

NSF/ENC-86003

PB86-18787

EVALUATION OF MITIGATION STRATEGIES FOR DISASTER EVENTS

February 17, 1986

NSF Grant CEE 8316567

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9072-101

REPORT DOCUMENTATION PAGE		1. REPORT NO. NSF/ENG-86003	2.	3. Recipient's Accession No. PB86 187877 IAS
4. Title and Subtitle Evaluation of Mitigation Strategies for Disaster Events			5. Report Date February 1986	
7. Author(s) P. Gordon; T. Banerges; L. Wingo			6.	
8. Performing Organization Name and Address University of Southern California School of Urban and Regional Planning Los Angeles, CA 90089-0042			9. Performing Organization Rept. No.	
10. Sponsoring Organization Name and Address Directorate for Engineering (ENG) National Science Foundation 1800 G Street, N.W. Washington, DC 20550			10. Project/Task/Work Unit No.	
			11. Contract(C) or Grant(G) No. (C) (G) CEE8316567	
			12. Type of Report & Period Covered	
15. Supplementary Notes			14.	
16. Abstract (Limit: 300 words) This research applies policy analysis to the problem of designing a policy to mitigate the costs to residents of Los Angeles of any one of several prospective major earthquakes threatening the Los Angeles basin. A model is developed in the form of an accounting framework that processes data inputs from a variety of sources and generates policy rankings as outputs. Viewed as an attempt to start closing the gap between formal research in the earthquake field and applied policy testing, this model answers questions, for each of a number of suggested policies, such as: (1) Who will benefit? (2) Who will bear the cost? and (3) Is the policy cost effective? Annualized per capita costs and benefits of expected damage reduction are forecast for income groups as well as for geographic districts.				
17. Document Analysis a. Descriptors				
Earthquakes		Earthquake warning systems	Economic analysis	
Disasters		Earthquake resistant structures	Computer programs	
Damage		Mathematical models	Land use	
Injuries		Urban planning	Taxes	
b. Identifiers/Open-Ended Terms				
c. CBATS Field/Group				
18. Availability Statement NTIS			19. Security Class (This Report)	20. No. of Pages
			20. Security Class (This Page)	21. Price

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CHAPTER I: INTRODUCTION

BACKGROUND, OVERVIEW, LIMITS OF THE STUDY

This research applies policy analysis to the problem of designing a policy to mitigate the costs to residents of Los Angeles of any one of several prospective major earthquakes threatening the Los Angeles basin. We develop a model in the form of an accounting framework that processes data inputs from a variety of sources and generates policy rankings as outputs. We view this model as an attempt to start closing the gap between formal research in the earthquake field and applied policy testing.

The model answers questions like the following: for each of a number of suggested policies, are the costs saved worth the cost of saving them? if so, which policy is the most efficient, offering 'the biggest bang for the buck'? who enjoy the benefits and who bear the costs from such policies?

This policy analysis imposes four technical responsibilities:

1. to identify a set of prospective policies;
2. to determine the effects of each on the anticipated level of damages;
3. to assess the worth of these effects; and
4. to estimate the costs to the community of carrying out those policies.

Meeting these responsibilities would make it possible to rank various policies in terms of their relative efficiencies. Yet, governments do not always adopt the 'best' policy; efficiency is not the only (nor even necessarily the dominant) criterion. Indeed, the most efficient policy choice may have undesirable distributional consequences. Most simply, we tend to accept the notion that a new policy should not 'worsen' the distribution of income.

This adds another responsibility to the policy analyst:

5. to assess the losses and gains each policy promises to impose on different groups in society

The consequences of policy for individuals is the proper concern for political representatives whose constituencies are defined by the spatial of political district boundaries. A policy's political feasibility depends on the likelihood that it can garner enough legislative votes to be enacted. Thus, treatment of political feasibility in our policy analysis entails yet another responsibility:

6. To determine the geographical distribution of affected groups among the political relevant domains.

In short, the policy analysis of earthquake mitigation strategies confronts us with six tasks.

Legislators and public officials have interests in the prospective consequences that would flow from the implementation of policies. Some plans and policies can be developed on a trial and error basis, allowing a government to 'learn' by its mistakes and correct them 'in

flight'. For others which have massive initial commitments of resources or lagged consequences that only become apparent after a number of years the learning mode is too costly. To substitute for the learning mode we have come to rely on models of conditional prediction. These models do not 'chose' a best policy; they simply present putative consequences flowing from different sets of assumptions to which valualational notions can be applied.

In this research we have developed a model of conditional prediction for policies proposed to mitigate the damages to be expected from the occurrence of any of the major seismic events predicted for Southern California. While we have no policy tools powerful enough to change the likelihood that one or more such events will occur, we can change the exposure of property and persons to damage or injury. Thus our modeling of the problem takes the estimated likelihoods of ground-shaking at various intensities at specific points in the city as given. It then asks the question: how will the expected losses and the costs of policy implementation vary by each of four prospective policies?

To answer these questions, we forecast the annualized per capita costs and benefits of expected damage reduction for income groups as well as for geographic districts.

LITERATURE REVIEW

Current research on the relationship of development patterns to post-earthquake recovery -- and of land use and development controls to seismic risk -- is still in its infancy. Until very recently the bulk of research on earthquake hazard mitigation was devoted solely to building-scale vulnerability, focusing on structural design and life-line engineering. Recent interest in research related to urban-scale vulnerability is encouraging. Most effort to date has focused on three topics:

1. application of seismic and geological information and micro-zonation techniques in land use planning;
2. case studies of anticipatory or post-earthquake land use planning and reconstruction efforts;
2. organizational and institutional contexts of local decision-making in disaster mitigation.

The first category of research draws on existing scientific and technical information related to local seismic phenomena and geological characteristics. Studies for seismic zonation in the San Francisco Bay area (Borchert, 1975; Brabb, 1979) provided important ground-work for subsequent studies by Blair and Spangle (1979), Erley and Kockelman (1981) and Jaffe et al. (1981). The United States Geological Survey has played a leading role in generating technical and scientific information related to microzonation studies and in disseminating this information. At least in the San Francisco Bay area, the seismic microzonation information made available by USGS has had some impact on city and county land use planning (Kockelman, 1980). What is lack-

ing is a ready methodology for microzonation analysis for specific sites and generalized performance characteristics of land use acceptable to sites of different risk levels.

Under the second topic, systematic inventory of post-earthquake losses and community impacts of recent earthquakes has begun. The Jones and Avgar (1977) protocol on the Rumania earthquake, the Lagorio and Mader (1981) and the Stratta et al. (1980) reports on the Campania-Basilicata earthquake, and the excellent paper on Managua by Kates et al. (1973) are noteworthy examples. Preliminary reports on the Coalinga quake are also available (French, et al., 1984).

While most of these studies focus on the community consequences of earthquake events and their implications for short-term and long-term planning, a monograph by Spangle et al. (1980) examines specifically the nature of reconstruction planning based on case studies of earthquake damaged cities (San Fernando, Anchorage, Valdez, Santa Rosa). It concludes that options for significant land use changes after an earthquake often are limited, which suggests that much greater emphasis should be placed on anticipatory, pre-earthquake policy measures for vulnerable areas.

Research on the organizational and institutional contexts of decision-making related to earthquake planning has barely begun. There have, however, been several efforts to develop land use planning techniques based on earthquake risk analysis. French and Isaacson (1984) describe a probabilistic approach to earthquake risk analysis which begins by estimating and mapping the expected level of ground motion and geological vulnerabilities of study areas. Damage estimates are

then made based on existing land use patterns or future scenarios. In another such damage and loss estimation system developed under the auspices of FEMA (Moore et al., 1985) several types of loss estimates were offered:

1. the expected physical damage caused by groundshaking;
2. the expected percent loss of function or usability;
3. the expected percent of population killed and injured;
4. expected losses due to collateral earthquake hazards such as ground-failure, fire, inundation.

This work is similar, and presumably related, to a technically sophisticated approach to earthquake damage evaluation developed by Rojahn, Sharpe, Nutt (1984). In that study a fairly detailed inventory was developed to estimate direct physical damage, deaths, injuries, loss of function and relocation time. Thus the study attempts not only to measure direct physical losses but also collateral social and economic losses. The investigation combines empirical with judgmental data; the authors emphasize that many of the damage estimates are based on a combination of judgmental and empirical inputs.

It appears from the previous studies of human loss prediction that results relied heavily on judgmental extrapolations of limited empirical data. Some of these efforts have tried to introduce different factors or weightings to reflect technological developments in building construction since much of the information is dated. (Whitman et al., 1975).

We believe that our approach to the problem is more elaborate than any preceding it. Munroe and Blair (1975) and Hirschberg et al. (1978)

offer formal models which assess the costs and benefits of various earthquake mitigation measures. Rose offers an input-output model framework to estimate the role of utility lifelines in sustaining economic activity after an earthquake. Ellison, et al. (1984) also developed an accounting system to measure long-term economic effects of an earthquake in the Charleston metropolitan area. They concluded that the economy was quite resilient and could absorb the impact.

While our model can be distinguished from these in many ways, the key difference is its ability to compute costs and benefits by income group as well as by geographic district. Our approach contains a distributional dimension and is able to answer questions of who pays and who benefits, as well as of total costs and total benefits. This capability makes it possible to achieve a basic purpose of policy analysis -- that efficiency and distributional patterns for a variety of policies and assumptions can be compared.

Several observations can be made from a review of some of the current literature. First, judgmental data and expert opinion continue to be major inputs to most models of earthquake estimation and risk analysis. Second, no state of the art method for estimating secondary and collateral damages (loss of life, injury, etc.) exhibit much reliability or validity. Third, despite some sophisticated modeling efforts, impacts on economic activities remain difficult to pin down. Fourth, although some work has begun in estimating damages to lifeline infrastructure, little information exists on physical damage beyond building types. In fact, there has been no attempt to estimate the damage to the non-building stock in the larger urban context.

Finally, there has been little work on the distributional -- by income class or by geographical areas -- effects of the earthquake hazard risk.

POLICIES TO BE EVALUATED

A novel aspect of the research was the need to design a set of prospective 'new' earthquake mitigation policies to test. The candidate policies needed to be prospective in the sense that they fell within the powers normally available to local governments in California, 'new' in the sense that they were not currently in place with consequences already detectable.

We begin by reviewing the general policy instruments available to California local governments. They may utilize their police powers (P) to regulate the conduct of private parties to protect the health, safety, welfare and morals of the community. They may tax personal and real property (and other objects as state constitution and the legislature permit) (T) to provide revenue for the operation of government and the implementation of policies. Local governments may also purchase (or acquire by eminent domain), own, and convey land and structures for proper public purposes (L). They may enter into contracts and agreements with private parties or other public agencies to carry out proper public purposes (C). They may construct public works (infrastructure) appropriate to the production of public goods and services (I). Finally, they may do all things otherwise necessary and proper to administer the activities appropriate to such matters (X). While these six sources of authority do not exhaust the menu of powers available to local government, they are the most important from the standpoint of their potential for earthquake loss mitigation.

Another policy taxonomy is embedded in the nature of damages attributable to a major seismic disturbance. Structures collapse or are rendered uninhabitable or useless for their purposes as a result of severe ground-shaking, subsidence or slides. These primary damages occur in the first few moments of the quake's onset. In the private sector many residences, commercial buildings and industrial structures will suffer damages depending in large part on their construction (S1). The losses here include the costs of repair (or site clearance and structure replacement). In the public sector it is the basic infrastructure of the community which is damaged -- streets, water delivery and impoundment systems, bridges and overpasses, power and communications lines, natural gas pipelines, sewerage systems and treatment plants, airport runways and traffic control systems, as well as hospitals, schools, fire and police stations (I1). The losses from such damage may take the form of severe deprivation and hardship visited upon the community from the loss of services dependent upon such infrastructure. Furthermore, the ultimate costs of replacing such facilities is a major dimension of these losses.

Collapsing structures can kill and injure large numbers of people (M1), depending on the moment the disaster strikes: early on a weekday afternoon schools, stores and places of employment will be crowded while the pre-dawn hours will find most at home in bed. The morbidity and mortality consequences of such disasters -- the human costs -- are what really engage public concern. The value of lives lost is difficult if not impossible to calculate: it would be virtually impossible to achieve any consensus on such a figure. The losses occasioned by

injury only begin with medical costs of recovery; they must also include incomes foregone and the more intangible losses from pain and suffering.

Also prominent among the losses would be those resulting from secondary damage to private property (S2), infrastructure (I2) and human beings (M2), caused by fire, looting, collapse of reservoirs, the exposure of structure contents to the elements -- all following the initial seismic shock.

Finally, large losses would result from the massive interruption of the local economy (E). Enormous stocks of capital are erased, the transportation of goods and services brought to a halt, the delivery of labor services impaired. The output of goods and services may be substantially reduced for weeks. An even more visible interruption in the economy may take the form of large numbers of homeless people who may become totally dependent on public support during the early stages of recovery.

It is beyond the scope of this project to address the full catalog of losses possible in a major earthquake in the Los Angeles area. Our attention was focused on direct and secondary impacts on the private building stock as well as on other elements of the built environment (elements of public 'infrastructure'). The shaded areas of figure I.1 denote the limits of the study.

FIGURE I.1: LIMITS OF THE STUDY

CAUSES OF LOSSES	shaking	interaction	fire	other
TYPES OF LOSSES				
buildings	shaded	shaded	shaded	
non-building stock	shaded	shaded	shaded	
life				
injury				
economic activities				

The systematic process of identifying policies for study began with the proposition that the mitigation of prospective loss from a major seismic event is an appropriate object of public policy for local governments in California, including the invocation of the police powers to protect the health, safety and welfare of the citizenry of Los Angeles. We were then led to examine each of these powers, raising the question each time: can this power support one or more activities which might have the effect of reducing the losses the community would suffer from a disaster event?

While the fundamental purpose of the power to tax is to raise revenue for the conduct of the affairs of government, its ability to create incentives to induce private individuals and firms to conduct themselves in manners consistent with public policy has often been exploited as a positive instrument of policy. With respect to the seismic threat, the induced behavior that would mitigate prospective losses might be manifested as private parties making structural improvements in structures they own to increase the ability to withstand seismic shaking. Accordingly, we designed a policy which would offer limited financial support to property owners to do things that are both in their interests and those of the general public:

1. offer a tax subsidy of up to 20% of current local property tax obligations to offset the costs to property owners of improving the ability of residential structures to withstand seismic shaking, subject to prior approval by the appropriate city agency (ta).

The regulatory authority available to local government under the implicit police powers of the state is the basis for negative incen-

tives with respect to private behavior inconsistent with the health, safety, welfare and morals of the community. To mitigate the losses from seismic events that authority could support regulations governing matters ranging from construction standards and zoning to the provision for the safety of structure occupants. Our review of the mitigatory potential of public regulations led us to add the following measures to the project's policy menu:

2. require retrofitting and upgrading of all nonconforming private buildings to current structural standards within five years (pa);
3. require every large private establishment in the city limits to file with the City a 'private agency crisis plan' to minimize injuries and loss of life at such sites in the event of a major earthquake (pb);
4. prohibit the storage of materials of prospectively hazardous to life and property within the city limits of Los Angeles (pc);
5. rezone to low-density uses all areas within the city limits threatened by slides, liquifaction, subsidence, or inundation (pd);
6. restrict the current occupancy of places of public assembly to reduce the exposure of individuals to death and injury in the event of a major earthquake (pe).

The eminent domain powers of local government would make it possible both to take private structures for direct public use and to take property which presented unusual hazards to the public in the event of an earthquake. To the list of possible policies, we add measures that would:

7. set up within two years 500 fully equipped emergency shelters around the city to take care of those homeless and displaced after major earthquakes (ea);
8. acquire and demolish nonconforming structures posing unusual threats to life and property (eb);

9. acquire land and structures in hazardous areas subject to liquefaction and sliding to prevent further development of such areas (ec).

The authority to use public funds to do all things reasonable and proper to implement proper public purposes provides the general powers that support much local public policy and is restrained only by the rule of reason. In the case of earthquake loss mitigation we would include in the list the following measures authorized by this local government power:

10. provide standby capacity or redundancy at vulnerable points in public infrastructure systems, such as roads, electric power distribution systems, sewerage, etc. (ia);
11. participate in joint powers agreements with other local jurisdictions in the metropolitan area to pool emergency powers, facilities, equipment and command responsibilities (ib);
12. conduct in schools, neighborhood centers and through the media public information programs with respect to the threats posed to life, limb and property by a major earthquake and to individual actions to minimize its effects on them, their families, employees and other charges (ic).

The power to enter into contracts allows the local government to solicit commitment from private (or public) agencies and groups to carry out provisions of public policy not easily achievable under other powers. For earthquake loss mitigation, we add the following measure to policy list:

13. offer public subsidies to firms and other private agencies to enter into emergency response training programs for employees, students and regular clients (ca).

Administrative powers offer a broad grant of authority to local governments. These powers can provide substantive support for programs in terms of planning, coordination and consultation. Several such

internal measures are appropriate to our policy menu:

14. join with the California Department of Transportation and other local transportation agencies in the area to create a regional emergency transportation plan (xa);
15. establish a working interdepartmental coordinating committee to integrate plans for emergency public health and safety services (xb);
16. purchase insurance on local public buildings and structures against loss attributable to earthquakes (xc);
17. develop a standby emergency communications net among key administrators within the city and among local governments in Southern California (xd);
18. require all departments and agencies of the City to submit to the Mayor within six months emergency management plans assuming a major earthquake within ten years (xe);
19. establish a crisis command system within the City government with appropriate standby emergency powers to deal with the aftermath of a major earthquake (xf).

Our approach to generating a policy menu of nineteen elements was to take the policies generically rather than specifically, even though the expression of them was made specific to give panelists some sense of concreteness. Figure I.2 places these policies in a matrix which relates them to the policy taxonomies discussed at the outset.

FIGURE I.2: POLICIES BY SOURCE OF AUTHORITY AND FORM OF LOSS ADDRESSED

(see text for key to symbols)

FORM OF LOSS:		POWERS:					
		p	t	e	i	c	x
(included)							
S1	a, d	a	b, c				
S2	c		b				
I1	a			a		a, b, c	
I2				a		a, b	
(omitted)							
M1	b, c		a, b	c	a	b	
M2	b, c		a, b	a, c	a	b	
E				b		d, e, f	

The first task for panelists engaged in the policy review exercise was to help us distinguish policies-in-effect from the list in order to identify candidate policies from which we would chose those to test in the policy model. The concept of 'policy-in-effect' is more elusive than at first appears. To begin, the term 'policy' has two common meanings:

1. a policy may be an authoritative statement about intentions to pursue a certain course of action;
2. a policy may also mean a public purpose toward achievement of which activities have been organized and resources made available.

Our approach was to invoke the judgments of a panel of experts on seismic risk in Southern California. We invited a number of officials of state and local government agencies and private firms who were responsible for rendering advice to their institutions on earthquake preparedness to serve on a Delphi panel to explore this and other relevant matters. Sixteen panelists served throughout this exercise.

To help us identify the City of Los Angeles' policies-in-effect to mitigate losses from earthquakes, we asked panelists to examine the policy menu to give us their understanding of the status of each using the key in Figure I.3.

FIGURE 1.3: POLICY STATUS KEY

10. I know that this (or a very similar) policy has been adopted and is being aggressively implemented.
9. I know that this (or a very similar) policy has been adopted and is being routinely implemented.
8. I know that this (or a very similar) policy is on the books but is not being currently implemented.
7. While such a policy is currently in effect, I expect it to be rescinded or substantially modified within the next year or two.
6. While I do not know for sure, my impression is that such a policy is in effect.
5. I know that such a policy has been adopted.
4. I know that a policy very similar to this is currently being seriously considered and there is a strong likelihood that it will be adopted within the next year.
3. I know that such a policy has been seriously entertained by responsible authorities within the last two or three years and rejected.
2. To the best of my knowledge such a policy has not been seriously considered for adoption in recent years.
1. In my judgment such a policy is never likely to be a serious candidate for adoption.
0. I have insufficient information to make any judgment about the policy status of this item.

Answers to these questions generated a set of frequencies by categories which reflected:

1. the nature of the respondent's information;
2. his or her sense of certainty;
3. his or her conclusion about whether the policy was in effect.

Since these categories were not so structured as to permit any direct scaling of the answers, we employed a set of weights designed: [1] to distinguish judgments that a policy was in effect (+) from those for which that was not the case (-); and [2] to measure the conviction of the respondent about his or her answer. This exercise resulted in an interpretive scale of positive values $0 < a < 10$ and negative values $0 > b > -10$. From this we defined a measure of the strength of conviction (S), which was simply the sum of the absolute values of the positive and negative scores:

$$0 < S = |a| + |b| < 10$$

Low values of S represented uncertainty and lack of conviction on the part of the panel as a whole. We then defined a measure (I) of the degree to which positive responses dominated the panel's collective opinion. That is, responses asserting that the policy element was, indeed, part of the City's earthquake damage mitigation policy were defined:

$$I = |a|/S.$$

This model suggests that if all of our policy elements were clearly described, if all of the experts were perfectly informed and if there

was no ambiguity in the facts about policy elements, we would expect the following results:

$$I(j) = 1, \quad S(j) = 10, \text{ and}$$

$$I(k) = 0, \quad S(k) = 10$$

where, j designates a member of the set (m) of policies-in-effect;
 k designates a member of the set ($19-m$) of policies-not-in-effect.

At the other extreme, the experts could be ignorant but honest, the facts difficult to ascertain and the statements of policy elements ambiguous. We would then expect to find:

$$I(j) = 0, \quad S(j) = 0, \text{ and}$$

$$I(k) = 0, \quad S(k) = 0.$$

What we actually discovered in the results is not the discontinuity between J and K that the ideal case would have us expect, nor the collapse of the distinction between J and K, but an irregular gradient between what appear to be 'j-cases' and 'k-cases'. J and K are taken to consist of six and eight policy elements respectively, while the remaining five cases fall between J and K. These five cases reveal the effects of ignorance on the part of the panel, ambiguity in the actual cases or a failure in our description of the policies. The results of this analysis are in column 1 of Table I.1.

TABLE I.1: ANALYSIS OF POLICY DELPHI*

Policy element	1 Is policy NOT in effect? (i)		2 Is policy Feasible? (f)		3 Is Adoption Urgent? (u)		4 A POLICY (Policy OPTIM? Object Reasons?		(See Key Below)
	01 0 to 1 rank	02 0 to 10 rank	03 0 to 10 rank	04 0 to 10 rank	05 0 to 10 rank	06 0 to 10 rank	07 0 to 10 rank	08 0 to 10 rank	
1 20% Tax Credit to Earth-quake-proof Residences	Yes .81	2.5	No 3.5	16.0	No 4.3	16.0	NO f,u	S	
2 Emergency Transportation Plan for Region	? .25	11.0	Yes 6.6	5.0	Yes 6.4	6.5	YES e	E	
3 Upgrade National Seismic Structures	No .19	16.0	No 1.6	16.5	Yes 6.0	6.5	NO e,f	S	
4 Establish Interdepartmental Coordinating Committee	No .00	10.0	Yes 6.2	7.0	Yes 6.4	6.5	NO e	H,E	
5 Acquire Property in Hazardous Areas	Yes 1.00	1.0	No 1.6	16.5	No 2.3	19.0	NO f,u	S	
6 Joint Powers Agreements among Local Govts.	No .25	12.5	Yes 6.5	6.0	Yes 6.0	10.0	NO e	E	
7 Acquire Insurance for Public Buildings	? .50	7.5	? 4.7	11.0	No 2.6	16.0	NO e,f,u	E	
8 Prohibit Storage of Hazardous Materials	Yes .81	3.5	No 4.0	12.0	No 4.0	12.0	NO f,u	S,H	
9 Restrict Occupancy of Public Places	? .50	7.5	No 1.2	19.0	No 2.0	17.0	NO e,f,u	H	
10 Standby Capacity in Infrastructure	No .25	12.5	? 4.4	12.0	Yes 6.4	6.5	NO e,f	E	
11 Subsidize Emergency Response Training	Yes .80	2.0	Yes 2.6	8.0	No 4.2	15.0	NO u	H	
12 Emergency Communications Net	No .11	16.0	Yes 6.9	2.5	Yes 7.2	2.5	NO e	E	
13 Rezone Hazardous Areas to Low Density	Yes .83	2.0	No 2.5	15.0	No 2.0	18.0	NO f,u	S	
14 Condemn & Demolish Nonconforming Bldgs.	? .56	9.0	No 1.3	18.0	No 5.2	11.0	NO e,f,u	S	
15 City Agencies to Draw Up Emergency Management Plans	No .12	15.0	Yes 6.9	2.5	Yes 7.0	4.0	NO e	E	
16 Require Crisis Response Plans of Private Firms	Yes .72	6.0	? 4.9	9.0	Yes 6.2	9.0	YES e	H	
17 Public Information and Instruction Program	No .00	18.0	Yes 7.4	1.0	Yes 7.2	2.5	NO e	H	
18 Set Up a Crisis Command Center in City govt.	No .00	18.0	? 4.6	10.0	Yes 7.9	1.0	NO e,f	E	
19 System of Emergency Shelters	? .52	10.0	? 4.0	10.0	No 4.5	12.0	NO e,f,u	S	

3 Benchmark Policies 01 = Score on Question 1 S = Structures
 02 = Score on Question 2 H = Life and Limb
 03 = Mean Score on Questions 4, 6, and 9. E = Recovery

*Readers should beware of the convention in the table that poses the interpretive question about policies-in-effect in the negative: policies-in-effect are identified by a NO answer to the question, 'Is policy NOT in effect?' By our rating system, perfect panel consensus on the judgment that a policy is in effect is reflected by a score of zero or a NO answer; similarly, perfect consensus that a policy is not in effect is represented by a score of 1.00 and a YES answer.

Policies on which the panel was unable to agree were defined as those with scores between 0.35 and 0.65, and identified with a question-mark. Of the nineteen policies on our menu, the panel concluded that eight were already in effect. These constituted our policy benchmark. Of the remainder, the panel concluded that six were not now in effect and that five more might not be, leaving a modified policy menu of eleven candidate policies.

Columns 3 and 4 of Table I.1 set out the panel's assessment of the feasibility and the urgency of the various policies. Some of the conclusions appear to be:

1. eight of the ten policies rated 'urgent' by the panel are benchmark policies and all of the benchmark policies were considered to be urgent;
2. only one of the six clear candidate policies was construed to be urgent by the panel; of the remaining five, four were considered infeasible;
3. only one of the five uncertain candidate policies was considered to be feasible;

Overall the panel's response seemed to tell us that the City of Los Angeles has done a good job of designing and adopting an overall earthquake mitigation policy. It would, by inference, recommend only two additional policies from our policy menu as being urgent and feasible:

2. an emergency transportation plan for the region;
16. a requirement that large private firms prepare crisis response plans to be filed with the City.

Our review of the panel's judgments leads us to the conclusion that it is conservative with respect to prospective mitigation policies. Of

the six policies listed in Figure 1.2 that required governmental intervention in the exercise of private property rights, five were found to be infeasible and deferrable. The sixth, 'upgrading nonconforming buildings', was found to be urgent if infeasible (and probably already in effect).

It was clear that we could not limit ourselves to those policies which met the three-fold test:

1. is the policy already in effect?
2. is the policy feasible?
3. is the adoption of the policy urgent?

No policy meets that test beyond any doubt. Even stretching things would only have added #2 and #16. These are basically planning activities whose consequences for damage mitigation strategies would be difficult to specify.

Once we made the decision to confine this part of the research effort to policies which would have an effect on the prospective damage to structures, our attention was drawn to the five policies which dealt with structures and which were not in effect ('S' in column 5 of Table 1.1). From these we selected three quite different policies to test:

1. 20% tax credit to earthquake-proof private residential structures;
8. prohibition storage of hazardous materials within city limits;
14. condemnation and demolition by the City of buildings nonconforming with respect to capacity to withstand seismic shaking.

In addition, we chose a policy which the panel took to be in effect:

10. adding standby capacity to key infrastructure systems.

We added this policy because we found that we had not conveyed effectively to the panel what the policy should include and because we wished to test our policy model on a more 'conventional' policy.

These four policies play a major role in the exercises discussed in the following chapters and, of course, in simulations by the policy model.

CHAPTER II: DATA INPUTS

A major contribution of our model is to be found in its accounting framework which can process diverse kinds of available information to develop outputs having a substantial policy interest. From the beginning we hypothesized that the data to implement such a model are available, even if from diverse and non-traditional sources. This prototype application of the model to the city of Los Angeles was an attempt to test that hypothesis. We can now report that enough information does exist and can be developed to implement the model. This chapter reviews the data sources as well as the data manipulations that implemented the model.

For the policy model to carry out its functions, its accounts require information on such matters as:

1. the likelihood that major seismic events will occur within the regional fault structure within the near future;
2. the ground-shaking that any such event(s) will generate in all parts of the city;
3. the effects of ground-shaking on all significant classes of structures (privately owned and others) within the city;
4. valuation of various levels of damage to various classes of structures;
5. distributions of structure classes among various parts of the city;
6. the effects of alternative policies on the expectation of loss by the various classes of structures;

7. the costs to the taxpayers of carrying out such policies;
8. the income distribution of households in the various parts of the city;
9. the incidence of policy costs on the taxpaying public, by income class.

Such a data specification can only be satisfied by bringing into the model data from a multitude of sources. In some cases, the data were readily available in a form consistent with the model. The City's LUPAMS file, for example, presented us with information about the distribution of structures in the various planning districts. Other information had to be inferred from data from more general sources such as that about non-private structure values and the incidence of various taxes on income groups in the taxpayer population.

Some kinds of information, however, were not directly derivable from data sources we had searched out, nor could they be easily derived from conventional information available to us. Algermissen et al. had generated the well-known structure-type damage curves by MMI measures. To work within our model, these curves had to be adjusted to basic assumptions implied by the model and extended to account for damage to public facilities and infrastructure. We also required general information about the costs, status and usefulness of various policy options; these would have been laborious and expensive to estimate, item by item. The need for information having these characteristics was satisfied by employing judgmental inputs, disciplined by the methodology of the Delphi technique.

There is nothing arcane about the way in which individuals or organizations deal with uncertainty when they have no easy way to reduce

that uncertainty: we ask those who are in a better position than we are -- the 'experts' -- to give us their best informed guess. We recognize that the experts probably disagree to some degree on the matters at hand. Given some such 'dissensus' among experts, most of us would like to think that judgments close to the center of the array of opinion are somehow more credible than those at the extremes, and thus we would like 'our' expert to be in the middle. But how do we know that until he or she has rendered judgment? We don't. So, to protect ourselves from ending out on a limb, we elect to consult either the whole universe of experts or a carefully selected panel thereof. We propose to consider that consensual values elicited therefrom are 'truer' than the judgment of any individual. We could, of course, assemble the panel in a hall and ask them to agree on the appropriate answers to the questions we put to them. These could be discussed, debated and put to the vote. If the vote in favor were unanimous, then the content of that resolution would represent the judgment of experts. But what if the vote were 56 to 44? The notion of majority rule has some relevance to social preferences but does very little to validate the 'truth' or establish fact, per se. In both of these cases polemic and personal force can count for as much as reasoned discourse.

The Delphi technique offers a means of minimizing these hazards. It permits us to interrogate each of the members of a 'panel' of experts in such a way that the judgments of each are treated confidentially but are disciplined by the knowledge that the other members of the panel will respond to the same questions. Typically, the analyst will aggregate the panel's individual responses into a two-dimensional

'panel response', in terms of the central value of the panel's answers and a measure of dispersal (dissensus). In the conventional Delphi procedure such summary statistics are shared with each panelist, who is given an opportunity to review and amend his or her answer in light of this information. This modified information is analyzed and the new central value construed as the panel's answer, especially if smaller measures of dispersion suggest some degree of 'convergence' toward a consensual value by panelists.

Needless to say, the outcome of such a procedure depends heavily on the composition of the panel and on the breadth and complexity of the matter of the inquiry, factors which are not unrelated: the narrower the question the easier it is to specify the relevant domain of expertise. Not only that, but the more specialized the expertise appropriate to a collective best judgment and the greater the degree of consensus that one would expect to find. On the other hand, broad, messy questions often involve several kinds of specialized experience and make appropriate a more heterogeneous panel and a lower expectation of consensus. What becomes important with such questions is that the panel be larger than one dealing with narrow questions and be selected without obvious bias.

Our policy questions were broad and messy, so a small, professionally homogeneous panel would not have served us well. We wanted judgmental information about the current status of policies, about their costs and benefits and the incidences of same, about their feasibility, and about their urgencies. Engineers, planners, geologists, economists, and public officials, all have something to say on matters dealing

with the mitigation of earthquake damages, and the authority of no one can be said to preempt the authority of the others.

Accordingly, we first compiled a master list of approximately sixty experts in public and private sectors in Southern California identified by inquiries to the relevant agencies of local governments in Southern California and by our own experience with the field. We then asked each member on the master list independently to nominate a slate of experts on local earthquake matters, and the master list was reorganized by the frequency with which individuals were identified in the nomination exercise. We invited the thirty top-ranked experts to serve on our policy Delphi panel; of these approximately twenty accepted, and sixteen of these participated throughout the policy inquiry. We could perceive nothing in this panel selection process that would 'stack the deck' with respect to answers to any of our inquiries. At the same time we have no reason to believe that the panel was an ideal sample of the universe of the sixty experts on the master list.*

The form of the policy questions posed to the panel provided devices by which panelists could disqualify themselves from answering specialized questions on which they themselves wished to deny any imputation of expertise (the '0-option'). In assessing the private (nontaxpayer) costs panelists availed themselves of this option in approximately one-third (32.2%) of judgments sought, while in making judgments about

*This procedure was modified somewhat with respect to the damage curve inquiries (below). The panel was enriched by adding several engineers and geologists from other parts of the state to the group.

public (taxpayer-borne) costs less than a fourth of the answers were 0. We accepted each panelist's implied judgment that he or she themselves as expert on any matter on which they submitted an answer > 1. We are not protected from the prospect that some experts may overestimate their expertise, or that some entering zero may underestimate themselves and so deprive us of the value of their judgment.

The Delphi technique produces judgments not 'facts', but where 'facts' cannot be established such judgments are the best informed guesses about the facts that expertise in a particular matter has to offer. Only history or the emergence of superior information can preempt them.

URBAN 'CONTEXTS'

Building damages usually attract much of the attention after major earthquakes. Yet the ravages of a major earthquake are not limited to buildings; they include significant damages to the rest of the physical city. While there are empirically based methods for predicting damages by different building types for different magnitudes of earthquake, no such methods exist for estimating the residual non-building damage. In this study we are interested in both the building damage and the residual non-building damage since the policy evaluation model is predicated on damage estimates for the entire physical city. We needed, then, to develop damage loss curves similar to those existing for building types for the rest of the urban physical stock -- underpasses, retaining walls, signs and billboards, transmission towers, streets, pavements, utilities, street furnitures, and the like.

Previous damage loss studies (Algermissen et al., 1978; Spangle and Degenkolb, 1984) have almost exclusively focused on building types, begging the question of estimating the residual damages to the physical city. Yet, there is ample evidence that non-building stock damages could be substantial.

During the 1971 San Fernando earthquake, for example, damage to one of the dams reduced its effective height by about 30 feet; displacement along the fault line sheared or damaged underground water, sewer and gas pipes; forty-two bridges or overpasses were damaged of which five collapsed. Although dollar costs are not known for many of these non-

building stock damages, it is known that damage to the highway and road system alone amounted to over \$36 million dollars.

We have found no damage ratios for the non-building stock. There are several reasons for this. It is possible that, given the miscellaneous nature of the residual damage, estimates are never quite complete or systematic. Many of these items fall in the public sector so that costs of repair or replacement are likely to be distributed among several agencies. Unlike individual buildings, these urban hardware items may not always be covered by insurance, and hence may not be subject to damage assessment. Furthermore, many of these elements are not as discrete as individual buildings, or are too numerous to be categorized under manageable headings. In other words, the absence of damage ratios for the non-building stock items can be attributed to both inadequate data and an absence of methodology.

In the absence of objectively derived damage curves for non-building stock losses, we adapted the Delphi methodology to generate damage curves for such losses from expert judgments. It was obvious at the outset that some form of abstraction was necessary to describe the variable features of the physical city. The critical assumption was that given some plausible global descriptions of urban settings, panelists would be able to estimate the overall residual environmental damage by such types of urban settings. What was needed, then, was an operational typology of urban settings similar to the widely used building typologies.

The development of such a conceptual model was considerably facilitated by the availability of a citywide data file (LUPAMS; Land Use

and Planning Analysis and Management Systems) from the Los Angeles City Planning Department. The LUPAMS file is a master data file which compiles information from records of the Los Angeles County Assessor's file, the City Clerk's office, the City Planning Department and the Community Analysis Bureau. It includes land use, property values, zoning, jurisdictional and city services data for most of the over half-million parcels in the city of Los Angeles. As shown in Table II.1, LUPAMS offered enormous amounts of physical and wealth data, useful not only in estimating value at risk, but also in constructing physical descriptors of possible classes of urban areas.

In Figure II.1 we show our general approach to develop the typology of urban settings or contexts. The methods for classification were statistical, based on a set of relevant variables taken from the LUPAMS file. We chose census tracts rather than individual parcels as the appropriate unit of analysis since the state of parcel level data were enormous. This required creation of a new file of aggregated parcel data by census tracts for selected variables.

There was another justification for using census tract-level data. For all of its richness, the LUPAMS data file has one deficiency -- it does not include any assessment-related information on parcels under public ownership or other parcels with tax-exempt status. At the census tract level problems of missing data would be reduced.

URBAN CONTEXT CLASSIFICATION

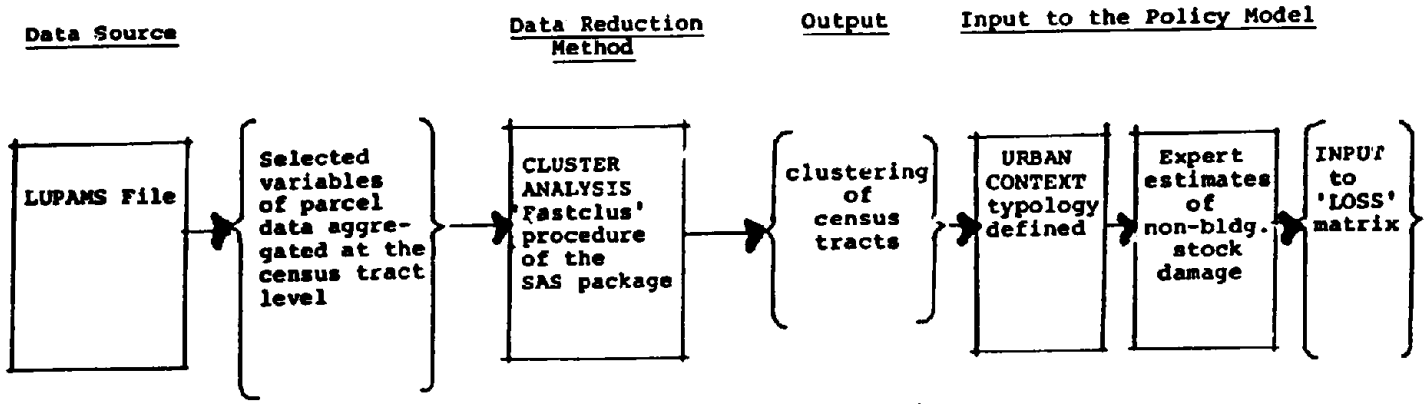


FIGURE II.1: PROCEDURE FOR DEVELOPING TYPOLOGY OF URBAN CONTEXTS

TABLE II.1: GENERAL ITEMS OF THE LUPAMS FILE

Los Angeles County Tax Assessor's Secured File

1. Agency classification number
2. Assessor' mapbook, page and parcel number
3. "Market" land value
4. "Market" improvement value
5. Personal property value and exemption
6. Inventory value and exemption
7. Fixture value and exemption
8. Other exemptions -- real estate and homeowner
9. Tax code area
10. Owners' name
11. Owners' mailing address
12. Parcel address (Situs)
13. Zoning
14. Assessor's property use classification
15. Building data line items (first five improvements)
 - (a) Design type
 - (b) Construction class and building shape
 - (c) Year built
 - (d) Number of units
 - (e) Number of bedrooms
 - (f) Number of bathrooms
 - (g) Gross floor area

City Clerk's Office

1. Parcel area
2. Census Tract
3. Census Block
4. Council District
5. Community Number
6. SCAG Land Use Code

Planning Department

1. Planned land use
2. Zoning corrections
3. Public owned land/land use
4. Home ownership code
5. Condominium indicators

Community Analysis Bureau

1. Police reporting district
2. Fire engine first-in-district
3. Corrected Dwelling Unit counts
4. Weed abatement key

Table II.2 lists the variables which were selected from the LUPAMS file. The parcel-level data were aggregated to the census tract level for each of these variables. Furthermore, population and acreage data for census tracts were available from 1980 Census files and merged with the reduced LUPAMS file for the 734 census tracts which lie within the city boundaries. Several new variables using derived measures were also created; these are explained in the note accompanying Table II.2.

TABLE II 2: LIST OF VARIABLES USED IN CLUSTER ANALYSIS

- Parcel Area (PCL-AREA)
- Building Improvement Area (BDGIMPRV) *
- Class "A" Structures (BCA)
- Class "B" Structures (BCB) *
- Class "C" Structures (BCC) *
- Class "D" Structures (BCD) *
- Class "S" Structures (BCS) *
- Building Quality 1 (BDC1)
- Building Quality 2 (BDC2)
- Building Quality 3 (BDC3)
- Building Quality 4 (BDC4)
- Building Quality 5 (BDC5)
- Building Quality 6 (BDC6)
- Building Quality 7 (BDC7)
- Building Quality 8 (BDC8)
- Residential Land Use (UC0) *
- Commercial Land Use I (UC1) *
- Commercial Land Use II (UC2) *
- Industrial Land Use (UC3) *
- Irrigated Farm Land Use (UC4)
- Dry Farm Land Use (UC5)
- Recreational Land Use (UC6)
- Institutional Land Use (UC7)
- Miscellaneous Land Use (UC8)
- Structures Built Before 1935 (YR1T34) *
- Structures Built Between 1934 and 1973 (YR35T73) *
- Structures Built After 1973 (YR74T85)
- Property Value (VALPAR) *
- Floor Area Ratio (FAR) *
- 1980 Census Gross Density (DEN80) *
- Year Built (BYEAR)

NOTE: (a) Property Value and Floor Area Ratio are derived measures. They were computed as follows: Property Value = (Market Land Value + Market Improvement Value)/Parcel Area. Floor Area Ratio = Building Improvement Area/Parcel Area.

(b) Asterisks indicate variables with R-squared values greater than .25 (See Table II.3, Results of the FASTCLUS procedure). These variables were used to name the six operational clusters (See Table II.4 and Figures II.3 and II.4)

Of the available statistical classification techniques (Figure II.1), cluster analysis is the most appropriate. It has emerged as a popular strategy for taxonomic analysis in such widely diverse fields as botany, education and clinical psychology. Recently it has been applied successfully in urban social area analysis (Kendig, 1976) using demographic data at the census tract level. We used the 'FASTCLUS' procedure of the SAS (Statistical Analysis System) package, which is best suited to large-scale files.

FASTCLUS requires a priori specification of the number of clusters desired. Our experiment with several different specifications indicated the categories at the level of six clusters were consistent with our operational point of view.

Results from the six-cluster FASTCLUS calculation are reported in Table II.3.

The table shows that the bulk of the census tracts were included in the two main clusters -- one and six -- with memberships of 470 and 210 tracts, respectively. The next largest cluster included 31 census tracts. The remaining clusters are much smaller; the fourth cluster only contains one census tract. It should also be noted that only about half of the variables played any significant role in differentiating cluster characteristics. Only these variables were used to name the clusters.

FASTCLUS PROCEDURE

TABLE II.3: RESULTS FROM THE FASTCLUS CALCULATION

CLUSTER SUMMARY

CLUSTER	MEMBERS	RMS ST DEV	MAX DISTANCE FROM SEED
1	470	931.7771	9088.806
2	3	1828.08	10891.82
3	31	1538.818	18141.56
4	1		0
5	17	848.003	13814.73
6	210	732.8804	12200.27

STATISTICS FOR VARIABLES

VARIABLE	TOTAL STD	WITHIN STD	R-SQUARED	VAR RATIO
PCL_AREA	418.68904	411.43814	0.04553	0.04771
BDGIMPRV	2840.78173	1823.18568	0.73396	2.75322
BCA	3.07858	3.03073	0.02560	0.03891
BCB	2.84428	2.85214	0.20037	0.29087
BCC	9.75802	6.68844	0.53349	1.14360
BCD	12.28822	8.13607	0.56511	1.28848
BCE	1.41612	1.21853	0.25345	0.35767
BDC2	3.88176	3.80627	0.02887	0.03048
BDC4	19.31787	17.88037	0.14916	0.17931
BDC5	21.88786	21.85642	0.02683	0.02792
BDC6	21.38517	20.88555	0.05004	0.08268
BDC7	18.13761	18.18952	0.1280	0.11458
BDC1	3.88728	3.88857	0.00618	0.00622
BDC2	0.57838	0.57967	0.00242	0.00243
BDC8	11.22650	11.10429	0.02825	0.02817
UC0	16.15382	12.06253	0.44621	0.80574
UC4V	7.15481	7.15444	0.00697	0.00702
UC1	5.18103	4.06904	0.38977	0.63872
UC2	4.80321	3.65743	0.42418	0.73664
UC3	4.81028	6.98905	0.38720	0.62186
UC4	0.02778	0.02774	0.01017	0.01027
UC5	0.18680	0.18685	0.00526	0.00528
UC6	0.62187	0.61727	0.02180	0.02228
UC7	0.75650	0.68648	0.15819	0.18792
UC8	2.36783	2.22122	0.12580	0.14291
YR1734	32.31824	27.83834	0.36214	0.39711
YR2572	30.81482	26.63240	0.26242	0.28762
YR7478	10.31600	10.27568	0.01489	0.01480
VALPAR	1035.37033	810.84195	0.75814	3.12456
FAR	10.83754	6.68642	0.62006	1.71785
DEN80	7584.20281	3485.27161	0.78026	3.76790
BYEAR	13.07974	11.53628	0.22728	0.28431
OVER-ALL	1451.52368	682.40020	0.78049	3.55965

APPROXIMATE EXPECTED OVER-ALL R-SQUARED = (0.8730) CUBIC CLUSTERING CRITERION = -17.3854
 WARNING: THESE VALUES ARE INVALID IF VARIABLES ARE CORRELATED.

FASTCLUS PROCEDURE

CLUSTER MEANS

CLUSTER	PCL_AREA	BDGIMPRV	BCA	BCB	BCC	BCD	BCE	BDC3	BDC4	BDC5	BDC6
1	357.48	2058.47	0.33	0.15	2.30	96.73	0.43	1.48	8.50	27.26	30.82
2	292.85	30422.26	8.32	8.77	28.27	53.31	2.69	1.48	18.76	15.88	28.80
3	284.20	5861.37	0.18	0.58	10.21	88.44	0.37	0.71	10.83	32.07	30.51
4	428.00	32888.60	3.45	24.14	5.17	65.82	1.72	0.00	24.14	24.48	6.90
5	888.03	8283.67	1.88	8.78	47.84	38.14	8.15	1.28	8.44	38.83	13.00
6	183.82	2700.84	0.19	0.13	2.53	96.68	0.41	2.88	25.24	35.88	21.28

CLUSTER	BDC7	BDC1	BDC2	BDC8	UC0	UC4V	UC1	UC2	UC3	UC4	UC5
1	19.75	1.12	0.11	6.68	80.86	5.47	2.81	1.81	2.45	0.01	0.03
2	4.82	0.20	0.00	1.10	26.30	2.18	28.65	27.27	3.12	0.00	0.00
3	7.15	0.47	0.03	8.20	77.87	4.06	10.73	6.53	2.78	0.00	0.00
4	1.72	0.00	0.00	0.00	41.18	1.18	18.82	20.98	2.35	0.00	0.00
5	5.41	0.81	0.18	8.45	26.48	3.80	8.33	17.40	28.80	0.01	0.00
6	7.18	0.47	0.15	3.92	88.16	4.41	6.80	3.82	1.38	0.00	0.00

CLUSTER	UC6	UC7	UC8	YR1734	YR2572	YR7478	VALPAR	FAR	DEN80	BYEAR
1	0.10	0.43	1.41	22.15	68.79	8.06	425.06	8.28	6547.58	48.18
2	1.22	1.64	1.71	43.28	48.49	7.23	5183.45	112.13	14248.67	41.10
3	0.18	1.22	0.61	81.39	39.48	8.20	1182.18	39.71	34846.13	39.02
4	0.00	2.35	4.71	28.41	38.71	30.88	23087.88	76.82	1463.00	52.87
5	0.18	0.68	6.80	28.06	62.82	8.12	780.47	22.84	4836.26	47.16
6	0.21	0.88	0.85	58.38	25.41	6.21	137.82	14.88	16277.03	34.45

CLUSTER STANDARD DEVIATIONS

CLUSTER	PCL_AREA	BDGIMPRV	BCA	BCB	BCC	BCD	BCE	BDC3	BDC4	BDC5	BDC6
1	488.208	817.374	3.527	1.222	4.221	8.284	0.821	3.678	14.710	22.480	21.810
2	71.820	8393.021	8.012	27.219	20.628	25.881	2.371	2.577	17.258	7.866	32.658
3	288.818	3192.688	0.428	1.027	16.184	17.684	1.168	1.312	18.148	21.221	18.256
4											
5	401.074	4380.584	4.747	15.852	27.847	30.000	8.888	3.073	10.670	22.488	11.327
6	115.484	1718.651	1.220	0.817	3.688	4.558	0.810	4.688	23.784	18.828	18.284

CLUSTER	BDC7	BDC1	BDC2	BDC8	UC0	UC4V	UC1	UC2	UC3	UC4	UC5
1	20.886	4.541	0.438	12.448	11.174	8.318	3.045	2.444	6.644	0.034	0.232
2	2.711	0.618	0.000	1.218	24.801	0.772	16.738	8.518	4.178	0.000	0.000
3	8.408	2.408	0.081	11.189	17.974	3.282	6.188	8.184	11.412	0.000	0.000
4											
5	8.031	1.160	0.483	18.285	28.921	2.610	9.184	18.880	24.820	0.024	0.000
6	13.142	2.208	0.848	8.077	10.218	4.508	4.497	3.968	2.488	0.000	0.017

In Table II.4 we present the distinguishing characteristics of the six clusters in descriptive fashion using such terms as 'few' or 'many', 'small', 'moderate', 'large' to make it easier for the reader to follow the variable profiles of the six clusters. These descriptors simply highlight the major features or tendencies among the clusters, and represent the mean cluster values of the fifteen critical variables used in their identification.

It is evident from these cluster summaries and the map of Los Angeles shown in Figure II.2, that the first cluster which includes more than half of all census tracts in the city, essentially represents the almost ubiquitous low-density single family residential areas that typify much of Los Angeles' suburban 'sprawl'. In the areas characterized by this cluster more than two-thirds of the buildings are likely to have been built after the Long Beach earthquake. About a fifth of the buildings are likely to be of the pre-1934 era. Relatively few structures are likely to have been built since the San Fernando earthquake. Over ninety percent of these areas are in residential use. With a gross density of ten persons per acre, this cluster can be best described as Low-Density Residential Neighborhood.

The second cluster contains only three census tracts and represents Medium-Density, Built-Up, Mixed-Use Districts, typically found in the inner city areas. This setting is characterized by older buildings, about two-fifths of which were built before 1934. Most of the remaining building stock were built between 1934 and 1973. About a third of

TABLE II.4: SUMMARY OF SIX CLUSTERS

Variables	Cluster Number					
	1	2	3	4	5	6
Suggested name	Low density residential neighborhood	Medium density built-up mixed use district	Older high density transitional neighborhood	High rise high value business district	Mixed industrial commercial district	Older medium density mixed neighborhood
Year Built	Late forties	Early forties	Late thirties	Early fifties	Late forties	Mid thirties
Value of Improvements	small	large	moderate	large	moderate	small
Reinforced concrete fireproof structures	----	few	----	many	few	----
Masonry, concrete brick structures	----	some	some	----	many	----
Wood or wood & steel structures	mostly	some	many	some	some	many
Specialized other types of bldgs.	----	few	----	----	some	----
Residential use	mostly	small amt	mostly	some	few	mostly
Stores, offices, shopping centers, hotels, motels	----	many	some	some	some	few
Restaurants, banks wholesale, parking	----	many	few	many	many	few
Industrial, MFG, warehousing, etc.	----	few	few	----	many	----
Built before 1934	----	many 43%	mostly 51%	some	some	mostly 35%
Built between 1934-73	mostly 70%	many 36%	some	some	mostly 33%	some
Average property value (land and improvement)	low	moderate	moderate	very high	low	low
Floor area ratio	low	high	moderate	high	moderate	low
1980 Census gross density	18	23	14	2	7	21

FIGURE II.2 City of Los Angeles

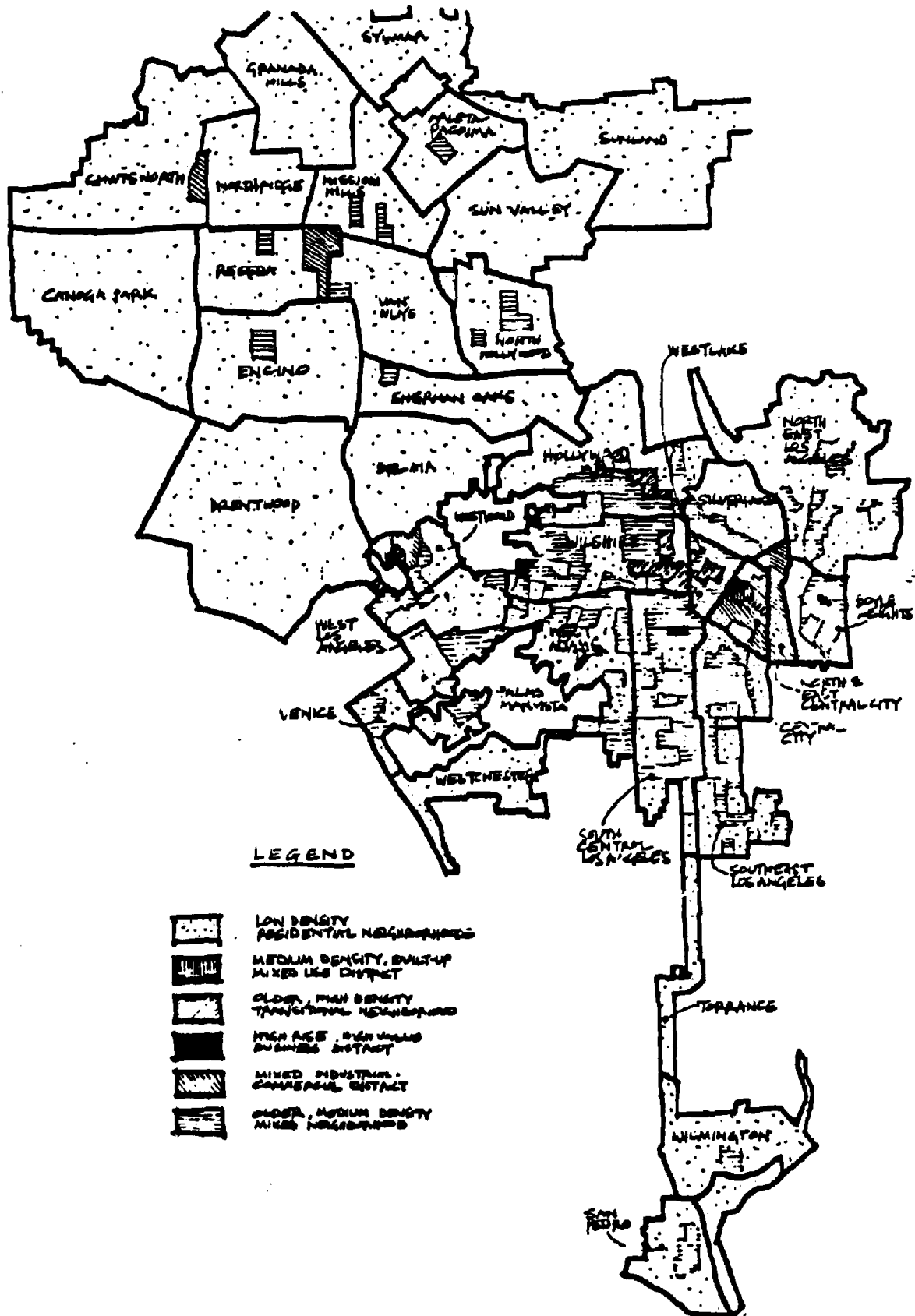


FIGURE II.2 (contd.) KEY TO MAP OF LOS ANGELES' PLANNING DISTRICTS
(order corresponds to matrices in Appendix E)

1. Arleta
2. Brentwood
3. Bel Air
4. Boyle Heights
5. Canoga Park
6. Central City
7. Chatsworth
8. Encino
9. Granada Hills
10. Hollywood
11. Mission Hills
12. N.E. Central City
13. North Hollywood
14. North-East L.A.
15. Northridge
16. Palms
17. Reseda
18. San Pedro
19. Sherman Oaks
20. Silver Lake
21. South-Central L.A.
22. South-East L.A.
23. Sunland
24. Sun Valley
25. Sylmar
26. Torrance Corridor
27. Van Nuys
28. Venice
29. West L.A.
30. West Adams
31. Westchester
32. Westlake
33. Westwood
34. Wilshire
35. Wilmington

all buildings are brick or concrete block masonry structures. More than half of all buildings, however, are wood, or wood and steel structures. Land use is mixed. About a fourth of the acreage is devoted to residential land use. Stores, offices, shopping centers, hotels, motels comprise about two-fifths of the acreage. Restaurants, banks, wholesale and manufacturing outlets, service stations, service shops, parking lots and garages, automobile and equipment sales and service areas, constitute more than a fourth of all land use.

The third cluster is the third largest cluster with a membership of 31 census tracts. These settings can be best described as, Older, High-Density, Transitional Neighborhood. More than half of all buildings in these areas were built before 1934, and another two-fifths since 1934. Despite the large number of pre-1934 buildings, only about a tenth of all structures are brick or concrete masonry structures. Almost all of the rest are wood, or wood and steel structures. Residential use dominates, occupying over three-fourths of the land use acreage. But the fact that almost a quarter of total land use acreage is devoted to various types of commercial and industrial uses, suggests that the area is in transition. With a gross density of 54 persons per acre, this setting represents a high concentration of resident population.

The fourth cluster with membership of only one census tract represents the very heart of the downtown business district. It can be called High-Rise, High-Value, Business District. About two-fifths of the buildings were built between 1934 and 1973. Almost a quarter of all buildings are reinforced concrete fire-proof structures, representing

the highest concentration of this type of building among all settings; most of the rest are wood or wood and steel-framed structures. Residential use represents only two-fifths of all land use acreage. Almost half of all land use is characterized by stores, offices, shopping centers, hotels, motels, restaurants, banks, wholesaling, parking, and the like. Both the value of real estate and the floor area ratio are the highest among all six settings.

With a membership of seventeen census tracts, the fifth cluster can be seen as Mixed Industrial-Commercial District. Three-fifths of the structures were built between 1934 and 1973. About a third were built before 1934. Almost half of all buildings are brick or concrete block masonry structures. Nearly two-fifths of the buildings, however, are wood or wood and steel structures. Industrial and warehousing uses dominate, accounting for about two-fifths of land use acreage. Commercial uses cover about a quarter of the area, while the residential uses occupy another quarter.

The sixth setting is the second largest cluster, with a membership of 210 census tracts. These areas can be best described as Older, Medium-Density, Mixed Neighborhoods. They have the largest concentration of older building stock with about three-fifths built before 1934. About a third of all their buildings were built between 1934 and 1973. Almost all of the structures are of wood or wood and steel construction. Residential uses dominate, with about nine-tenths of the land use acreage. The remainder consists of business and commercial establishments. With about 26 persons per acre gross density they are medium-density neighborhoods.

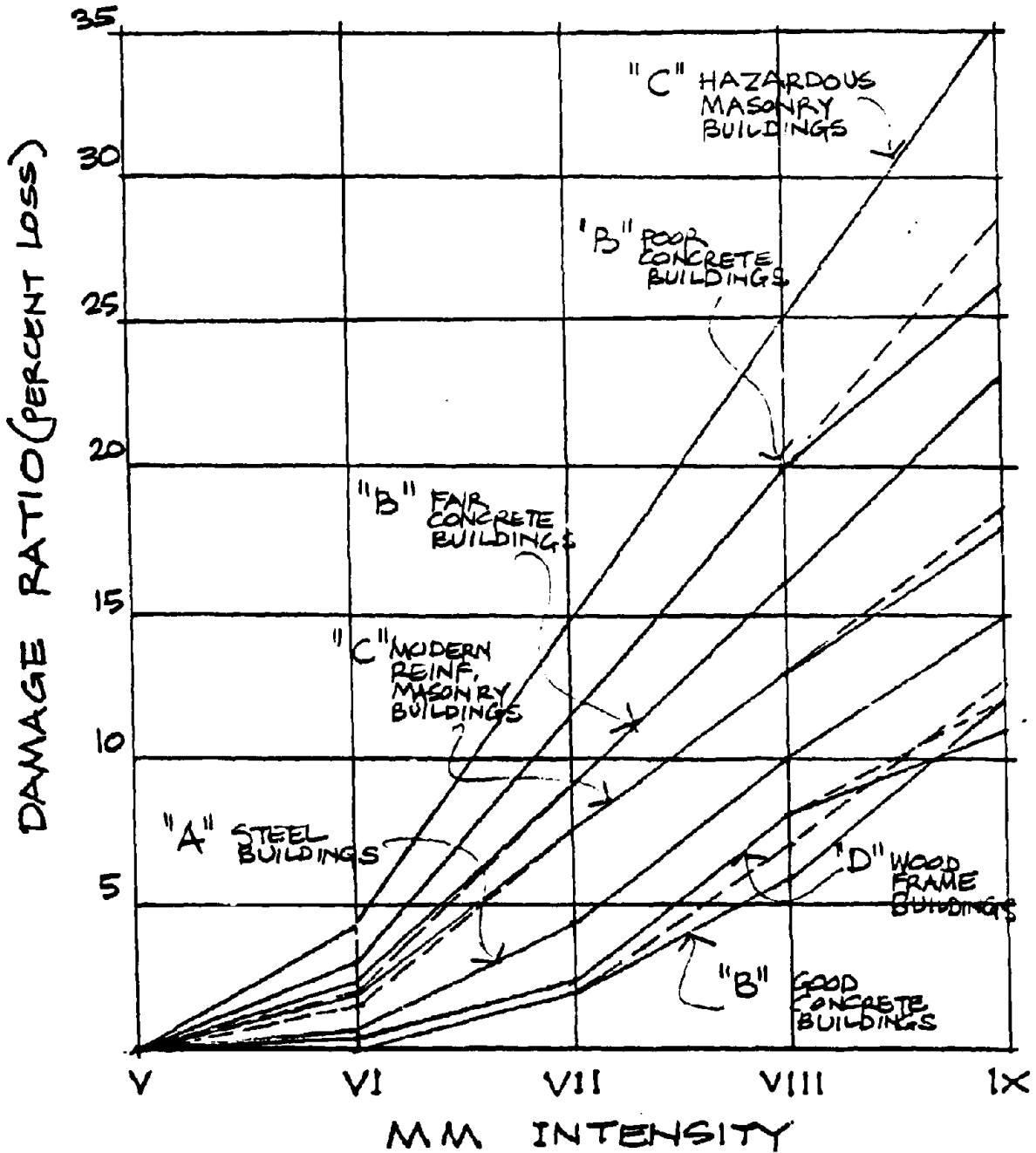
The summary characteristics of the six different setting types are included in materials in Appendix A3. These include cluster means for variables used naming the clusters and representative aerial photographs of each of the six settings.

DAMAGE-CURVE ESTIMATES

To describe the derivation of the damage estimate curves for different setting types requires a brief review of the purposes and results of our three-cycle Delphi exercise.

We began the Delphi exercises by presenting the panelists a set of damage curves by building types developed by William Spangle and Associates and H.J. Degenkolb Associates for the Los Angeles area -- the Degenkolb-Spangle curves in what follows. They are apparently a modified version of the well-known Algermissen and Steinbrugge (1978) damage curves, whose classifications could not be applied to the taxonomy of building types in the LUPAMS data file. We asked panelists to validate these curves in light of the information we had about policies-in-effect (Chapter I) in the City of Los Angeles (page 2, Questionnaire A, Appendix A1.) Figure II.3 summarizes the modified Degenkolb-Spangle curves based on the collective judgments of panelists who were asked whether these curves were accurate given the effects of current ('baseline') policies-in-effect in Los Angeles. The resulting curves should be only slightly different from the original D-S curves shown in Figure 1 on page 5 of the Questionnaire A, Appendix A2.

FIGURE 11.3: MODIFIED D-S CURVES -- RESULTS FROM DELPHI CYCLE ONE*



*Dotted line segments are from the original D-S curves.

From these revised building damage curves and the known (from the cluster analysis) distribution of building types by the six urban settings (Table II.5) we were able to develop 'composite' building damage curves for each of the six settings (Table II.6).

In the second cycle of the Delphi exercise panelists were presented these composite curves and were asked to modify them to reflect additional damages resulting from the interaction effects between buildings in close proximity and possible fires resulting from primary ground shaking. Results of panel judgments are shown in Figure II.4.

In this second cycle panelists were asked also to estimate the damage to the non-building stock (infrastructure, etc.) for each of the six urban settings. These damage curves are shown in Figure II.5.

TABLE II.5: DISTRIBUTION OF BUILDING TYPES BY URBAN SETTINGS

	SETTING					
	1	2	3	4	5	6
BUILDING-TYPE:						
good concrete	0	0.28	0	7.24	0.15	0.17
wood frame	99.27	24.53	94.06	65.52	32.84	84.92
steel buildings	0	15.09	0	3.45	0	0
modern masonry	0.40	29.97	2.43	3.65	43.41	5.11
fair concrete	0	2.83	0	9.66	0.89	0.39
poor concrete	0	2.55	0	7.24	0.45	0.67
hazardous masonry	0.12	22.86	2.57	1.52	17.78	7.18
type-S	0.21	0	0.94	1.72	4.48	1.68

(columns should add up to 100%)

TABLE II.6: COMPOSITE BUILDING DAMAGE CURVES BY SIX URBAN SETTINGS

	SETTING					
	1	2	3	4	5	6
MM INTENSITY:						
VI	0.5	2.0	0.6	1.0	1.8	0.9
VII	2.5	7.5	3.0	4.0	6.7	3.6
VIII	8.0	13.8	8.5	9.8	12.7	9.4
IX	11.0	19.3	11.6	13.8	17.7	13.0

FIGURE II.4: BUILDING DAMAGE INCLUDING INTERACTION EFFECTS -- RESULTS FROM DELPHI CYCLE TWO

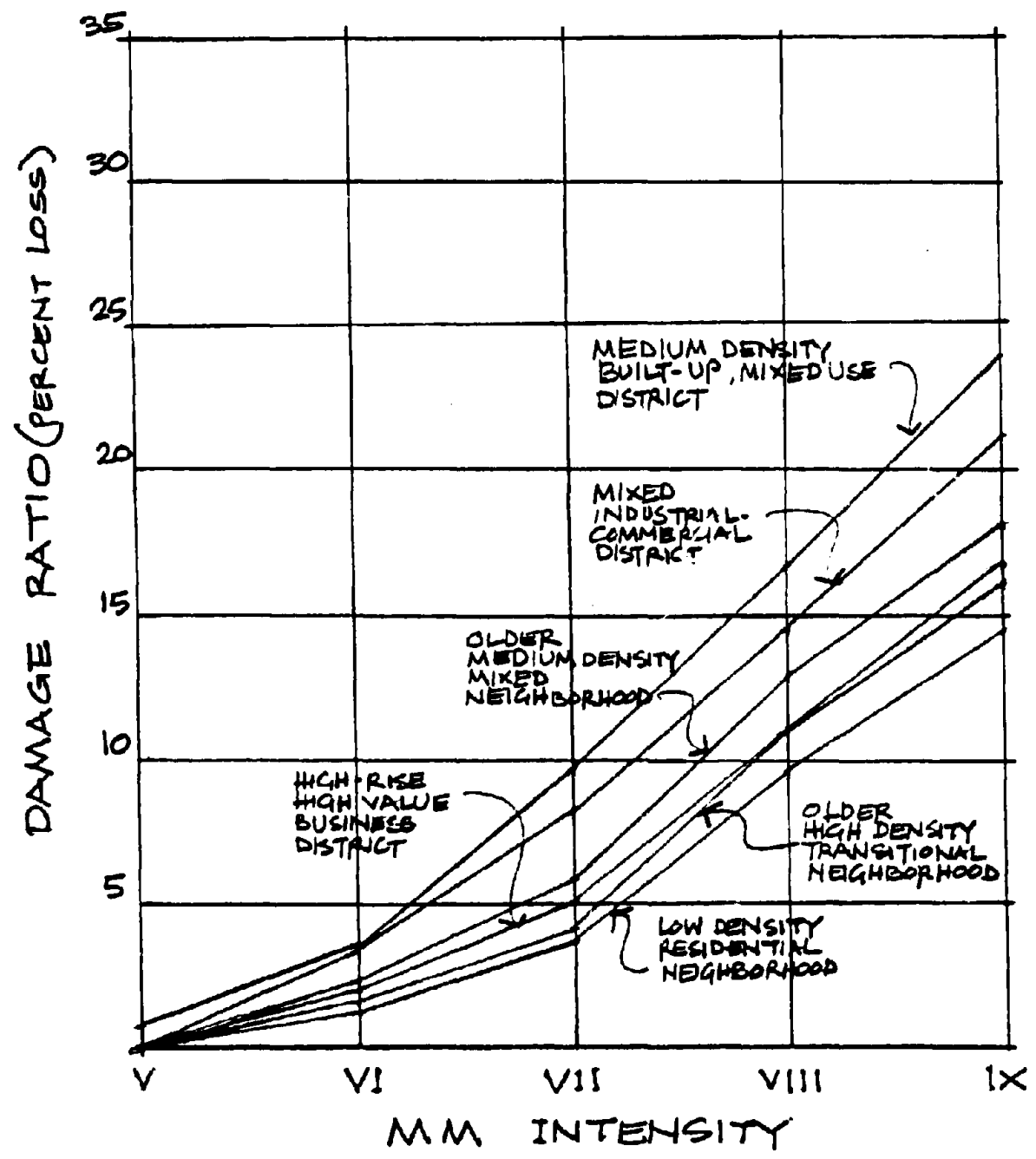
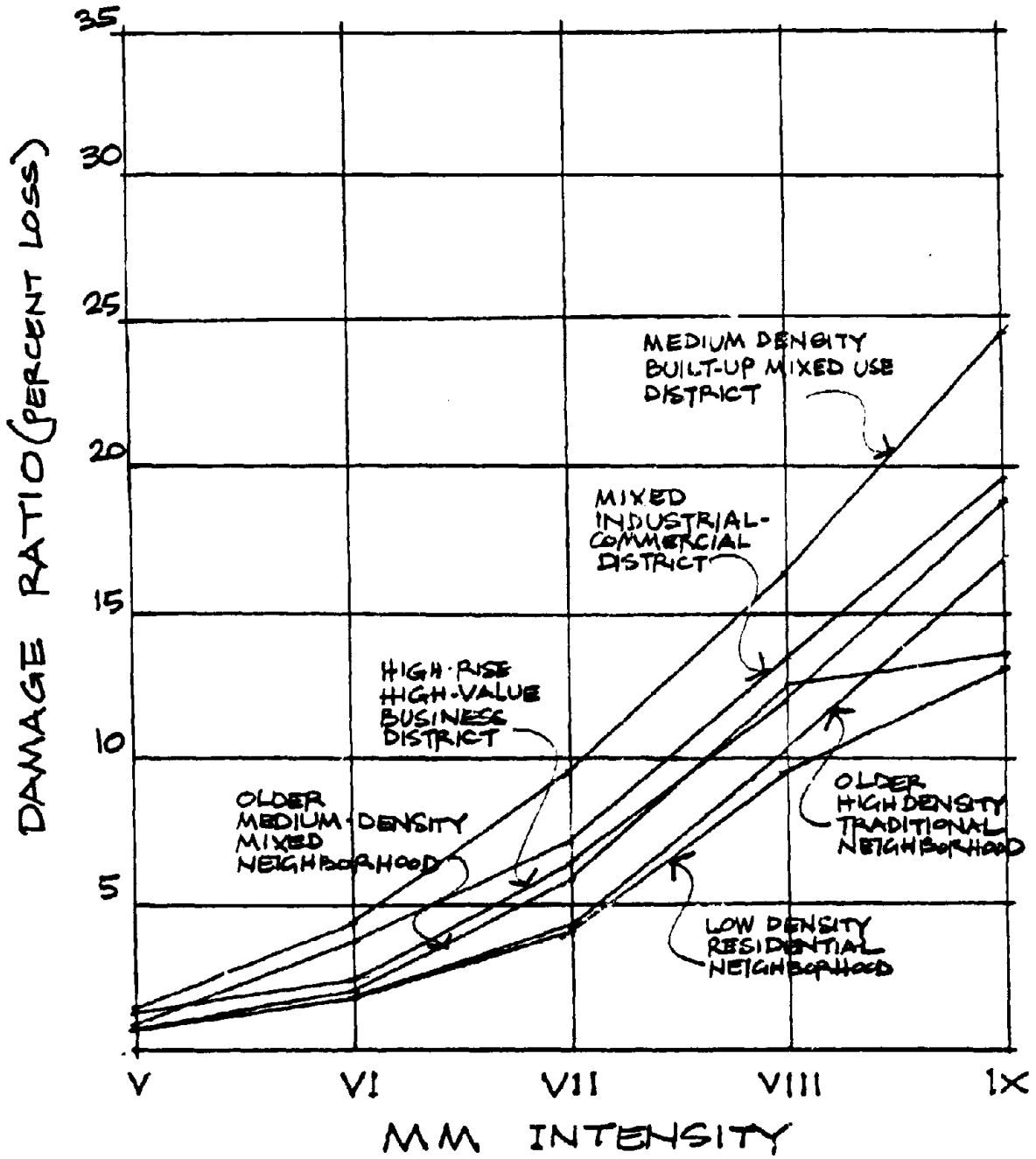


FIGURE II.5: DAMAGE TO NON-BUILDING STOCK -- RESULTS FROM DELPHI CYCLE TWO



Finally in the third cycle of the Delphi exercise, panelists were asked to consider the mitigating effects of the four suggested damage mitigation policies derived in our policy exercise (Chapter I). They were asked to adjust the building damage and the non-building stock damage curves accordingly. Results are summarized in Figures B.1 through B.12 of Appendix B. These graphs then serve as the baseline damage data and damage mitigation-by-policy data inputs to the policy model.

For the sake of the model, the damage estimates and policy-contingent mitigation estimates had also to be introduced via a matrix of thirty-five planning districts, each of which included various shares of the six urban settings. In Figure II.2, the previous map, we show Los Angeles as a mosaic of different urban settings superimposed on the thirty-five planning districts. These damage curves contain at least as much uncertainty as the more familiar curves which are specific to building types and which are silent on any policy context.

As superior methodologies are developed to generate more reliable curves, these can and should be entered into our evaluation model, replacing the ones developed here. An interim procedure might be to make levels of uncertainty explicit. This could be done if questionnaires similar to those developed in this research were elaborated so that panelists could posit levels of likelihood that might be associated with various curve shifts.*

* Similar approaches have recently been implemented by Rojahn, et al., 1984.

TAX-INCIDENCE INFORMATION

The most up-to-date and exhaustive review of state and local tax incidence is by Phares (1980), who sought to determine who pays state and local taxes; he develops incidence information for as many as ninety-three types of state and local taxes for all of the fifty states. Needless to say, Phares' work requires him to make a number of economic assumptions, all of which we accept by utilizing his results.

Of the many taxes that we could have selected, we restricted ourselves to: 1) the state individual income tax; 2) the state general sales and gross receipts tax; 3) total real property taxes; and 4) the local general sales and gross receipts taxes. Together, according to Phares, these accounted for just over 68% of the total state and local tax liability of Californians. We simplified tax incidence calculations by dividing the amount that Californians paid of each of the four taxes by the sum of the four. This 'normalization' ruled out financing any of the earthquake mitigation policies via any but these four tax sources. We calculated that tax #1 accounted for approximately 17% of the total (four-tax) liability of all Californians; tax # 2 was 26%; taxes #3 and #4 accounted for 51% and 6%, respectively.

All of Phares' incidence results were for fourteen family-income groups. Population data for the thirty-five districts of our study area were for seven household-income groups. Luckily, we were able to collapse Phares' fourteen groups to our seven; there were no uncommon group cut-off points. Yet, we were forced to assume that the result-

ing seven groups had the same distribution by families as by households.

The tax-incidence information which was finally used is shown in matrix TAX of Appendix E.

PROBABILITIES

The policy model relies on annual probability-of-risk data for each of the thirty-five analysis zones, with the likelihood of five levels of (MMI) ground-shaking itemized.

The calculations were developed by M.D. Trifunac and his associates. Their methods are documented in Anderson and Trifunac (1978a, 1978b). The seismicity model which Trifunac's group used is made up of 29 faults that span the Southern California region. Their program ('EQRISK') calculates seismic risk at the coordinates of the centroids of the 35 planning areas studied. The probabilities produced are shown in matrix PROB of Appendix F.

VALUE OF CAPITAL AT-RISK

Data on the value of privately owned improvements are from the LUPAMS file described in Chapter II. As suggested there, the large LUPAMS file was reduced to 731 observations, aggregating to create census tract-average values. The sum of dollar values for all tract averages for each of the six clusters was multiplied by the number of parcels believed to be in each cluster. That dollar value was annualized by multiplying by 10%.

Valuations arrived at by this procedure for the each of the six clusters are shown in matrix V of Appendix F. According to LUPAMS (and our processing), the sum of (unexempted) improvement values for the city (counting only the 'first' improvement on each parcel) was approximately \$42 billion.

One of the innovations of our approach is the inclusion of infrastructure at-risk. Dollar valuations of these structures are not included in the LUPAMS file. Our approach was to use data from Musgrave (1980). These are data were assembled by the Bureau of Economic Analysis of the U.S. Department of Commerce. The various BEA reports offer time-series information for the value of government-owned fixed capital for the nation. We selected the series: net value of total (federal as well as state and local) structures in constant (1972) dollars. Though including more than simply the value of public infrastructure, this series picks up other government-owned buildings.

Since these were not included in the LUPAMS file, this is an advantage.

Musgrave's series went to 1979. We extrapolated the trend to 1982 and converted to current dollars. The results were converted to a per capita basis. Per capita values and 1980 population data were then used to assign these wealth data to the thirty-five planning districts.

The resulting values are shown in matrix W of Appendix E. The city's total value of government-owned structures is shown to amount to approximately \$12 billion.

Flowcharts showing model computations (see Appendix C) reveal that damage mitigations to publicly owned and privately owned capital were treated differently since the former were available for the six settings while the latter were available for the thirty-five planning districts. More troubling is the fact that the two sources were quite distinct and conceivably not consistent. Problems with value-of-capital data make these the weakest link in the chain of information inputs that were assembled.

COSTS OF IMPLEMENTATION

A full accounting of the costs of carrying out local earthquake damage mitigation policies would distinguish among three classes of bearers:

1. direct costs borne by private firms and individuals in the local community;
2. outlays of state and local governments passed on to taxpayers through state and local fiscal structures;
3. subventions by the federal government in the form of emergency relief grants and loans to private parties and to local governments.

The federal responsibility is limited to post-event emergency assistance to expedite recovery and consequently is not relevant to our concerns, but the property-oriented mitigation policies we test in this model generate costs which:

- a. are unambiguous charges against the general funds of the city or the state, such as implementation costs;
- b. can only be borne by parties in the private sector, such as costs of compliance with building and zoning standards; or
- c. can be borne either by individuals or be picked up by the local government, as in the case where poor tenants would be evicted from a nonconforming structure to permit it to be demolished who might thereafter be relocated or rehoused at public expense.

Under ideal circumstances we would carry out a careful analysis of each policy, tracing out its consequences and evaluating its impacts on both taxpayers and citizens at-large, account for costs falling under (a) and (b), and allow for a more detailed expression of policies to allow us to attribute each cost under (c) above either to (a)

or (b). The terms of our proposal did not require such intensive policy study for the purpose of simply testing our aggregate model, nor would we have been justified in allocating any substantial share of the resources for the project to that end.

Instead, we turned to our Delphi panel and asked it to use its expertise to make some general estimates for us on these matters, assuming that each policy would be adopted in the near future and would be in effect over the succeeding twenty years. In each case, we provided the panelists with a general statement of each of nineteen prospective policies and a suggestive accounting structure to assist them in their estimation. Furthermore, we asked them to assume that:

1. any capital expenditures will be amortized over a twenty-year period;
2. future nominal interest rates will remain near 10%;
3. the relative contribution of revenue sources to gross revenues will not be influenced by the policy under consideration.

The panelists were provided with a key that allowed them to 'default' if they doubted their ability to render confident judgment, or if they were convinced that the answers depended on unspecified future developments. Otherwise, they were to choose from the answer key the value interval within which their estimates of aggregate costs would fall. The intervals were specified logarithmically rather than arithmetically as a result of a judgment that 'order of magnitude' estimates were sufficiently precise for our purposes, especially where there was a comparatively high degree of consensus.

Table II.7 summarizes how the sixteen panelists responded to our inquiry, policy-by-policy for public and private costs. Tables II.8 and II.9 summarize the actual estimates and their dispersion (ranked by decreasing consensus as measured by the standard deviation of the estimates for each). Table II.10 repeats public cost estimates for each of the four policies we tested.

TABLE II.7: COST RESPONSE SUMMARIES

Policy No.	PRIVATE COST					PUBLIC COS. (ordered by no. of estimates/32)				
	Don't Know	Depends <\$100M		>\$100M	Some est.	Don't Know	Depends <\$100M		>\$100M	Some Est.
	0	1,2	3,4	5-10	3-10	0	1,2	3,4	5-10	3-10
1	7	0	8	1	9	5	2	1	8	9
2	4	1	9	2	11	3	1	5	7	12
3	4	1	1	10	11	4	1	3	8	11
4	6	1	9	0	9	3	0	9	4	13
5	7	1	6	2	8	4	5	1	6	7
6	5	1	9	1	10	3	2	8	3	11
7	6	1	8	1	9	4	1	1	10	11
8	3	1	1	11	12	6	1	5	4	9
9	6	1	0	9	9	3	1	5	7	12
10	6	1	7	2	9	5	0	1	10	11
11	5	0	8	3	11	3	1	1	11	12
12	5	1	9	1	10	4	0	2	10	12
13	5	2	2	7	9	4	3	2	7	9
14	6	2	2	6	8	6	2	0	8	8
15	5	1	10	0	10	3	0	5	8	13
16	5	0	0	11	11	3	0	5	8	13
17	5	1	5	5	10	3	0	1	12	13
18	4	1	9	2	11	3	0	5	7	12
19	4	2	7	3	10	2	1	0	13	13
Sum	98	15	110	77	187	71	21	60	151	211
%	32.24	6.25	36.18	25.33	61.51	23.36	6.91	19.74	49.67	69.41

TABLE II.10: PUBLIC COSTS OF FOUR POLICIES TESTED

POLICY DESCRIPTION	MEAN PANEL ESTIMATE OF PUBLIC COST
Storing Hazardous Substances	\$7,700.
Demolish Nonconforming Buildings	23,713,700.
System Standby Capacity	38,986,000.
20% Tax Credit	16,681,000.

After a careful review of the cost results we decided to omit the private cost calculations from the model exercise on the basis of two considerations:

1. we deemed the basis we had provided the panelists inadequate for us to have confidence that all panelists were approaching each policy with the same 'model' of private costs;
2. estimating the incidence of private costs was not an exercise parallel to or consistent with our treatment of public costs by way of tax incidence, and we required further study if it were not to weaken our confidence in the policy outputs.

This decision in no way reflects any sense that such costs are unimportant in our analysis; it simply respects the complexity of the problem involved. In its final form, our model should treat fully the incidence of both costs and benefits. In its present form, our model operates with the tax incidence, not the cost incidence of the tested earthquake damage mitigation policies.

Finally, the reader should keep in mind that a city might adopt several of such separate policy elements in its efforts to reduce earthquake damage. If so, estimation of costs and benefits would not be simply additive. Both in terms of implementation and physical effects, many of these policy elements are complementary. To explore such combinations of policy elements would require us to treat each feasible combination as a separate policy. This might present some cognitive difficulty for the panelists.

An alternative to addressing this issue is to include an additional questionnaire for 'cross-impact analysis', a common follow-up task for Delphi-based studies. Panelists would be asked to indicate:

1. the probabilities that policy *i* will have an effect on policy *j*; and
2. the percentage by which effects of policy *j* would be enhanced or offset by policy *i*, if they were to interact.

The results are two contingency matrices, one showing probabilities and the other showing percentage changes in the effects. The information available from the two matrices can be used in various ways to reflect the effects of one policy or the cumulative effects of *n* policies on any other. Results could modify the mitigation effects of policies first first made by panelists.

Since one of the fundamental purposes of this exercise was to obtain the judgment of experts on cost figures which would be appropriate to the project's general cost-benefit framework, we now look at the actual aggregated estimates submitted by the panel. We use the rating key means to represent the aggregated judgment of the panel. We are comfortable with this definition where all members of the panel agree; our discomfort grows with dissensus (measured by the standard deviation of panelist responses). If two appraisers value a painting, one saying it is worthless, the other maintaining its worth at \$100,000, it is hard to conclude the painting is worth \$50,000 since both would probably object. We would require more information to determine which valuation is authoritative. In Tables II.7 and II.8 we identify three points: a mean value, which can be interpreted loosely as the least value that half of the respondents would accept, the mean minus one standard deviation -- the least value that roughly 5/6 of the respondents would accept and the mean minus one standard deviation --

the least value that roughly 1/6 of the respondents would accept. If the standard deviation of the answers to a particular question is relatively low (<1, say) we can take the mean to represent the group's estimate.

In other cases, we may be able to discern differences in assumption or perspective among panelists which account for the differences in their answers. We then find a position on one side or the other of the mean in accordance with our own points of view. Without such information, we are justified in adopting a conservative posture and finding a tentative figure on the high side of the mean.

In applying the valuation procedure we need to distinguish 'trivial' from 'significant' costs. If, as Table II.9 suggests, the public costs of achieving joint powers agreements among local jurisdictions reach \$36,300 at most, they would warrant little attention from the City Council and have virtually no effect on any benefit-cost calculations. On the other hand, the decision to construct an emergency shelter system is bound to be significant in both fiscal and welfare terms. In Table II.9 we have extracted the details from Table II.8 of the nineteen policies whose means exceed \$1,000,000, or whose means plus one standard deviation exceed \$10,000,000. Such policies we construe as having fiscally significant costs attached to them. The public costs of all other policies can be taken to be near zero.

One of the seven significant policies has already been identified as a 'benchmark' -- #3, upgrading structurally nonconforming buildings. Two others -- #11 emergency training and #19 the emergency shelter program -- are aimed at preserving life and limb.

Given our present concern about mitigating damage to physical structures, three of our significant policies are addressed to such objectives while the last is concerned with maintaining the local infrastructure so that it would be operational in time of disaster (those designated 'ps' and 'pi' in the 'code' column of Table II.9). These have very substantial costs associated with them at the level of the mean. They also enjoy very substantial standard deviations.

The next chapter which describes the model's workings shows where and how these cost data are utilized.

CHAPTER III: RESULTS

LAYOUT OF THE MODEL

A flow-chart which summarizes computations is shown in Appendix C.

Without going into all of the detail, here are the important steps:

1. A 5x5x6 damage matrix was constructed (for private structures) using results from the damage-curve questionnaires; the dimensions were the six neighborhood types, five MMI levels and five policy states -- the four policy changes discussed in Chapter I as well as a 'baseline' of current mitigation policies-in-place in Los Angeles;
2. A 6x6 diagonal matrix of dollar values denoting the value of private structures at-risk in each of the neighborhood types was used to convert all damage data to dollar loss figures;
3. The results were converted to a thirty-five area format, using information on the distribution of the six generic neighborhoods through the city's thirty-five planning areas;
4. The (5x35) probability matrix converted all of the previous results to expected (annual) values;
5. All of the data were converted to expected annual dollar benefits by subtracting from the baseline-policy matrix; the five dimensions of policy (#1, above) were reduced to four dimensions by computing the difference that the four policy innovations would make;
6. All of these (expected annual dollar benefit) data for the thirty-five planning areas were converted to per capita information, using 1980 population data;
7. The known distribution of each planning area's population among seven income-groups was used to collapse the benefits information to a 4x7 matrix -- of expected annual dollar benefits per capita;
8. A 4x7 tax-incidence matrix, summarizing our use of Phares' data was combined with the four (annualized) policy costs;

9. Another four-dimensional (diagonal) matrix (labeled TP) included the proportion of Californians' (four-tax) liability due to each of the four individual taxes; these data (along with the results from #8) allowed us to compute a 4x7 matrix of annual dollar cost per capita -- for each of the four policies and seven income groups;
10. Annual costs per capita were subtracted from annual expected dollar benefits per capita (#7) to arrive at a (4x7) matrix of net benefits -- or, annual expected net dollar benefits per person -- for each of four policies, for each of the seven income groups;
11. A similar (though not identical) procedure generated net benefits data from reduced damage to infrastructure -- were the same four mitigation policies in place;
12. Results in #10 and #11 were combined to form a 'bottom-line' matrix of net annual expected dollar benefits per person;
13. Bottom-line information was distributed to the thirty-five planning areas -- using information on each income-groups' presence in each of the areas;
14. All of these steps were reproduced with 'conservative' benefits estimates; whereas panelists' mean judgments of damage curve shifts from each policy were used in the three-dimensional matrix (in #1, above), a completely separate calculation was carried out with one-standard-deviation-less-than-the-mean damage curve shifts (actually, the latter curves have a slightly higher displacement).

Great care must be taken in interpreting the models' outputs. It should be noted that none of the data on 'what is on the ground', e.g. the LUPANS and the census data were updated or forecast into the future. That task awaits further work; our model has no real time dimension. The most appropriate interpretation of model results would, therefore, be along the following lines:

The model provides estimates of net annual expected dollar benefits per person given the known distributions of structures and people, presuming that each of the prospective policies is fully in place and that its annualized (public) capital cost is borne by taxpayers according to funding assumptions we make explicit and given the associated tax-incidence, as described in Chapter II.

MODEL OUTPUTS

The policy model was structured in such a way that four sets of outputs are emphasized for each 'run' (each sensitivity test). These are: 1) LASTAV1 which is a 4x7 matrix showing net annual per capita expected dollar benefits of the four policies as they accrue to the seven income groups -- using panelists' average damage curve shift assessments (Table III.1); 2) LASTAV2 which is a 4x35 rearrangement of the same data -- the same four policies by thirty-five planning areas; 3) LASTCON1, also 4x7 but using 'conservative' benefits assessments (Table III.2); 4) LASTCON2, 4x35 also using 'conservative' assessments.

The most striking aspect of results in Table III.1 is that all of the elements are positive; all of the policies impact all of the income groups positively. Using standard benefit-cost criteria, the results say: 'do something', even 'do anything'. Using political criteria, they say almost the same thing since there appear to be no net loser groups.

Looking more closely, the results seem to make a great deal of sense. The poorest do best under all four policy regimes. The model 'knows' where the most vulnerable structures are; it also 'knows' the areas of the City where the poorest people live; it surmises that the most damage-prone buildings are likely to house the poorest people. The model also 'knows' that all of the four taxes are progressive at the lower end.

TABLE III.1: ANNUAL EXPECTED DOLLAR NET BENEFITS PER CAPITA
FOR ANY OF FOUR POLICIES -- ACCRUING TO SEVEN INCOME GROUPS

(average benefits assessment of panelists)

POLICY:	annual household income-group:						
	(i)	(ii)	(iii)	(iv)	(v)	(vi)	(vii)
20% tax credit	7.14	6.03	4.47	4.40	4.34	4.09	4.58
hazard. mats. storage	8.82	7.91	6.57	6.68	6.72	6.58	7.01
demolish nonconf.struc.	11.02	9.68	7.80	7.57	7.46	7.11	7.86
standby infrastruct.	7.74	6.52	4.89	4.52	4.41	4.06	4.95

income groups:

(i)	less than \$5,000
(ii)	\$5,000 - \$9,999
(iii)	\$10,000 - \$14,999
(iv)	\$15,000 - \$19,999
(v)	\$20,000 - \$24,999
(vi)	\$25,000 - \$34,999
(vii)	greater than \$35,000

All of the income groups do best under policy #3, 'demolish non-conforming structures'. This also makes sense since it is well known (Sarin, 1982) that these structures pose an unusual threat in Los Angeles. The model does not 'know' who occupies the most hazardous structures; it simply contains data which denotes that those areas of the city with the oldest buildings are usually the areas populated by lower income groups. Hence, the poorest benefit from policy #3 the most.

Matrix LASTAV2 shows the distribution of these benefits to the thirty-five planning areas. A discussion of the plausibility of these results is beyond the scope of this report, requiring close acquaintance with the nature of each area. Nevertheless, we expect that these results are of interest to political leaders.

As suggested earlier, a more elaborate version of this model would process probabilistic information on any damage curve shifts in response to the various policy initiatives. In the interim, we simply tested more conservative panel responses on damage curve shifts. These utilize smaller-than-average curve shifts; we tested shifts one standard deviation less than the mean.

The 4x7 matrix of 'conservative' results (Table III.2) is interesting since it does suggest political trade-offs. There are negative as well as positive elements suggesting gainers as well as losers. Again, the model presents us with plausible results: where there are gainers and losers, the greatest gains are to the lowest income groups; the greatest losses are to middle-income individuals, reflecting the distribution of tax burdens as well as the occupancy of the

most perilous structures.

LASTCON2 shows which planning areas gain and which lose under the regime of 'conservative' benefits. Again, there are areas with expected gains as well as areas with expected losses.

TABLE III.2: ANNUAL EXPECTED DOLLAR NET BENEFITS PER CAPITA
FOR ANY OF FOUR POLICIES -- ACCRUING TO SEVEN INCOME GROUPS

('conservative' benefits assessment of panelists)

POLICY:	annual household income-group:						
	(i)	(ii)	(iii)	(iv)	(v)	(vi)	(vii)
20% tax credit	0.35	0.19	0.01	-0.15	-0.15	-0.21	0.10
hazard. mats. stor.	0.16	0.15	0.12	0.13	0.14	0.15	0.17
demolish nonconf.struc.	1.40	1.07	0.67	0.37	0.28	0.13	0.49
standby infrastr. cap.	0.67	0.24	-0.27	-0.66	-0.74	-0.91	-0.30

income groups:

(i)	less than \$5,000
(ii)	\$5,000 - \$9,999
(iii)	\$10,000 - \$14,999
(iv)	\$15,000 - \$19,999
(v)	\$20,000 - \$24,999
(vi)	\$25,000 - \$34,999
(vii)	greater than \$35,000

SENSITIVITY TESTS

A large number of sensitivity tests are, of course, possible. That is the attractiveness of the model to policy analysts. The 'test' that has already been described had to do with alternate assumptions on (panelists' views of) policy potency. In this section, we add another test having to do with alternate funding proposals for the same four policies.

Table III.3 repeats outputs already discussed ('baseline' funding) and contrasts these with outputs from four runs of the model which assume that policies are funded wholly by each of the four taxes (state income, state sales, local sales, property), one at a time. Columns 'A' and 'C' repeat the distinction between the panel's average and conservative benefits judgments.

Results for each of the two sales taxes are practically identical; these taxes have almost the same incidence. Funding for the second policy appears not to matter since this is the least expensive of the four proposals. Also, the 'A' columns are only moderately sensitive to funding assumptions. In these cases, benefits so dominate costs that it appears to matter little how the policies are to be financed.

The 'C' columns do suggest some contrasts. The relative progressivity of the state income tax accounts for the greater variability of net benefits were all of the public costs to be financed by it. Financing via real property taxes would distribute net benefits less progres-

sively than for any of the other taxes.

We expect that policy makers would exploit the power of the model by testing a large number of assumptions on funding levels as well as tax source combinations.

TABLE III.3: SENSITIVITY TESTS --
ALTERNATE POLICY FUNDING ASSUMPTIONS

Policies	Sole Funding by:									
	baseline funding		state inc. tax		state sales tax		local sales tax		real prop. tax	
	A	C	A	C	A	C	A	C	A	C
	(seven rows: net annual expected dollar benefits; from lowest bracket in first row to highest bracket)									
20% tax credit	7.14	.35	7.46	.67	7.19	.40	7.19	.40	7.01	.21
	6.03	.19	6.45	.61	6.00	.16	6.00	.16	5.91	.06
	4.47	.01	4.81	.34	4.39	-.08	4.39	-.07	4.41	-.06
	4.40	-.14	4.38	-.16	4.33	-.21	4.32	-.22	4.44	-.10
	4.34	-.15	3.89	-.61	4.34	-.16	4.34	-.16	4.50	-.00
	4.09	-.21	3.40	-.90	4.12	-.17	4.13	-.17	4.29	-.00
	4.58	.10	4.34	-.14	4.67	.19	4.67	.19	4.60	.12
hazardous materials storage	8.82	.16	8.82	.16	8.82	.16	8.82	.16	8.82	.16
	7.91	.15	7.91	.15	7.91	.15	7.91	.15	7.91	.15
	6.57	.12	6.57	.12	6.57	.12	6.57	.12	6.57	.12
	6.68	.13	6.68	.13	6.68	.13	6.68	.13	6.68	.13
	6.72	.14	6.72	.14	6.72	.14	6.72	.14	6.72	.14
	6.58	.15	6.58	.15	6.58	.15	6.58	.15	6.58	.15
	7.01	.17	7.00	.17	7.01	.17	7.01	.17	7.00	.17
demolish nonconf. structures	11.02	1.40	11.47	1.85	11.09	1.47	11.09	1.47	10.82	1.20
	9.68	1.07	10.28	1.67	9.64	1.03	9.64	1.03	9.51	.90
	7.80	.67	8.28	1.15	7.68	.55	7.69	.55	7.71	.58
	7.57	.37	7.55	.35	7.47	.27	7.47	.27	7.64	.44
	7.46	.28	6.81	-.37	7.45	.27	7.45	.27	7.68	.50
	7.11	.13	6.12	-.85	7.16	.18	7.16	.18	7.40	.42
	7.86	.49	7.53	.16	8.00	.63	7.99	.63	7.89	.52
infrastr. standby capacity	7.74	.67	8.48	1.42	7.85	.79	7.85	.79	7.41	.35
	6.52	.24	7.50	1.22	6.45	.17	6.45	.17	6.23	-.05
	4.89	-.27	5.68	.52	4.69	-.46	4.70	-.45	4.74	-.42
	4.52	-.66	4.49	-.69	4.36	-.83	4.35	-.83	4.63	-.55
	4.41	-.74	3.34-1.81		4.40	-.75	4.39	-.76	4.77	-.38
	4.06	-.91	2.45-2.53		4.15	-.82	4.16	-.82	4.55	-.43
	4.95	-.30	4.40	-.86	5.18	-.08	5.17	-.08	5.00	-.25

note: Columns 'A' refer to mean values of damage as assessed by panelists; columns 'C' refer to conservative values, one standard deviation less assessed mitigation.

CHAPTER IV: CONCLUSIONS

We have tried to take a first step toward 'closing the gap' between formal earthquake research and applied policy analysis. The model we developed helped to illuminate the long chain of informational inputs required to move us to usable evaluational assessments. We can report that it is possible to gather and manipulate available information to provide outputs which are a basis for policy contrasts and rankings.

Exercises such as this have the added usefulness that they help investigators focus on weak links which deserve further attention. In fact, it can be argued that one of our model's main capabilities is to identify research priorities for policy analysis in the earthquake mitigation field.

Our own conclusions on immediate priorities were formed with the help of our panelists who met as an Advisory Group to review an early draft of this report (Appendix F). Of immediate concern are the following:

1. more elaborate damage curve estimating procedures should be developed; experts should be prompted to indicate subjective probabilities associated with synthetic curve shifts; model results should be qualified by the appropriate likelihoods;
2. more comprehensive implementation cost data should be gathered; private as well as public cost incidence should be specified as part of model outputs;
3. a larger conception of loss and damage should be considered; loss of life, injuries, and secondary economic losses must be handled by the model.

A number of other tasks will certainly emerge as these three are addressed. Yet, we are optimistic that useable results will be forthcoming and that planning for earthquake damage mitigation can take a large step forward. We hope that the ideas introduced in this research will form a basis for the next generation of policy models.

REFERENCES

- Algermissen, S.T., K.V. Steinbrugge and H.L. Lagorio (1978) Estimation of Earthquake Losses to Buildings U.S. Geological Survey, Open File Report, 78-441.
- Anagnostopoulos, S.A. and R.V. Whitman 'On Human Loss Prediction in Buildings During Earthquakes' Proceedings of the ASCE 671-677.
- Anderson, J.G. and M.D. Trifunac (1978a) 'Uniform Risk Functionals for Characterization of Strong Earthquake Ground Motion' Bulletin of the Seismological Society of America 205-218.
- Anderson, J.G. and M.D. Trifunac (1978b) 'Application of Seismic Risk Procedures to Problems in Microzonation' Proceedings of the Second International Conference on Microzonation for Safer Construction, Research and Applications 559-569.
- Blair, M.L. and W.E. Spangle (1979) Seismic Safety and Land Use Planning -- Selected Examples Geological Survey Paper 941-B. Washington, D.C.: U.S. Government Printing Office.
- Borcherdt, R.D., ed. (1979) Studies of Seismic Zonation of the San Francisco Bay Region Geological Survey Paper 941-A. Washington, D.C.: U.S. Government Printing Office.
- Brabb, E.E., ed., (1979) Progress on Seismic Zonation in the San Francisco Bay Region Geological Survey Circular 807. Washington, D.C.: U.S. Government Printing Office
- Earth Sciences Associates, Inc. (1982) Seismic Environment and Geologic Effects: Technical Report prepared for William Spangle and Associates
- Ellison, R.W. et al. (1984) 'Measuring the Regional Economic Effects of Earthquakes and Earthquake Predictions' Journal of Regional Science 559-578.
- Erley, D. and W.J. Kockelman (1981) Reducing Landslide Hazards: A Guide for Planners Planning Advisory Service Report 359. Chicago: American Planning Association.
- French, S. P. and M.S. Isaacson (1984) 'Applying Earthquake Risk Analysis Techniques to Land Use Planning' The Journal of the American Planning Association 509-522.
- Hirschberg, J.G., et al. (1978) Natural Hazards: Socio-Economic Impact Assessment Model prepared for the National Science Foundation.

Jaffe, M.S. et al. (1981) Reducing Earthquake Risks: A Planner's Guide Chicago: American Planning Association.

Kates, R.W. et al. (1973) 'Human Impact of the Managua Earthquake' Science 981-989.

Kockelman, W.J. (1980) Examples of the Use of Earth-Science Information by Decisionmakers in the San Francisco Bay Region Geological Survey, Open File Report 80-124.

Kendig, Hal (1976) 'Cluster Analysis to Classify Residential Areas: A Los Angeles Application' Journal of the American Institute of Planners 286-294.

Lagorio, H.J. and G.A. Mader (1981) Preliminary Observations Regarding Architectural and Planning Aspects of the November 20, 1980 Earthquake in Southern Italy: Regions of Campania and Basilicata Berkeley: Earthquake Engineering Research Institute.

Moore, D., T. Okamoto, J. Russo, R. Wilson, C. Rojahn (1985) The FEMA Earthquake Damage and Loss Estimation System Prepared for Proceedings of the 1985 Multiconference of the Society for Computer Simulation, San Diego.

Munroe, Tapan and Cynthia Blair (1975) 'Economic Impact in Seismic Design Analysis: A Preliminary Investigation' Seismic Design Decision Analysis report no. 20.

Musgrave, John C. (1980) 'Government-Owned Fixed Capital in the United States' Survey of Current Business, 33-43.

Phares, Donald (1980) Who Pays State and Local Taxes? Cambridge, Massachusetts: Oelgeschlager, Gunn and Hain, Publishers, Inc.

Panel on the Public Policy Implications of Earthquake Prediction of the Advisory Committee on Emergency Planning, Commission on Sociotechnical Systems, National Research Council (1975) Earthquake Prediction and Public Policy Washington, D.C.: National Academy of Sciences

Preble, John F. (1983) 'Public Sector Use of the Delphi Technique' Technological Forecasting and Social Change, 75-88.

Rojahn, C. and R.L. Sharpe. R.V. Nutt (1984) Earthquake Damage Evaluation Data for California Palo Alto: Applied Technology Council.

Rose, Adam 'Utilities, Lifelines and Economic Activities in the Context of Earthquakes' ASCE Conference on Earthquakes and Lifelines.

Sarin, Rakesh (1982) 'Risk Management Policy for Earthquake Reduction' UCLA-ENG-8244.

Schulze, William D., et al. (1985) 'A Benefit-Cost Analysis of Seismic Building Codes' Unpublished, University of Colorado: Department of Economics.

Spangle, W. and Associates (1980) Land Use Planning after Earthquakes Portola Valley, California.

Stratta, J.L. et al. (1981) Earthquake in Campania-Basilicata, Italy, November 23, 1980 Washington, D.C.: National Academy Press.

U.S. Department of Commerce: National Oceanic and Atmospheric Administration (1973) A Study of Earthquake Losses in the Los Angeles, California Area prepared for the Federal Disaster Assistance Administration of the Department of Housing and Urban Development.

Whitman, R.V., N.S. Remmer, B.Schumacker (1980) Feasibility of Regulatory Guidelines for Earthquake Hazard Reduction in Existing Buildings in the Northeast Cambridge: Department of Civil Engineering, MIT.

A P P E N D I X A 1

Policy Delphi (questionnaire and detailed results available
on request)

A P P E N D I X A 2

Damage curves -- cycle I

QES	NAMES	SHIFT	SAS							COM1	
			CURVE1	CURVE2	CURVE3	CURVE4	CURVE5	CURVE6	CURVE7		
1	ARNOLD	0	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1
2	ASAKURA	1	-0.05	-0.16	-0.15	-0.35	-0.16	-0.10	0.00	0.00	1
3	BACHARAC	1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1
4	BLANCK	1	0.00	-0.20	0.00	0.00	-0.05	0.00	0.00	0.00	1
5	BRAUNS	1	0.00	-0.10	0.00	0.00	0.00	0.00	0.00	0.00	1
6	DONOVAN	1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1
7	JOHNSON	0	0.00	-0.20	0.00	0.00	0.00	0.00	0.00	0.00	1
8	JOHNSTON	1	-0.05	-0.25	-0.05	-0.05	-0.05	-0.25	0.00	0.00	1
9	JONES	0	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1
10	JULIAN	1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1
11	KCKELNA	1	0.00	-0.10	-0.05	0.00	-0.16	0.00	0.00	0.00	1
12	MASRI	00	0.00	0.20	0.00	0.00	0.00	0.00	0.00	0.00	1
13	MATINGLY	00	0.20	0.00	0.00	0.00	-0.05	0.00	0.00	0.00	1
14	MCCOY	1	0.00	-0.05	0.00	0.00	0.00	0.00	0.00	0.00	1
15	MEUNIER	1	0.00	-0.05	-0.05	0.00	-0.05	0.00	0.00	0.00	1
16	ETIAK	1	0.00	-0.05	-0.05	0.00	0.00	0.00	0.00	0.00	1
17	SPICER	1	-0.05	-0.05	-0.05	0.00	-0.05	0.00	0.00	0.00	1
18	KATTHAN	0	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1

20

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	SAS	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
COB	16	00000000	02015664	05000000		00000000	00000000	00000000	00000000	00000000
IB	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
STANDARD	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
COB	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
IB	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
STANDARD	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
COB	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
IB	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000
STANDARD	16	00000000	00000000	00000000		00000000	00000000	00000000	00000000	00000000

SAS									
OJS	ID	BLDGTYPE	POLICY	HRV	HRVI	HRVII	HRVIII	HRIX	CON2
58	BACHRACH	6	A	0	0.015	0.075	0.130	0.190	0
59	BACHRACH	6	B	0	0.000	0.035	0.055	0.150	0
60	BACHRACH	6	C	0	0.030	0.000	0.000	0.110	0
61	BACHRACH	7	A	0	0.000	0.005	0.000	0.095	0
62	BACHRACH	7	B	0	0.000	0.000	0.000	0.050	0
63	BACHRACH	7	C	0	0.005	0.025	0.000	0.115	0
64	BLANCK	1	A	0
65	BLANCK	1	B	0
66	BLANCK	1	C	0
67	BLANCK	2	A	0	0.000	0.065	0.150	0.235	0
68	BLANCK	2	B	0
69	BLANCK	2	C	0
70	BLANCK	3	A	0
71	BLANCK	3	B	0
72	BLANCK	3	C	0
73	BLANCK	4	A	0
74	BLANCK	4	B	0
75	BLANCK	4	C	0
76	BLANCK	5	A	0
77	BLANCK	5	B	0
78	BLANCK	5	C	0	0.000	0.100	0.200	0.305	0
79	BLANCK	6	A	0
80	BLANCK	6	B	0
81	BLANCK	6	C	0	0.000	0.000	0.030	0.070	0
82	BLANCK	7	A	0
83	BLANCK	7	B	0
84	BLANCK	7	C	0
85	BLANCK	8	A	0
86	BLANCK	8	B	0
87	BLANCK	8	C	0
88	BLANCK	9	A	0
89	BLANCK	9	B	0
90	BLANCK	9	C	0
91	BLANCK	10	A	0
92	BLANCK	10	B	0
93	BLANCK	10	C	0
94	BLANCK	11	A	0
95	BLANCK	11	B	0
96	BLANCK	11	C	0
97	BLANCK	12	A	0	0.020	0.070	0.135	0.175	0
98	BLANCK	12	B	0	0.015	0.075	0.130	0.190	0
99	BLANCK	12	C	0	0.015	0.075	0.130	0.190	0
100	BLANCK	13	A	0	0.000	0.010	0.055	0.100	0
101	BLANCK	13	B	0
102	BLANCK	13	C	0
103	BLANCK	14	A	0
104	BLANCK	14	B	0
105	BLANCK	14	C	0
106	DOHOVAN	1	A	0	0.010	0.040	0.100	0.150	0
107	DOHOVAN	1	B	0	0.010	0.040	0.100	0.150	0
108	DOHOVAN	1	C	0	0.030	0.120	0.200	0.250	0
109	DOHOVAN	2	A	0	0.035	0.115	0.200	0.250	0
110	DOHOVAN	2	B	0	0.035	0.115	0.200	0.250	0
111	DOHOVAN	2	C	0	0.020	0.090	0.160	0.200	0
112	DOHOVAN	3	A	0	0.020	0.090	0.160	0.200	0
113	DOHOVAN	3	B	0	0.000	0.040	0.100	0.150	0
114	DOHOVAN	3	C	0	0.000	0.040	0.100	0.150	0

OBS	ID	BLDGTYPE	POLICY	SAS					CON2
				HHV	HHVI	HHVII	HHVIII	HHIX	
115	DO NOVAN	#	A	0	0	0	0.070	0.125	1
116	DO NOVAN	#	A	0	0	0	0.070	0.125	1
117	DO NOVAN	#	A	0	0	0	0.070	0.125	1
118	DO NOVAN	#	A	0	0	0	0.070	0.125	1
119	DO NOVAN	#	A	0	0	0	0.070	0.125	1
120	DO NOVAN	#	A	0	0	0	0.070	0.125	1
121	DO NOVAN	#	A	0	0	0	0.070	0.125	1
122	DO NOVAN	#	A	0	0	0	0.070	0.125	1
123	DO NOVAN	#	A	0	0	0	0.070	0.125	1
124	DO NOVAN	#	A	0	0	0	0.070	0.125	1
125	DO NOVAN	#	A	0	0	0	0.070	0.125	1
126	DO NOVAN	#	A	0	0	0	0.070	0.125	1
127	DO NOVAN	#	A	0	0	0	0.070	0.125	1
128	DO NOVAN	#	A	0	0	0	0.070	0.125	1
129	DO NOVAN	#	A	0	0	0	0.070	0.125	1
130	DO NOVAN	#	A	0	0	0	0.070	0.125	1
131	DO NOVAN	#	A	0	0	0	0.070	0.125	1
132	DO NOVAN	#	A	0	0	0	0.070	0.125	1
133	DO NOVAN	#	A	0	0	0	0.070	0.125	1
134	DO NOVAN	#	A	0	0	0	0.070	0.125	1
135	DO NOVAN	#	A	0	0	0	0.070	0.125	1
136	DO NOVAN	#	A	0	0	0	0.070	0.125	1
137	DO NOVAN	#	A	0	0	0	0.070	0.125	1
138	DO NOVAN	#	A	0	0	0	0.070	0.125	1
139	DO NOVAN	#	A	0	0	0	0.070	0.125	1
140	DO NOVAN	#	A	0	0	0	0.070	0.125	1
141	DO NOVAN	#	A	0	0	0	0.070	0.125	1
142	DO NOVAN	#	A	0	0	0	0.070	0.125	1
143	DO NOVAN	#	A	0	0	0	0.070	0.125	1
144	DO NOVAN	#	A	0	0	0	0.070	0.125	1
145	DO NOVAN	#	A	0	0	0	0.070	0.125	1
146	DO NOVAN	#	A	0	0	0	0.070	0.125	1
147	DO NOVAN	#	A	0	0	0	0.070	0.125	1
148	DO NOVAN	#	A	0	0	0	0.070	0.125	1
149	DO NOVAN	#	A	0	0	0	0.070	0.125	1
150	DO NOVAN	#	A	0	0	0	0.070	0.125	1
151	DO NOVAN	#	A	0	0	0	0.070	0.125	1
152	DO NOVAN	#	A	0	0	0	0.070	0.125	1
153	DO NOVAN	#	A	0	0	0	0.070	0.125	1
154	DO NOVAN	#	A	0	0	0	0.070	0.125	1
155	DO NOVAN	#	A	0	0	0	0.070	0.125	1
156	DO NOVAN	#	A	0	0	0	0.070	0.125	1
157	DO NOVAN	#	A	0	0	0	0.070	0.125	1
158	DO NOVAN	#	A	0	0	0	0.070	0.125	1
159	DO NOVAN	#	A	0	0	0	0.070	0.125	1
160	DO NOVAN	#	A	0	0	0	0.070	0.125	1
161	DO NOVAN	#	A	0	0	0	0.070	0.125	1
162	DO NOVAN	#	A	0	0	0	0.070	0.125	1
163	DO NOVAN	#	A	0	0	0	0.070	0.125	1
164	DO NOVAN	#	A	0	0	0	0.070	0.125	1
165	DO NOVAN	#	A	0	0	0	0.070	0.125	1
166	DO NOVAN	#	A	0	0	0	0.070	0.125	1
167	DO NOVAN	#	A	0	0	0	0.070	0.125	1
168	DO NOVAN	#	A	0	0	0	0.070	0.125	1
169	DO NOVAN	#	A	0	0	0	0.070	0.125	1
170	DO NOVAN	#	A	0	0	0	0.070	0.125	1
171	DO NOVAN	#	A	0	0	0	0.070	0.125	1

OBS	ID	BLDGTYPE	POLICY	SAS				SBIK	COM2
				MMV	MMVI	MMVII	MMVIII		
172	JOCHEM	3	A	0	0.0335	0.100	0.175	0	0.000
173	JOCHEM	3	A	0	0.0335	0.100	0.160	0	0.000
174	JOCHEM	3	A	0	0.0335	0.070	0.110	0	0.000
175	JOCHEM	3	A	0	0.0220	0.080	0.145	0	0.000
176	JOCHEM	3	A	0	0.0220	0.075	0.125	0	0.000
177	JOCHEM	3	A	0	0.0000	0.060	0.100	0	0.000
178	JOCHEM	3	A	0	0.0000	0.020	0.060	0	0.000
179	JOCHEM	3	A	0	0.0000	0.020	0.060	0	0.000
180	JOCHEM	3	A	0	0.0330	0.090	0.160	0	0.000
181	JOCHEM	3	A	0	0.0330	0.115	0.190	0	0.000
182	JOCHEM	3	A	0	0.0220	0.040	0.060	0	0.000
183	JOCHEM	3	A	0	0.0110	0.050	0.110	0	0.000
184	JOCHEM	3	A	0	0.0110	0.060	0.120	0	0.000
185	JOCHEM	3	A	0	0.0000	0.020	0.050	0	0.000
186	JOCHEM	3	A	0	0.0000	0.020	0.050	0	0.000
187	JOCHEM	3	A	0	0.0000	0.020	0.050	0	0.000
188	JOCHEM	3	A	0	0.0000	0.020	0.050	0	0.000
189	JOCHEM	3	A	0	0.0000	0.020	0.050	0	0.000
190	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
191	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
192	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
193	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
194	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
195	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
196	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
197	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
198	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
199	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
200	JOCHEM	3	A	0	0.0000	0.040	0.100	0	0.000
201	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
202	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
203	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
204	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
205	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
206	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
207	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
208	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
209	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
210	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
211	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
212	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
213	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
214	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
215	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
216	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
217	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
218	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
219	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
220	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
221	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
222	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
223	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
224	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
225	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
226	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
227	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000
228	JOCHEM	3	A	0	0.0000	0.020	0.070	0	0.000

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OBS	ID	BLDGTYPE	POLICY	SAS					CON2
				HHV	HHVI	HHVII	HHVIII	HHIX	
343	SPICIR	#####	B	0	0.0020	0.0000	0.0160	0.2250	#####
344	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
345	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
346	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
347	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
348	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
349	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
350	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
351	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
352	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
353	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
354	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
355	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
356	SPICIR	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
357	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
358	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
359	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
360	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
361	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
362	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
363	WHITPAN	#####	B	0	0.0150	0.0050	0.1550	0.2250	#####
364	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
365	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
366	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
367	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
368	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
369	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
370	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
371	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
372	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
373	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
374	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
375	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
376	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####
377	WHITPAN	#####	B	0	0.0000	0.0000	0.0000	0.0000	#####

DATA FROM DELPHI BY BUILDING TYPE AND POLICY

OBS	ID	BLDGTYPE	POLICY	HNV	HVII	HVII	HVIII	HXII	COB2
1	ARNOLD	1	A	0.	0.010	0.040	0.100	0.150	1
2	ASAKUEA	1	A	0.	0.010	0.040	0.100	0.150	1
3	BACHAFAC	1	A	0.	0.	0.	0.	0.	1
4	BLANCK	1	A	0.	0.	0.	0.	0.	1
5	BRAUNS	1	A	0.	0.010	0.040	0.100	0.150	1
6	DOHOVAN	1	A	0.	0.010	0.040	0.100	0.150	1
7	JOHNSCN	1	A	0.	0.010	0.040	0.100	0.150	1
8	JOHNSCN	1	A	0.	0.010	0.040	0.100	0.150	1
9	JONES	1	A	0.	0.010	0.040	0.100	0.150	1
10	JULIAN	1	A	0.	0.000	0.000	0.000	0.075	1
11	KOCKELMA	1	A	0.	0.000	0.000	0.000	0.145	1
12	MASRI	1	A	0.	0.001	0.030	0.090	0.145	1
13	MATINGLY	1	A	0.	0.000	0.030	0.085	0.135	1
14	MCCOY	1	A	0.	0.000	0.030	0.085	0.135	1
15	MEUNIER	1	A	0.	0.010	0.040	0.100	0.150	1
16	PETAK	1	A	0.	0.010	0.040	0.100	0.150	1
17	SPICZF	1	A	0.	0.000	0.035	0.095	0.145	1
18	WHITMAN	1	A	0.	0.000	0.035	0.095	0.145	1
19	ARNOLD	1	B	0.	0.010	0.040	0.100	0.150	1
20	ASAKUEA	1	B	0.	0.000	0.000	0.040	0.095	1
21	BACHAFAC	1	B	0.	0.	0.	0.	0.	1
22	BLANCK	1	B	0.	0.	0.	0.	0.	1
23	BRAUNS	1	B	0.	0.010	0.040	0.100	0.150	1
24	DOHOVAN	1	B	0.	0.010	0.040	0.100	0.150	1
25	JOHNSCN	1	B	0.	0.010	0.040	0.100	0.150	1
26	JOHNSCN	1	B	0.	0.010	0.040	0.100	0.150	1
27	JONES	1	B	0.	0.010	0.040	0.100	0.150	1
28	JULIAN	1	B	0.	0.000	0.000	0.000	0.150	1
29	KOCKELMA	1	B	0.	0.000	0.000	0.000	0.175	1
30	MASRI	1	B	0.	0.000	0.000	0.000	0.150	1
31	MATINGLY	1	B	0.	0.000	0.000	0.000	0.100	1
32	MCCOY	1	B	0.	0.000	0.000	0.000	0.180	1
33	MEUNIER	1	B	0.	0.000	0.000	0.000	0.135	1
34	PETAK	1	B	0.	0.000	0.000	0.000	0.150	1
35	SPICZF	1	B	0.	0.010	0.040	0.100	0.150	1
36	WHITMAN	1	B	0.	0.010	0.040	0.100	0.150	1
37	ARNOLD	1	C	0.	0.010	0.040	0.100	0.150	1
38	ASAKUEA	1	C	0.	0.000	0.000	0.000	0.050	1
39	BACHAFAC	1	C	0.	0.	0.	0.	0.	1
40	BLANCK	1	C	0.	0.	0.	0.	0.	1
41	BRAUNS	1	C	0.	0.010	0.040	0.100	0.150	1
42	DOHOVAN	1	C	0.	0.010	0.040	0.100	0.150	1
43	JOHNSCN	1	C	0.	0.010	0.040	0.100	0.150	1
44	JOHNSCN	1	C	0.	0.010	0.040	0.100	0.150	1
45	JONES	1	C	0.	0.010	0.040	0.100	0.150	1
46	JULIAN	1	C	0.	0.000	0.000	0.000	0.150	1
47	KOCKELMA	1	C	0.	0.000	0.000	0.000	0.000	1
48	MASRI	1	C	0.	0.000	0.000	0.000	0.000	1
49	MATINGLY	1	C	0.	0.000	0.000	0.000	0.000	1
50	MCCOY	1	C	0.	0.000	0.000	0.000	0.000	1
51	MEUNIER	1	C	0.	0.000	0.000	0.000	0.000	1
52	PETAK	1	C	0.	0.010	0.040	0.100	0.150	1
53	SPICZF	1	C	0.	0.010	0.040	0.100	0.150	1
54	WHITMAN	1	C	0.	0.010	0.040	0.100	0.150	1
55	ARNOLD	2	A	0.	0.015	0.065	0.150	0.235	1
56	ASAKUEA	2	A	0.	0.030	0.120	0.200	0.275	1
57	BACHAFAC	2	A	0.	0.	0.	0.	0.	1

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DATA FROM DELPHI BY BUILDING TYPE AND POLICY

OBS	ID	BLDGTYPE	POLICY	HMV	HMVI	HMVII	HMVIII	HMIX	COR2
58	BLANCK	U	A	0.00000000
59	BRAUNS	U	A	0.00000000
60	DONOVAN	U	A	0.00000000
61	JOHNSON	U	A	0.00000000
62	JOHNSON	U	A	0.00000000
63	JONES	U	A	0.00000000
64	JULIAN	U	A	0.00000000
65	KOCKELMA	U	A	0.00000000
66	NASRI	U	A	0.00000000
67	HATINGLY	U	A	0.00000000
68	HCCOY	U	A	0.00000000
69	NEUNIER	U	A	0.00000000
70	PETAK	U	A	0.00000000
71	SPICEE	U	A	0.00000000
72	WHITMAN	U	A	0.00000000
73	ARNOLE	U	B	0.00000000
74	ASAKUEA	U	B	0.00000000
75	BACHAFAC	U	B	0.00000000
76	BLANCK	U	B	0.00000000
77	BRAUNS	U	B	0.00000000
78	DONOVAN	U	B	0.00000000
79	JOHNSON	U	B	0.00000000
80	JOHNSON	U	B	0.00000000
81	JONES	U	B	0.00000000
82	JULIAN	U	B	0.00000000
83	KOCKELMA	U	B	0.00000000
84	NASRI	U	B	0.00000000
85	HATINGLY	U	B	0.00000000
86	HCCOY	U	B	0.00000000
87	NEUNIER	U	B	0.00000000
88	PETAK	U	B	0.00000000
89	SPICEE	U	B	0.00000000
90	WHITMAN	U	B	0.00000000
91	ARNOLE	U	C	0.00000000
92	ASAKUEA	U	C	0.00000000
93	BACHAFAC	U	C	0.00000000
94	BLANCK	U	C	0.00000000
95	BRAUNS	U	C	0.00000000
96	DONOVAN	U	C	0.00000000
97	JOHNSON	U	C	0.00000000
98	JOHNSON	U	C	0.00000000
99	JONES	U	C	0.00000000
100	JULIAN	U	C	0.00000000
101	KOCKELMA	U	C	0.00000000
102	NASRI	U	C	0.00000000
103	HATINGLY	U	C	0.00000000
104	NEUNIER	U	C	0.00000000
105	PETAK	U	C	0.00000000
106	SPICEE	U	C	0.00000000
107	WHITMAN	U	C	0.00000000
108	ARNOLE	U	A	0.00000000
109	ASAKUEA	U	A	0.00000000
110	BACHAFAC	U	A	0.00000000
111	BLANCK	U	A	0.00000000
112	BRAUNS	U	A	0.00000000
113	DONOVAN	U	A	0.00000000
114	JOHNSON	U	A	0.00000000

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DATA FROM DELPHI BY BUILDING TYPE AND POLICY

OBS	ID	BLDGTYPE	POLICY	BHV	BHVI	BHVII	BHVIII	BHIX	CON2
115	JOHNSTON		A	0	0.020	0.065	0.110	0.150	0
116	JONES		A	0	0.020	0.080	0.145	0.210	0
117	JULIAN		A	0	0.020	0.080	0.160	0.230	0
118	KOCKEINA		A	0	0.020	0.080	0.085	0.150	0
119	HASBI		A	0	0.020	0.080	0.135	0.200	0
120	HATINGLY		A	0	0.020	0.080	0.180	0.210	0
121	HCCOY		A	0	0.020	0.080	0.130	0.180	0
122	HEUNIER		A	0	0.020	0.080	0.140	0.205	0
123	PETAK		A	0	0.020	0.080	0.160	0.230	0
124	SPICER		A	0	0.020	0.080	0.155	0.225	0
125	WHITMAN		A	0	0.020	0.080	0.080	0.150	0
126	ARNOLD		B	0	0.000	0.010	0.100	0.170	0
127	ASAKUBA		B	0	0.020	0.080	0.160	0.230	0
128	BACHAFAC		B	0	0.020	0.080	0.080	0.150	0
129	BLANCH		B	0	0.020	0.080	0.080	0.150	0
130	BRAUNS		B	0	0.020	0.080	0.160	0.230	0
131	DONOVAN		B	0	0.020	0.080	0.160	0.230	0
132	JOHNSON		B	0	0.020	0.080	0.160	0.230	0
133	JOHNSTON		B	0	0.020	0.080	0.125	0.175	0
134	JONES		B	0	0.020	0.080	0.125	0.230	0
135	JULIAN		B	0	0.020	0.080	0.100	0.150	0
136	KOCKELNA		B	0	0.020	0.080	0.150	0.120	0
137	HASBI		B	0	0.020	0.080	0.130	0.140	0
138	HATINGLY		B	0	0.020	0.080	0.150	0.220	0
139	HCCOY		B	0	0.020	0.080	0.150	0.210	0
140	HEUNIER		B	0	0.020	0.080	0.160	0.205	0
141	PETAK		B	0	0.020	0.080	0.160	0.230	0
142	SPICER		B	0	0.020	0.080	0.160	0.230	0
143	WHITMAN		B	0	0.020	0.080	0.160	0.230	0
144	ARNOLD		C	0	0.000	0.000	0.060	0.125	0
145	ASAKUBA		C	0	0.020	0.080	0.160	0.230	0
146	BACHAFAC		C	0	0.020	0.080	0.160	0.230	0
147	BLANCH		C	0	0.020	0.080	0.160	0.230	0
148	BRAUNS		C	0	0.000	0.000	0.110	0.190	0
149	DONOVAN		C	0	0.000	0.000	0.120	0.190	0
150	JOHNSON		C	0	0.000	0.000	0.075	0.140	0
151	JOHNSTON		C	0	0.000	0.000	0.160	0.180	0
152	JONES		C	0	0.000	0.000	0.100	0.180	0
153	JULIAN		C	0	0.000	0.000	0.100	0.100	0
154	KOCKEINA		C	0	0.000	0.000	0.050	0.120	0
155	HASBI		C	0	0.000	0.000	0.050	0.180	0
156	HATINGLY		C	0	0.000	0.000	0.160	0.230	0
157	HCCOY		C	0	0.000	0.000	0.050	0.125	0
158	HEUNIER		C	0	0.000	0.000	0.050	0.130	0
159	PETAK		C	0	0.000	0.000	0.050	0.125	0
160	SPICER		C	0	0.000	0.000	0.050	0.125	0
161	WHITMAN		C	0	0.000	0.000	0.050	0.125	0
162	ARNOLD		A	0	0.000	0.000	0.020	0.125	0
163	ASAKUBA		A	0	0.000	0.000	0.070	0.125	0
164	BACHAFAC		A	0	0.000	0.000	0.070	0.125	0
165	BLANCH		A	0	0.000	0.000	0.070	0.125	0
166	BRAUNS		A	0	0.000	0.000	0.070	0.125	0
167	DONOVAN		A	0	0.000	0.000	0.060	0.115	0
168	JOHNSON		A	0	0.000	0.000	0.060	0.100	0
169	JOHNSTON		A	0	0.000	0.000	0.060	0.120	0
170	JONES		A	0	0.000	0.000	0.070	0.120	0
171	JULIAN		A	0	0.000	0.000	0.070	0.125	0

DATA FROM DELPHI BY BUILDING TYPE AND POLICY

OBS	ID	BLDGTYPE	POLICY	HMV	HMVI	HMVII	HMVIII	HMIX	COR2
172	ROCKELMA	4	A	0	0.0000	0.0000	0.0000	0.0000	0
173	HASRI	4	A	0	0.0000	0.0000	0.0000	0.0000	0
174	HATINGLY	4	A	0	0.0000	0.0000	0.0000	0.0000	0
175	HCCOY	4	A	0	0.0000	0.0000	0.0000	0.0000	0
176	HEUNIER	4	A	0	0.0000	0.0000	0.0000	0.0000	0
177	PETAK	4	A	0	0.0000	0.0000	0.0000	0.0000	0
178	SPICER	4	A	0	0.0000	0.0000	0.0000	0.0000	0
179	WHITHAM	4	A	0	0.0000	0.0000	0.0000	0.0000	0
180	ABMOLF	4	B	0	0.0000	0.0000	0.0000	0.0000	0
181	ASAKUFA	4	B	0	0.0000	0.0000	0.0000	0.0000	0
182	BACHAFAC	4	B	0	0.0000	0.0000	0.0000	0.0000	0
183	BLANCH	4	B	0	0.0000	0.0000	0.0000	0.0000	0
184	BRAUNS	4	B	0	0.0000	0.0000	0.0000	0.0000	0
185	DONOVAN	4	B	0	0.0000	0.0000	0.0000	0.0000	0
186	JOHNSON	4	B	0	0.0000	0.0000	0.0000	0.0000	0
187	JOHNSTON	4	B	0	0.0000	0.0000	0.0000	0.0000	0
188	JONES	4	B	0	0.0000	0.0000	0.0000	0.0000	0
189	JULIAN	4	B	0	0.0000	0.0000	0.0000	0.0000	0
190	ROCKELMA	4	B	0	0.0000	0.0000	0.0000	0.0000	0
191	HASRI	4	B	0	0.0000	0.0000	0.0000	0.0000	0
192	HATINGLY	4	B	0	0.0000	0.0000	0.0000	0.0000	0
193	HCCOY	4	B	0	0.0000	0.0000	0.0000	0.0000	0
194	HEUNIER	4	B	0	0.0000	0.0000	0.0000	0.0000	0
195	PETAK	4	B	0	0.0000	0.0000	0.0000	0.0000	0
196	SPICER	4	B	0	0.0000	0.0000	0.0000	0.0000	0
197	WHITHAM	4	B	0	0.0000	0.0000	0.0000	0.0000	0
198	ABMOLF	4	C	0	0.0000	0.0000	0.0000	0.0000	0
199	ASAKUFA	4	C	0	0.0000	0.0000	0.0000	0.0000	0
200	BACHAFAC	4	C	0	0.0000	0.0000	0.0000	0.0000	0
201	BLANCH	4	C	0	0.0000	0.0000	0.0000	0.0000	0
202	BRAUNS	4	C	0	0.0000	0.0000	0.0000	0.0000	0
203	DONOVAN	4	C	0	0.0000	0.0000	0.0000	0.0000	0
204	JOHNSON	4	C	0	0.0000	0.0000	0.0000	0.0000	0
205	JOHNSTON	4	C	0	0.0000	0.0000	0.0000	0.0000	0
206	JONES	4	C	0	0.0000	0.0000	0.0000	0.0000	0
207	JULIAN	4	C	0	0.0000	0.0000	0.0000	0.0000	0
208	ROCKELMA	4	C	0	0.0000	0.0000	0.0000	0.0000	0
209	HASRI	4	C	0	0.0000	0.0000	0.0000	0.0000	0
210	HATINGLY	4	C	0	0.0000	0.0000	0.0000	0.0000	0
211	HCCOY	4	C	0	0.0000	0.0000	0.0000	0.0000	0
212	HEUNIER	4	C	0	0.0000	0.0000	0.0000	0.0000	0
213	PETAK	4	C	0	0.0000	0.0000	0.0000	0.0000	0
214	SPICER	4	C	0	0.0000	0.0000	0.0000	0.0000	0
215	WHITHAM	4	C	0	0.0000	0.0000	0.0000	0.0000	0
216	ABMOLF	4	A	0	0.0000	0.0000	0.0000	0.0000	0
217	ASAKUFA	4	A	0	0.0000	0.0000	0.0000	0.0000	0
218	BACHAFAC	4	A	0	0.0000	0.0000	0.0000	0.0000	0
219	BLANCH	4	A	0	0.0000	0.0000	0.0000	0.0000	0
220	BRAUNS	4	A	0	0.0000	0.0000	0.0000	0.0000	0
221	DONOVAN	4	A	0	0.0000	0.0000	0.0000	0.0000	0
222	JOHNSON	4	A	0	0.0000	0.0000	0.0000	0.0000	0
223	JOHNSTON	4	A	0	0.0000	0.0000	0.0000	0.0000	0
224	JONES	4	A	0	0.0000	0.0000	0.0000	0.0000	0
225	JULIAN	4	A	0	0.0000	0.0000	0.0000	0.0000	0
226	ROCKELMA	4	A	0	0.0000	0.0000	0.0000	0.0000	0
227	HASRI	4	A	0	0.0000	0.0000	0.0000	0.0000	0
228	HATINGLY	4	A	0	0.0000	0.0000	0.0000	0.0000	0

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DATA FROM DELPHI BY BUILDING TYPE AND POLICY

DES	ID	BUDGTYPE	POLICY	MNV	MNVI	MNVII	MNVIII	BNIX	COM2
206	SPICER	6	A	00	0.015	0.075	0.130	0.190	1.00
207	WHITMAN	6	B	00	0.035	0.070	0.135	0.135	0.00
208	ARNOLD	6	B	00	0.035	0.070	0.135	0.135	0.00
209	ASAKUFA	6	B	00	0.035	0.075	0.130	0.190	0.00
210	BACHRACH	6	B	00	0.000	0.035	0.050	0.150	0.00
211	BLANCK	6	B	00	0.035	0.075	0.130	0.190	0.00
212	BRAUNS	6	B	00	0.035	0.075	0.130	0.190	0.00
213	DONOVAN	6	B	00	0.035	0.075	0.130	0.190	0.00
214	JOHNSCN	6	B	00	0.035	0.075	0.130	0.190	0.00
215	JOHNSTON	6	B	00	0.035	0.075	0.130	0.190	0.00
216	JONES	6	B	00	0.035	0.075	0.130	0.190	0.00
217	JULIAN	6	B	00	0.035	0.075	0.130	0.190	0.00
218	KOCKELM	6	B	00	0.035	0.075	0.130	0.190	0.00
219	HASRI	6	B	00	0.035	0.075	0.130	0.190	0.00
220	HATINGLY	6	B	00	0.035	0.075	0.130	0.190	0.00
221	HCCOY	6	B	00	0.035	0.075	0.130	0.190	0.00
222	HEIMEF	6	B	00	0.035	0.075	0.130	0.190	0.00
223	PETAK	6	B	00	0.035	0.075	0.130	0.190	0.00
224	SPICER	6	B	00	0.035	0.075	0.130	0.190	0.00
225	WHITMAN	6	B	00	0.035	0.075	0.130	0.190	0.00
226	ARNOLD	6	C	00	0.035	0.075	0.130	0.190	0.00
227	ASAKUFA	6	C	00	0.035	0.075	0.130	0.190	0.00
228	BACHRACH	6	C	00	0.035	0.075	0.130	0.190	0.00
229	BLANCK	6	C	00	0.035	0.075	0.130	0.190	0.00
230	BRAUNS	6	C	00	0.035	0.075	0.130	0.190	0.00
231	DONOVAN	6	C	00	0.035	0.075	0.130	0.190	0.00
232	JOHNSCN	6	C	00	0.035	0.075	0.130	0.190	0.00
233	JOHNSTON	6	C	00	0.035	0.075	0.130	0.190	0.00
234	JONES	6	C	00	0.035	0.075	0.130	0.190	0.00
235	JULIAN	6	C	00	0.035	0.075	0.130	0.190	0.00
236	KOCKELM	6	C	00	0.035	0.075	0.130	0.190	0.00
237	HASRI	6	C	00	0.035	0.075	0.130	0.190	0.00
238	HATINGLY	6	C	00	0.035	0.075	0.130	0.190	0.00
239	HCCOY	6	C	00	0.035	0.075	0.130	0.190	0.00
240	HEIMEF	6	C	00	0.035	0.075	0.130	0.190	0.00
241	PETAK	6	C	00	0.035	0.075	0.130	0.190	0.00
242	SPICER	6	C	00	0.035	0.075	0.130	0.190	0.00
243	WHITMAN	6	C	00	0.035	0.075	0.130	0.190	0.00
244	ARNOLD	7	A	00	0.035	0.075	0.130	0.190	0.00
245	ASAKUFA	7	A	00	0.035	0.075	0.130	0.190	0.00
246	BACHRACH	7	A	00	0.035	0.075	0.130	0.190	0.00
247	BLANCK	7	A	00	0.035	0.075	0.130	0.190	0.00
248	BRAUNS	7	A	00	0.035	0.075	0.130	0.190	0.00
249	DONOVAN	7	A	00	0.035	0.075	0.130	0.190	0.00
250	JOHNSCN	7	A	00	0.035	0.075	0.130	0.190	0.00
251	JOHNSTON	7	A	00	0.035	0.075	0.130	0.190	0.00
252	JONES	7	A	00	0.035	0.075	0.130	0.190	0.00
253	JULIAN	7	A	00	0.035	0.075	0.130	0.190	0.00
254	KOCKELM	7	A	00	0.035	0.075	0.130	0.190	0.00
255	HASRI	7	A	00	0.035	0.075	0.130	0.190	0.00
256	HATINGLY	7	A	00	0.035	0.075	0.130	0.190	0.00
257	HCCOY	7	A	00	0.035	0.075	0.130	0.190	0.00
258	HEIMEF	7	A	00	0.035	0.075	0.130	0.190	0.00
259	PETAK	7	A	00	0.035	0.075	0.130	0.190	0.00
260	SPICER	7	A	00	0.035	0.075	0.130	0.190	0.00
261	WHITMAN	7	A	00	0.035	0.075	0.130	0.190	0.00
262	ARNOLD	7	B	00	0.035	0.075	0.130	0.190	0.00

DATA FROM DELPHI BY BUILDING TYPE AND POLICY

OBS	ID	BLDGTYPE	POLICY	MHV	MHVI	MHVII	MHVIII	MHIX	COB2
343	ASAKUFA	7	B	0	0.005	0.025	0.080	0.115	0
344	BACHRACH	7	B	0	0.000	0.000	0.010	0.050	0
345	BLANCH	7	B	0
346	BRAUNS	7	B	0
347	DONOVAN	7	B	0	0.005	0.025	0.080	0.115	0
348	JOHNSON	7	B	0	0.005	0.025	0.080	0.115	0
349	JOHNSTON	7	B	0	0.005	0.025	0.080	0.115	0
350	JONES	7	B	0	0.005	0.025	0.080	0.115	0
351	JULIAN	7	B	0	0.000	0.000	0.000	0.000	0
352	KOCKEIM	7	B	0	0.000	0.000	0.000	0.000	0
353	KOCHER	7	B	0	0.000	0.000	0.000	0.000	0
354	MASRI	7	B	0	0.000	0.000	0.000	0.000	0
355	NATLINGLY	7	B	0	0.000	0.000	0.000	0.000	0
356	MCCOY	7	B	0	0.000	0.000	0.000	0.000	0
357	REINER	7	B	0	0.000	0.000	0.000	0.000	0
358	REINER	7	B	0	0.005	0.025	0.080	0.115	0
359	SPICER	7	B	0	0.005	0.025	0.080	0.115	0
360	WHITMAN	7	B	0	0.005	0.025	0.080	0.115	0
361	ARNOLD	7	B	0	0.005	0.025	0.080	0.115	0
362	ASAKUFA	7	B	0	0.005	0.025	0.080	0.115	0
363	BACHRACH	7	B	0
364	BLANCH	7	B	0
365	BRAUNS	7	B	0
366	DONOVAN	7	B	0	0.005	0.025	0.080	0.115	0
367	JOHNSON	7	B	0	0.005	0.025	0.080	0.115	0
368	JOHNSTON	7	B	0	0.005	0.025	0.080	0.115	0
369	JONES	7	B	0	0.005	0.025	0.080	0.115	0
370	JULIAN	7	B	0	0.000	0.000	0.000	0.000	0
371	KOCKEIM	7	B	0	0.000	0.000	0.000	0.000	0
372	KOCHER	7	B	0	0.000	0.000	0.000	0.000	0
373	MASRI	7	B	0	0.000	0.000	0.000	0.000	0
374	NATLINGLY	7	B	0	0.000	0.000	0.000	0.000	0
375	MCCOY	7	B	0	0.000	0.000	0.000	0.000	0
376	REINER	7	B	0	0.000	0.000	0.000	0.000	0
377	REINER	7	B	0	0.005	0.025	0.080	0.115	0
378	SPICER	7	B	0	0.005	0.025	0.080	0.115	0
379	WHITMAN	7	B	0	0.005	0.025	0.080	0.115	0

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MEANS STD DEVIATION ETC FOR RM INTENSITY/BLDG TYPE/POLICY

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
----- BLDGTYPE=1 POLICY=A -----									
RMV	14	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	07.991
RMVI	14	0.02578571	0.00515628	0.0000000	0.1000000	0.00135126	0.0010000	0.0002557	31.284
RMVII	14	0.0346286	0.0103452	0.0000000	0.0400000	0.0026223	0.0045000	0.001777	23.031
RMVIII	14	0.0900000	0.0200000	0.0000000	0.1000000	0.0050000	1.5000000	0.0043866	10.031
RMIX	14	0.14142857	0.01934833	0.0000000	0.1500000	0.0033069	1.5000000	0.0039336	
----- BLDGTYPE=1 POLICY=B -----									
RMV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	09.311
RMVI	15	0.00597093	0.0057093	0.0000000	0.0100000	0.00130931	0.0000000	0.0002737	35.201
RMVII	15	0.01632559	0.0080000	0.0000000	0.0100000	0.0026689	0.0050000	0.0027510	24.001
RMVIII	15	0.02996667	0.0113242	0.0000000	0.0100000	0.0030736	1.2500000	0.0045162	13.081
RMIX	15	0.08533333	0.01898886	0.0000000	0.1000000	0.00507499	1.2500000	0.0036058	
----- BLDGTYPE=1 POLICY=C -----									
RMV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	04.311
RMVI	15	0.0060000	0.0057093	0.0000000	0.0100000	0.00130931	0.0000000	0.0002557	31.284
RMVII	15	0.0280000	0.0113242	0.0000000	0.0100000	0.0026689	0.0000000	0.0002557	31.284
RMVIII	15	0.0730000	0.0150000	0.0000000	0.0100000	0.0030736	0.0000000	0.0002557	31.284
RMIX	15	0.11933333	0.03909533	0.0000000	0.1000000	0.00507499	1.7500000	0.0015531	
----- BLDGTYPE=2 POLICY=A -----									
RMV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	06.000
RMVI	15	0.01733333	0.01428571	0.0000000	0.0100000	0.00310931	0.0000000	0.0002232	18.000
RMVII	15	0.12333333	0.0293320	0.0000000	0.0500000	0.0060117	1.8500000	0.0018052	10.000
RMVIII	15	0.23566667	0.0293320	0.0000000	0.2800000	0.0060117	3.5350000	0.0018052	10.000
RMIX	15	0.23566667	0.03802568	0.0000000	0.2800000	0.0060117	3.5350000	0.0018052	10.000
----- BLDGTYPE=2 POLICY=B -----									
RMV	16	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	07.000
RMVI	16	0.02532250	0.0129072	0.0000000	0.0400000	0.00396518	0.0050000	0.0007526	29.000
RMVII	16	0.09532250	0.0274509	0.0000000	0.1000000	0.0085127	1.5200000	0.0009073	17.000
RMVIII	16	0.17466667	0.03935238	0.0000000	0.2800000	0.0066125	3.8200000	0.0011128	13.000
RMIX	16	0.25466667	0.0335238	0.0000000	0.2800000	0.0066125	3.8200000	0.0011128	13.000
----- BLDGTYPE=2 POLICY=C -----									
RMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	11.000
RMVI	13	0.0130692	0.0129072	0.0000000	0.0100000	0.00101357	0.0000000	0.0002189	33.000
RMVII	13	0.07233333	0.0203320	0.0000000	0.0200000	0.0017844	0.0050000	0.0003623	28.000
RMVIII	13	0.12576923	0.0306732	0.0000000	0.2800000	0.0031666	1.6350000	0.0002270	28.000
RMIX	13	0.19576923	0.05544690	0.0000000	0.2800000	0.0031666	2.5450000	0.0002270	28.000
----- BLDGTYPE=3 POLICY=A -----									
RMV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	00.000
RMVI	15	0.01233333	0.0090000	0.0000000	0.0100000	0.0021319	0.0000000	0.0000838	27.000
RMVII	15	0.03966667	0.0100000	0.0000000	0.0900000	0.0031096	1.0000000	0.0002370	27.000
RMVIII	15	0.13966667	0.0850000	0.0000000	0.1600000	0.0031096	3.0000000	0.0002370	27.000
RMIX	15	0.20500000	0.02506348	0.0000000	0.1500000	0.00661888	3.0000000	0.0002370	27.000

MEANS STD DEVIATION ETC FOR MR INTENSITY/BLDG TYPE/POLICY

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.P.
----- BLDGTYPE=3 POLICY=B -----									
MRV I	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	97.84
MRV II	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	97.84
MRV III	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	97.84
MRV IV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	97.84
----- BLDGTYPE=3 POLICY=C -----									
MRV I	14	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	102.18
MRV II	14	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	102.18
MRV III	14	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	102.18
MRV IV	14	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	102.18
----- BLDGTYPE=4 POLICY=A -----									
MRV I	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	91.01
MRV II	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	91.01
MRV III	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	91.01
MRV IV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	91.01
----- BLDGTYPE=4 POLICY=B -----									
MRV I	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	95.52
MRV II	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	95.52
MRV III	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	95.52
MRV IV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	95.52
----- BLDGTYPE=4 POLICY=C -----									
MRV I	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	75.89
MRV II	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	75.89
MRV III	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	75.89
MRV IV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	75.89
----- BLDGTYPE=5 POLICY=A -----									
MRV I	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	19.24
MRV II	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	19.24
MRV III	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	19.24
MRV IV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	19.24
----- BLDGTYPE=5 POLICY=B -----									
MRV I	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	99.07
MRV II	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	99.07
MRV III	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	99.07
MRV IV	15	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	99.07

MEANS STD DEVIATION ETC FOR MH INTENSITY/BLDG TYPE/POLICY

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
----- BLDGTYPE=5 POLICY=C -----									
MV I	16	0.03000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	160.8371
MV II	16	0.03718350	0.01210630	0.00000000	0.00000000	0.00202659	0.11500000	0.00014656	190.0771
MV III	16	0.03656350	0.03585707	0.00000000	0.00000000	0.00396427	0.58500000	0.00128573	99.3960
MV IV	16	0.10933350	0.05487011	0.04000000	0.20000000	0.01371733	1.77500000	0.00301073	37.8255
MV V	17	0.18441176	0.06939206	0.08500000	0.30000000	0.01681005	3.13500000	0.00481526	
----- BLDGTYPE=6 POLICY=A -----									
MV I	16	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	65.3153
MV II	16	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	65.3153
MV III	16	0.03959350	0.03630311	0.00000000	0.00000000	0.00157578	0.15100000	0.00003973	10.8272
MV IV	16	0.17183500	0.02651258	0.00000000	0.00000000	0.00481454	0.95500000	0.00048562	14.1274
MV V	16	0.17062500	0.02205916	0.00000000	0.00000000	0.00428926	1.73300000	0.00063292	
----- BLDGTYPE=6 POLICY=B -----									
MV I	16	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	72.3241
MV II	16	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	72.3241
MV III	16	0.09337500	0.06800074	0.00000000	0.00000000	0.00370019	0.15000000	0.00046255	14.3398
MV IV	16	0.17812500	0.01938042	0.00000000	0.00000000	0.00462974	1.88000000	0.00037323	13.3372
MV V	16	0.17666667	0.01986262	0.00000000	0.00000000	0.00312651	2.62000000	0.00039452	
----- BLDGTYPE=6 POLICY=C -----									
MV I	16	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	85.3900
MV II	16	0.08437500	0.09723352	0.00000000	0.00000000	0.00180962	0.11500000	0.00052200	64.3290
MV III	16	0.09669750	0.03065522	0.00000000	0.00000000	0.00166486	0.76000000	0.00094800	42.3290
MV IV	16	0.09669750	0.04728846	0.00000000	0.00000000	0.00105182	1.59000000	0.00118823	30.8214
MV V	16	0.14812500	0.04505390	0.00000000	0.00000000	0.00112672	2.37000000	0.00202358	
----- BLDGTYPE=7 POLICY=A -----									
MV I	17	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	130.5680
MV II	17	0.00776871	0.02469966	0.00000000	0.00000000	0.00097336	0.03500000	0.00006697	174.3292
MV III	17	0.03233529	0.00983111	0.00000000	0.00000000	0.00071997	0.22500000	0.00096699	32.3504
MV IV	17	0.03352941	0.02104991	0.00000000	0.00000000	0.00070481	1.59000000	0.00044301	
----- BLDGTYPE=7 POLICY=B -----									
MV I	15	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	104.8227
MV II	15	0.02666667	0.02587399	0.00000000	0.00000000	0.00066667	0.04000000	0.00036667	104.8227
MV III	15	0.01633333	0.01114854	0.00000000	0.00000000	0.00038413	0.27000000	0.00030667	43.1522
MV IV	15	0.08666667	0.02448999	0.00000000	0.00000000	0.00081943	0.86000000	0.00070697	
MV V	15	0.10250000	0.03457155	0.00000000	0.00000000	0.00023168	1.43500000	0.00119519	
----- BLDGTYPE=7 POLICY=C -----									
MV I	15	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	119.6927
MV II	15	0.02333333	0.01238799	0.00000000	0.00000000	0.00066667	0.03500000	0.00036667	99.8930
MV III	15	0.01633333	0.01114854	0.00000000	0.00000000	0.00038413	0.27000000	0.00030667	39.8930
MV IV	15	0.08666667	0.02448999	0.00000000	0.00000000	0.00081943	0.86000000	0.00070697	
MV V	15	0.09666667	0.02448999	0.00000000	0.00000000	0.00081943	1.32500000	0.00071777	

A P P E N D I X A 3

Damage curves -- cycle II

7 SAS LOG 05 SAS 82.4 VS2/MVS JOB GKUA0501 STEP SAS PROC

NOTE: THE PROCEDURE MEANS USED 0.14 SECONDS AND 232K AND PRINTED PAGE 42
NOTE: SAS USED 240K MEMORY.

NOTE: SAS INSTITUTE INC.
SAS CIRCLE
PO BOX 8000
CARY, N.C. 27511-8000

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DATA FROM TASK ONE/SETTING ONE

FRIDAY, AUGUST 9, 1985 1

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	22.5	26.00	32.5	35.0	35.0	0
2	ARNOLD	0.0	0.50	2.5	8.0	11.8	0
3	ASAKURA	0.0	0.50	3.0	9.0	13.0	0
4	BACHRACH	0.0	1.00	3.0	9.3	13.0	0
5	BRAUN	0.0	1.50	5.0	16.0	21.5	0
6	GALLAGH	0.0	0.45	2.5	9.0	11.5	0
7	JOHNSON	0.0	0.50	3.0	9.5	12.5	0
8	JOHNSONR	0.0	0.45	2.5	8.0	13.5	0
9	KOCKELMN	0.0	1.00	3.0	10.0	16.5	0
10	MASRI	0.0	5.00	10.0	15.0	20.0	0
11	MTTINGLY	0.0	2.50	6.5	11.5	15.0	0
12	SANDY	0.0	0.45	3.0	10.0	15.5	1
13	WALLACE	0.0	0.45	2.5	10.0	15.0	0
14	WHITMAN	0.0	0.00	1.0	3.0	6.0	1

DATA FROM TASK ONE/SETTING ONE

FRIDAY, AUGUST 9, 1985 2

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	13	13	100.000	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	1	1	7.692	7.692
0.45	4	5	30.769	38.462
0.5	3	8	23.077	61.538
1	2	10	15.385	76.923
1.5	1	11	7.692	84.615
2.5	1	12	7.692	92.308
5	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
1	1	1	7.692	7.692
2.5	4	5	30.769	38.462
3	5	10	38.462	76.923
5	1	11	7.692	84.615
6.5	1	12	7.692	92.308
10	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
3	1	1	7.692	7.692
8	2	3	15.385	23.077
9	2	5	15.385	38.462
9.3	1	6	7.692	46.154
9.5	1	7	7.692	53.846
10	3	10	23.077	76.923
11.5	1	11	7.692	84.615
15	1	12	7.692	92.308
16	1	13	7.692	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
6	1	1	7.692	7.692
11.5	1	2	7.692	15.385
11.8	1	3	7.692	23.077
12.5	1	4	7.692	30.769
13	2	6	15.385	46.154
13.5	1	7	7.692	53.846
15	2	9	15.385	69.231
15.5	1	10	7.692	76.923
16.5	1	11	7.692	84.615
20	1	12	7.692	92.308
21.5	1	13	7.692	100.000

///

DATA FROM TASK ONE/SETTING ONE

FRIDAY, AUGUST 9, 1985 3

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	1.1000000	1.33322916	0.0000000	5.0000000	0.36977124	14.3000000	1.7775000	121.203
MMVII	13	3.65384615	2.32185823	1.0000000	10.0000000	0.64396761	47.5000000	5.39102564	63.546
MMVIII	13	9.86923077	3.19853733	3.0000000	16.0000000	0.88711464	128.3000000	10.23064103	32.409
MMIX	13	14.21538462	3.89162480	6.0000000	21.5000000	1.07934252	184.8000000	15.14474359	27.376

DATA FROM TASK ONE/SETTING TWO

FRIDAY AUGUST 9, 1985 4

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	12	16.0	21.0	25.5	30.5	0
2	ASAKIRA	0	3.0	10.0	17.5	24.0	0
3	ARNOLD	0	2.0	7.5	14.0	20.0	0
4	BACHRACH	0	6.0	12.5	20.5	27.0	0
5	BRAUN	0	3.5	11.0	20.5	29.0	0
6	GALLAGH	0	2.3	7.5	14.0	20.5	0
7	JOHNSONG	1	3.0	10.0	17.0	25.0	0
8	JOHNSONR	0	2.3	7.5	13.5	22.3	0
9	KOCKELMN	0	3.5	8.5	15.5	22.0	0
10	MASRI	0	6.5	12.5	18.5	25.0	0
11	MTTINGLY	5	10.0	17.5	23.0	27.5	0
12	SANDY	0	2.3	8.0	16.2	23.0	1
13	WALLACE	0	2.3	10.0	20.0	30.0	0
14	WHITMAN	0	0.5	2.5	6.0	9.0	0

DATA FROM TASK ONE/SETTING TWO

FRIDAY, AUGUST 9, 1985 5

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	11	11	84.615	84.615
1	1	12	7.692	92.308
5	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0.5	1	1	7.692	7.692
2	1	2	7.692	15.385
2.3	4	6	30.769	46.154
3	2	8	15.385	61.538
3.5	2	10	15.385	76.923
6	1	11	7.692	84.615
6.5	1	12	7.692	92.308
10	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
2.5	1	1	7.692	7.692
7.5	3	4	23.077	30.769
8	1	5	7.692	38.462
9.5	1	6	7.692	46.154
10	3	9	23.077	69.231
11	1	10	7.692	76.923
12.5	2	12	15.385	92.308
17.5	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
6	1	1	7.692	7.692
13.5	1	2	7.692	15.385
14	2	4	15.385	30.769
15.5	1	5	7.692	38.462
16.2	1	6	7.692	46.154
17	1	7	7.692	53.846
17.5	1	8	7.692	61.538
18.5	1	9	7.692	69.231
20	1	10	7.692	76.923
20.5	2	12	15.385	92.308
23	1	13	7.692	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
9	1	1	7.692	7.692
20	1	2	7.692	15.385
20.5	1	3	7.692	23.077
22	1	4	7.692	30.769
22.3	1	5	7.692	38.462
23	1	6	7.692	46.154
24	1	7	7.692	53.846
25	2	9	15.385	69.231
27	1	10	7.692	76.923
27.5	1	11	7.692	84.615
29	1	12	7.692	92.308
30	1	13	7.692	100.000

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DATA FROM TASK ONE/SETTING TWO

FRIDAY, AUGUST 9, 1985

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
MMV	13	0.46153846	1.39136531	0.00000000	5.00000000	0.38589531	6.00000000	1.93589744	301.462
MMVI	13	3.63076923	2.49645903	0.50000000	10.00000000	0.69239316	47.20000000	6.23230769	68.758
MMVII	13	9.69230769	3.50914556	2.50000000	17.50000000	0.97326187	126.00000000	12.31410256	36.205
MMVIII	13	16.63076923	4.32403835	6.00000000	23.00000000	1.19927246	216.20000000	18.69730769	26.000
MMIX	13	23.40769231	5.32939358	9.00000000	30.00000000	1.47810783	304.30000000	28.40243590	22.768

DATA FROM TASK ONE/SETTING THREE

FRIDAY, AUGUST 9, 1985 7

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	14.5	17.00	22.5	25.5	28.0	0
2	ASAKURA	0.0	0.09	4.0	10.0	14.0	0
3	ARNOLD	0.0	0.80	3.0	9.0	13.4	1
4	BACHRACH	0.0	2.20	5.0	11.0	14.0	0
5	BRAUN	0.0	1.50	5.0	14.5	20.0	0
6	GALLAGH	0.0	1.00	3.5	9.5	12.0	0
7	JOHNSONG	0.0	2.00	7.0	15.0	23.0	0
8	JOHNSONR	0.0	0.80	3.0	8.5	14.2	0
9	KOCKELMN	0.0	1.50	6.0	12.0	16.0	0
10	MASRI	0.0	0.80	3.0	12.2	20.0	0
11	MTTINGLY	0.0	3.50	6.0	11.0	16.0	0
12	SANDY	0.0	1.00	4.5	12.0	16.5	1
13	WALLACE	0.0	0.80	5.0	15.0	20.0	0
14	WHITMAN	0.0	0.50	1.5	4.5	9.0	0

DATA FROM TASK ONE/SETTING THREE

FRIDAY, AUGUST 9 1985 8

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	13	13	100.000	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0.09	1	1	7.692	7.692
0.5	1	2	7.692	15.385
0.8	4	6	30.769	46.154
1	2	8	15.385	61.538
1.5	2	10	15.385	76.923
2	1	11	7.692	84.615
2.2	1	12	7.692	92.308
3.5	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
1.5	1	1	7.692	7.692
3	3	4	23.077	30.769
3.5	1	5	7.692	38.462
4	1	6	7.692	46.154
4.5	1	7	7.692	53.846
5	3	10	23.077	76.923
6	2	12	15.385	92.308
7	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
4.5	1	1	7.692	7.692
8.5	1	2	7.692	15.385
9	1	3	7.692	23.077
9.5	1	4	7.692	30.769
10	1	5	7.692	38.462
11	2	7	15.385	53.846
12	2	9	15.385	69.231
12.2	1	10	7.692	76.923
14.5	1	11	7.692	84.615
15	2	13	15.385	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
9	1	1	7.692	7.692
12	1	2	7.692	15.385
13.4	1	3	7.692	23.077
14	2	5	15.385	38.462
14.2	1	6	7.692	46.154
15	2	8	15.385	61.538
16.5	1	9	7.692	69.231
20	3	12	23.077	92.308
23	1	13	7.692	100.000

DATA FROM TASK ONE/SETTING THREE

FRIDAY, AUGUST 9, 1985 9

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	1.26846154	0.88934289	0.09000000	3.50000000	0.24665934	16.49000000	0.79093077	70.112
MMVII	13	4.34615385	1.53276188	1.50000000	7.00000000	0.42511166	56.50000000	2.34935897	35.267
MMVIII	13	11.09230769	2.93441122	4.50000000	15.00000000	0.81385924	144.20000000	8.61076923	26.454
MMIX	13	16.00769231	3.87566371	9.00000000	23.00000000	1.07491571	208.10000000	15.02076923	24.211

DATA FROM TASK ONE/SETTING FOUR

FRIDAY, AUGUST 9, 1985 10

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	21	23.50	27.0	30.0	34.0	0
2	ASAKURA	0	1.75	5.0	12.0	17.0	0
3	ARNOLD	0	0.00	4.0	10.0	14.0	0
4	BACHRACH	0	2.00	5.5	11.5	16.0	0
5	BRAUN	0	2.00	5.0	12.5	19.0	0
6	GALLAGH	0	1.50	4.0	10.0	14.2	0
7	JOHNSONG	0	2.50	6.0	12.5	17.5	0
8	JOHNSONR	0	1.50	4.0	10.0	18.5	0
9	KOCKELMN	0	1.00	4.5	10.0	16.0	0
10	MASRI	0	2.00	6.0	12.0	16.0	0
11	MITTINGLY	0	3.50	6.0	10.5	14.0	0
12	SANDY	0	1.50	4.0	10.0	15.0	1
13	WALLACE	0	1.50	10.0	20.0	25.5	0
14	WHITMAN	0	0.50	1.5	3.5	9.0	0

DATA FROM TASK ONE/SETTING FOUR

FRIDAY, AUGUST 9, 1985 11

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	13	13	100.000	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	1	1	7.692	7.692
0.5	1	2	7.692	15.385
1	1	3	7.692	23.077
1.5	4	7	30.769	53.846
1.75	1	8	7.692	61.538
2	3	11	23.077	84.615
2.5	1	12	7.692	92.308
3.5	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
1.5	1	1	7.692	7.692
4	4	5	30.769	38.462
4.5	1	6	7.692	46.154
5	2	8	15.385	61.538
5.5	1	9	7.692	69.231
6	3	12	23.077	92.308
10	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
3.5	1	1	7.692	7.692
10	5	6	38.462	46.154
10.5	1	7	7.692	53.846
11.5	1	8	7.692	61.538
12	2	10	15.385	76.923
12.5	2	12	15.385	92.308
20	1	13	7.692	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
9	1	1	7.692	7.692
14	2	3	15.385	23.077
14.2	1	4	7.692	30.769
15	1	5	7.692	38.462
16	3	8	23.077	61.538
17	1	9	7.692	69.231
17.5	1	10	7.692	76.923
18.5	1	11	7.692	84.615
19	1	12	7.692	92.308
25.5	1	13	7.692	100.000

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DATA FROM TASK ONE/SETTING FOUR

FRIDAY, AUGUST 9, 1985 12

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	1.63461538	0.86971849	0.0000000	3.5000000	0.24121651	21.2500000	0.75641026	53.206
MMVII	13	5.03846154	1.93069006	1.5000000	10.0000000	0.53547708	65.5000000	3.72756410	38.319
MMVIII	13	11.11538462	3.52463855	3.5000000	20.0000000	0.97755885	144.5000000	12.42307692	31.710
MMIX	13	16.28461538	3.75207635	9.0000000	25.5000000	1.04063874	211.7000000	14.07807692	23.041

DATA FROM TASK ONE/SETTING FIVE

FRIDAY, AUGUST 9, 1985 13

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	10	13.00	16.5	20.5	25.0	0
2	ARNOLD	0	2.30	7.0	12.6	18.0	0
3	ASAKURA	0	2.75	8.0	15.5	22.5	0
4	BACHRACH	0	3.25	8.5	15.0	20.0	0
5	BRAUN	0	3.50	10.0	18.0	35.0	0
6	GALLAGH	0	2.50	7.0	13.0	18.0	0
7	JOHNSONG	0	3.50	10.0	20.0	27.0	0
8	JOHNSONR	0	2.50	7.0	12.5	20.0	0
9	KOCKELMN	0	2.50	7.5	14.0	20.0	0
10	MASRI	0	6.00	11.5	17.0	22.0	0
11	MITTINGLY	0	5.00	9.0	14.5	20.0	0
12	SANDY	0	2.50	7.0	14.0	20.0	1
13	WALLACE	0	2.50	10.0	20.0	25.0	0
14	WHITMAN	0	0.50	2.5	4.5	7.0	0

DATA FROM TASK ONE/SETTING FIVE

FRIDAY, AUGUST 9, 1986 14

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	13	13	100.000	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0.5	1	1	7.692	7.692
2.3	1	2	7.642	15.385
2.5	5	7	38.462	53.846
2.75	1	8	7.692	61.538
3.25	1	9	7.692	69.231
3.5	2	11	15.385	84.615
5	1	12	7.692	92.308
6	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
2.5	1	1	7.692	7.692
7	4	5	30.769	38.462
7.5	1	6	7.692	46.154
8	1	7	7.692	53.846
8.5	1	8	7.692	61.538
9	1	9	7.692	69.231
10	3	12	23.077	92.308
11.5	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
4.5	1	1	7.692	7.692
12.5	1	2	7.692	15.385
12.8	1	3	7.692	23.077
13	1	4	7.692	30.769
14	2	6	15.385	46.154
14.5	1	7	7.692	53.846
15	1	8	7.692	61.538
15.5	1	9	7.692	69.231
17	1	10	7.692	76.923
18	1	11	7.692	84.615
20	2	13	15.385	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
7	1	1	7.692	7.692
18	2	3	15.385	23.077
20	5	8	38.462	61.538
22	1	9	7.692	69.231
22.5	1	10	7.692	76.923
25	1	11	7.692	84.615
27	1	12	7.692	92.308
35	1	13	7.692	100.000

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DATA FROM TASK ONE/SETTING FIVE

FRIDAY, AUGUST 9, 1985 15

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V.
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	3.02307692	1.34189160	0.5000000	6.0000000	0.37217377	39.3000000	1.80067308	44.388
MMVII	13	8.07692308	2.23463414	2.5000000	11.5000000	0.61977600	105.0000000	4.99358974	27.667
MMVIII	13	14.67692308	3.97369556	4.5000000	20.0000000	1.10210485	190.8000000	15.79025641	27.074
MMIX	13	21.11538462	6.25883990	7.0000000	35.0000000	1.73588986	274.5000000	39.17307692	29.641

DATA FROM TASK ONE/FITTING SIX

FRIDAY, AUGUST 9, 1985 16

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	26.5	28.0	32.0	35.0	35.0	0
2	ARNOLD	0.0	1.0	3.9	9.7	14.0	0
3	ASAKURA	0.0	1.0	4.5	10.5	15.0	0
4	BACHRACH	0.0	1.5	5.0	12.0	16.5	0
5	BRAUN	0.0	2.0	6.5	14.0	20.0	0
6	GALLAGH	0.0	1.0	4.0	10.0	13.0	0
7	JOHNSONG	0.0	2.5	10.0	20.0	25.0	0
8	JOHNSONR	0.0	1.0	4.0	9.5	17.5	0
9	KOCKELMN	0.0	1.0	4.0	10.0	15.0	0
10	MASRI	0.0	4.5	9.0	15.0	20.0	0
11	MTTINGLY	2.0	6.5	10.5	16.0	20.0	0
12	SANDY	0.0	1.0	4.5	12.0	20.0	1
13	WALLACE	0.0	1.0	7.5	15.0	20.0	0

DATA FROM TASK ONE/SETTING SIX

FRIDAY, AUGUST 9, 1985 17

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
0	11	11	91.667	91.667
2	1	12	8.333	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
1	7	7	58.333	58.333
1.5	1	8	8.333	66.667
2	1	9	8.333	75.000
2.5	1	10	8.333	83.333
4.5	1	11	8.333	91.667
6.5	1	12	8.333	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
3.9	1	1	8.333	8.333
4	3	4	25.000	33.333
4.5	2	6	16.667	50.000
5	1	7	8.333	58.333
6.5	1	8	8.333	66.667
7.5	1	9	8.333	75.000
9	1	10	8.333	83.333
10	1	11	8.333	91.667
10.5	1	12	8.333	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
9.5	1	1	8.333	8.333
9.7	1	2	8.333	16.667
10	2	4	16.667	33.333
10.5	1	5	8.333	41.667
12	2	7	16.667	58.333
14	1	8	8.333	66.667
15	2	10	16.667	83.333
16	1	11	8.333	91.667
20	1	12	8.333	100.000

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
13	1	1	8.333	8.333
14	1	2	8.333	16.667
15	2	4	16.667	33.333
16.5	1	5	8.333	41.667
17.5	1	6	8.333	50.000
20	5	11	41.667	91.667
25	1	12	8.333	100.000

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DATA FROM TASK ONE/SETTING SIX

FRIDAY, AUGUST 9, 1985 15

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
MMV	12	0.1666667	0.57735027	0.0000000	2.0000000	0.1666667	2.0000000	0.3333333	346.410
MMVI	12	2.0000000	1.75809815	1.0000000	6.5000000	0.50751922	24.0000000	3.09090909	87.905
MMVII	12	6.1166667	2.51353910	3.9000000	10.5000000	0.72559624	73.4000000	6.31787879	41.093
MMVIII	12	12.8083333	3.24666496	9.5000000	20.0000000	0.93723144	153.7000000	10.54083333	25.348
MMIX	12	18.0000000	3.44436298	13.0000000	25.0000000	0.99430195	216.0000000	11.86363636	19.135

DATA FROM TASK TWO/SETTING ONE

FRIDAY, AUGUST 9, 1985 19

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.30	0.9	1.2
2	ASAKURA	0.0	1.2	5.00	15.0	21.0
3	BACHRACH	0.0	2.5	6.00	11.0	15.0
4	BLANCK	5.0	7.0	11.00	14.0	16.0
5	BRAUN	0.0	0.0	2.50	18.0	13.0
6	GALLAGH					
7	JOHNSONG	0.0	0.5	3.50	9.0	12.5
8	JOHNSONR	0.0	0.5	0.50	1.5	5.0
9	KOCKELMN	0.0	1.0	4.50	12.5	21.0
10	MASRI	0.0	3.0	6.50	10.0	15.0
11	MTTINGLY	2.5	6.0	9.50	14.0	17.5
12	SANDY	0.0	0.0	0.25	3.5	6.0
13	WALLACE	0.0	0.5	2.50	5.0	10.0
14	WHITMAN	0.0	0.5	2.50	9.0	13.5

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DATA FROM TASK TWO/SETTING ONE

FRIDAY, AUGUST 9, 1985 20

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	11	11	84.615	84.615
2.5	1	12	7.692	92.308
5	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	3	3	23.077	23.077
0.5	4	7	30.769	53.846
1	1	8	7.692	61.538
1.2	1	9	7.692	69.231
2.5	1	10	7.692	76.923
3	1	11	7.692	84.615
6	1	12	7.692	92.308
7	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.25	1	1	7.692	7.692
0.3	1	2	7.692	15.385
0.5	1	3	7.692	23.077
2.5	3	6	23.077	46.154
3.5	1	7	7.692	53.846
4.5	1	8	7.692	61.538
5	1	9	7.692	69.231
6	1	10	7.692	76.923
6.5	1	11	7.692	84.615
9.5	1	12	7.692	92.308
11	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.9	1	1	7.692	7.692
1.5	1	2	7.692	15.385
3.5	1	3	7.692	23.077
5	1	4	7.692	30.769
9	2	6	15.385	46.154
10	1	7	7.692	53.846
11	1	8	7.692	61.538
12.5	1	9	7.692	69.231
14	2	11	15.385	84.615
15	1	12	7.692	92.308
18	1	13	7.692	100.000

DATA FROM TASK TWO/SETTING ONE

FRIDAY, AUGUST 9, 1985 21

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1.2	1	1	7.692	7.692
5	1	2	7.692	15.385
6	1	3	7.692	23.077
10	1	4	7.692	30.769
12.5	1	5	7.692	38.462
13	1	6	7.692	46.154
13.5	1	7	7.692	53.846
15	2	9	15.385	69.231
16	1	10	7.692	76.923
17.5	1	11	7.692	84.615
21	2	13	15.385	100.000

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DATA FROM TASK TWO/SETTING ONE

FRIDAY, AUGUST 9, 1985 22

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V
MMV	13	0.57692308	1.49786172	0.00000000	5.00000000	0.41543210	7.50000000	2.24358974	259.629
MMVI	13	1.74615385	2.31034463	0.00000000	7.00000000	0.64077431	22.70000000	5.33769231	132.310
MMVII	13	4.19615385	3.38196575	0.25000000	11.00000000	0.93798853	54.55000000	11.43769231	80.597
MMVIII	13	9.49230769	5.39451288	0.90000000	18.00000000	1.49616868	123.40000000	29.10076923	56.830
MMIX	13	12.82307692	5.96226381	1.20000000	21.00000000	1.65363445	166.70000000	35.54858974	46.496

DATA FROM TASK TWO/SETTING TWO

FRIDAY, AUGUST 9, 1985 23

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.3	1.2	2.5
2	ASAKURA	0.0	6.0	15.0	27.5	40.0
3	BACHRACH	0.0	9.5	17.5	25.0	31.0
4	BLANCK	10.0	13.5	19.0	23.0	29.0
5	BRAUN	0.0	0.0	3.0	23.0	45.0
6	GALLAGH
7	JOHNSON	0.0	2.5	9.0	15.5	22.5
8	JOHNSONR	0.0	0.5	2.0	4.5	10.0
9	KOCKELMN	0.0	6.0	13.0	20.0	26.0
10	MASRI	0.0	4.5	10.0	16.0	22.0
11	MTTINGLY	7.5	13.0	17.5	23.5	29.0
12	SANDY	0.0	1.0	5.0	14.0	21.0
13	WALLACE	0.0	2.0	5.0	10.0	15.0
14	WHITMAN	0.0	2.0	8.5	16.0	26.0

DATA FROM TASK TWO/SETTING TWO

FRIDAY, AUGUST 9, 1985 24

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	11	11	84.615	84.615
7.5	1	12	7.692	92.308
10	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	2	2	15.385	15.385
0.5	1	3	7.692	23.077
1	1	4	7.692	30.769
2	2	6	15.385	46.154
2.5	1	7	7.692	53.846
4.5	1	8	7.692	61.538
6	2	10	15.385	76.923
9.5	1	11	7.692	84.615
13	1	12	7.692	92.308
13.5	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.3	1	1	7.692	7.692
2	1	2	7.692	15.385
3	1	3	7.692	23.077
5	2	5	15.385	38.462
8.5	1	6	7.692	46.154
9	1	7	7.692	53.846
10	1	8	7.692	61.538
13	1	9	7.692	69.231
15	1	10	7.692	76.923
17.5	2	12	15.385	92.308
19	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1.2	1	1	7.692	7.692
4.5	1	2	7.692	15.385
10	1	3	7.692	23.077
14	1	4	7.692	30.769
15.5	1	5	7.692	38.462
16	2	7	15.385	53.846
20	1	8	7.692	61.538
23	2	10	15.385	76.923
23.5	1	11	7.692	84.615
25	1	12	7.692	92.308
27.5	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING TWO

FRIDAY, AUGUST 9, 1985 25

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
2.5	1	1	7.692	7.692
10	1	2	7.692	15.385
15	1	3	7.692	23.077
21	1	4	7.692	30.769
22	1	5	7.692	38.462
22.5	1	6	7.692	46.154
26	2	8	15.385	61.538
29	2	10	15.385	76.923
31	1	11	7.692	84.615
40	1	12	7.692	92.308
45	1	13	7.692	100.000

DATA FROM TASK TWO/SETTING TWO

FRIDAY, AUGUST 9, 1985 26

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
MMV	13	1.34615385	3.32531086	0.00000000	10.00000000	0.92227529	17.50000000	11.05769231	247.023
MMVI	13	4.65384615	4.72310198	0.00000000	13.50000000	1.30995280	60.50000000	22.30769231	101.488
MMVII	13	9.60000000	6.36238949	0.30000000	19.00000000	1.76460935	124.80000000	40.48000000	66.275
MMVIII	13	16.86153846	7.98639869	1.20000000	27.50000000	2.21502846	219.20000000	63.78256410	47.365
MMIX	13	24.53846154	11.38248204	2.50000000	45.00000000	3.15693251	319.00000000	123.56089744	46.386

DATA FROM TASK TWO/SETTING THREE

FRIDAY, AUGUST 9, 1985 27

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.5	1.0	2.0
2	ASAKURA	0.0	2.5	7.5	14.0	22.5
3	BACHRACH	1.0	3.5	8.0	13.5	17.0
4	BLANCK	6.5	8.0	11.0	15.5	19.0
5	BRAUN	0.0	0.0	3.0	19.0	35.0
6	GALLAGH
7	JOHNSONG	0.0	1.5	5.5	9.0	21.5
8	JOHNSONR	0.0	0.5	1.0	2.5	8.0
9	KOCKELMN	0.0	3.5	9.0	15.0	21.0
10	MASRI	0.0	0.8	3.0	9.5	17.0
11	MTTINGLY	1.0	3.5	6.0	11.5	16.0
12	SANDY	0.0	0.0	0.0	2.0	4.5
13	WALLACE	0.0	0.5	3.0	8.0	11.0
14	WHITMAN	0.0	1.0	4.0	14.0	29.5

DATA FROM TASK TWO/SETTING THREE

FRIDAY, AUGUST 9, 1985 28

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	10	10	76.923	76.923
1	2	12	15.385	92.308
6.5	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	3	3	23.077	23.077
0.5	2	5	15.385	38.462
0.8	1	6	7.692	46.154
1	1	7	7.692	53.846
1.5	1	8	7.692	61.538
2.5	1	9	7.692	69.231
3.5	3	12	23.077	92.308
8	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	1	1	7.692	7.692
0.5	1	2	7.692	15.385
1	1	3	7.692	23.077
3	3	6	23.077	46.154
4	1	7	7.692	53.846
5.5	1	8	7.692	61.538
6	1	9	7.692	69.231
7.5	1	10	7.692	76.923
8	1	11	7.692	84.615
9	1	12	7.692	92.308
11	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1	1	1	7.692	7.692
2	1	2	7.692	15.385
2.5	1	3	7.692	23.077
8	1	4	7.692	30.769
9	1	5	7.692	38.462
9.5	1	6	7.692	46.154
11.5	1	7	7.692	53.846
13.5	1	8	7.692	61.538
14	2	10	15.385	76.923
15	1	11	7.692	84.615
15.5	1	12	7.692	92.308
19	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING THREE

FRIDAY, AUGUST 9, 1985 29

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
2	1	1	7.692	7.692
4.5	1	2	7.692	15.385
8	1	3	7.692	23.077
11	1	4	7.692	30.769
16	1	5	7.692	38.462
17	2	7	15.385	53.846
19	1	8	7.692	61.538
21	1	9	7.692	69.231
21.5	1	10	7.692	76.923
22.5	1	11	7.692	84.615
29.5	1	12	7.692	92.308
35	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING THREE

FRIDAY, AUGUST 9, 1985 30

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.65384615	1.79565001	0.00000000	6.50000000	0.49802371	8.50000000	3.22435897	274.629
MMVI	13	1.94615385	2.27582051	0.00000000	8.00000000	0.63118904	25.30000000	5.17935897	116.939
MMVII	13	4.73076923	3.43763112	0.00000000	11.00000000	0.95342733	61.50000000	11.81730769	72.665
MMVIII	13	10.34615385	5.68398560	1.00000000	19.00000000	1.57645396	134.50000000	32.30769231	54.938
MMIX	13	17.23076923	9.35105917	2.00000000	35.00000000	2.59351718	224.00000000	87.44230769	54.270

DATA FROM TASK TWO/SETTING FOUR

FRIDAY, AUGUST 9, 1985 31

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.5	1.0	2.0
2	ASAKURA	0.0	2.5	9.5	19.0	30.0
3	BACHRACH	0.0	3.5	9.0	14.5	20.0
4	BLANCK	16.0	19.5	23.0	26.0	30.0
5	BRAUN	0.0	0.0	2.5	13.0	40.0
6	GALLAGH					
7	JOHNSONG	0.0	2.0	4.5	10.0	15.0
8	JOHNSONR	0.0	0.5	1.0	3.5	10.0
9	KOCKELMN	0.0	2.5	6.5	12.5	19.0
10	MASRI	0.0	3.0	7.5	13.5	19.0
11	MTTINGLY	2.5	5.5	10.0	14.0	20.0
12	SANDY	0.0	1.0	1.5	5.0	6.0
13	WALLACE	0.0	1.5	5.0	15.0	20.0
14	WHITMAN	0.0	1.2	4.5	10.2	15.5

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DATA FROM TASK TWO/SETTING FOUR

FRIDAY, AUGUST 9, 1985 32

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	11	11	84.615	84.615
2.5	1	12	7.692	92.308
16	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	2	2	15.385	15.385
0.5	1	3	7.692	23.077
1	1	4	7.692	30.769
1.2	1	5	7.692	38.462
1.5	1	6	7.692	46.154
2	1	7	7.692	53.846
2.5	2	9	15.385	69.231
3	1	10	7.692	76.923
3.5	1	11	7.692	84.615
5.5	1	12	7.692	92.308
19.5	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.5	1	1	7.692	7.692
1	1	2	7.692	15.385
1.5	1	3	7.692	23.077
2.5	1	4	7.692	30.769
4.5	2	6	15.385	46.154
5	1	7	7.692	53.846
6.5	1	8	7.692	61.538
7.5	1	9	7.692	69.231
9	1	10	7.692	76.923
9.5	1	11	7.692	84.615
10	1	12	7.692	92.308
23	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1	1	1	7.692	7.692
3.5	1	2	7.692	15.385
5	1	3	7.692	23.077
10	1	4	7.692	30.769
10.2	1	5	7.692	38.462
12.5	1	6	7.692	46.154
13	1	7	7.692	53.846
13.5	1	8	7.692	61.538
14	1	9	7.692	69.231
14.5	1	10	7.692	76.923
15	1	11	7.692	84.615
19	1	12	7.692	92.308
26	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING FOUR

FRIDAY, AUGUST 9, 1985 33

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
2	1	1	7.692	7.692
6	1	2	7.692	15.385
10	1	3	7.692	23.077
15	1	4	7.692	30.769
15.5	1	5	7.692	38.462
19	2	7	15.385	53.846
20	3	10	23.077	76.923
30	2	12	15.385	92.308
40	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING FOUR

FRIDAY, AUGUST 9, 1985 34

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C V
NNV	13	1.42307692	4.43398877	0.00000000	16.00000000	1.22976722	18.50000000	13.66025641	311.578
NNVI	13	3.28461538	5.10666985	0.00000000	19.50000000	1.41633538	42.70000000	26.07807692	155.472
NNVII	13	6.53846154	5.90360603	0.50000000	23.00000000	1.63736571	85.00000000	34.85256410	90.290
NNVIII	13	12.09230769	6.56460986	1.00000000	26.00000000	1.82069519	157.20000000	43.09410256	54.287
NNIX	13	18.96153846	10.21938831	2.00000000	40.00000000	2.83434835	246.50000000	104.43589744	53.895

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DATA FROM TASK TWO/SETTING FIVE

FRIDAY, AUGUST 9, 1985 15

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0	0.0	0.3	0.8	1.0
2	ASAKURA	0	4.0	14.0	24.0	34.0
3	BACHRACH	0	5.0	11.0	18.5	23.0
4	BLANCK	10	13.0	16.5	20.5	25.0
5	BRAUN	0	0.0	2.5	18.5	35.0
6	GALLAGH
7	JOHNSONG	0	3.0	9.0	17.0	25.0
8	JOHNSONR	0	0.5	1.5	5.0	12.5
9	KOCKELMN	0	4.5	9.0	16.0	23.5
10	MASRI	0	4.5	9.5	14.5	19.0
11	MTTINGLY	3	4.7	9.0	14.2	20.0
12	SANDY	0	0.5	2.0	4.5	6.0
13	WALLACE	0	2.5	5.0	10.0	15.0
14	WHITMAN	0	2.0	7.0	13.5	20.0

DATA FROM TASK TWO/SETTING FIVE

FRIDAY, AUGUST 9, 1985 36

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	11	11	84.615	84.615
3	1	12	7.692	92.308
10	1	13	7.692	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	2	2	15.385	15.385
0.5	2	4	15.385	30.769
2	1	5	7.692	38.462
2.5	1	6	7.692	46.154
3	1	7	7.692	53.846
4	1	8	7.692	61.538
4.5	2	10	15.385	76.923
4.7	1	11	7.692	84.615
5	1	12	7.692	92.308
13	1	13	7.692	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.3	1	1	7.692	7.692
1.5	1	2	7.692	15.385
2	1	3	7.692	23.077
2.5	1	4	7.692	30.769
5	1	5	7.692	38.462
7	1	6	7.692	46.154
9	3	9	23.077	69.231
9.5	1	10	7.692	76.923
11	1	11	7.692	84.615
14	1	12	7.692	92.308
16.5	1	13	7.692	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.8	1	1	7.692	7.692
4.5	1	2	7.692	15.385
5	1	3	7.692	23.077
10	1	4	7.692	30.769
13.5	1	5	7.692	38.462
14.2	1	6	7.692	46.154
14.5	1	7	7.692	53.846
16	1	8	7.692	61.538
17	1	9	7.692	69.231
18.5	2	11	15.385	84.615
20.5	1	12	7.692	92.308
24	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING FIVE

FRIDAY, AUGUST 9, 1981 37

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1	1	1	7.692	7.692
6	1	2	7.692	15.385
12.5	1	3	7.692	23.077
15	1	4	7.692	30.769
19	1	5	7.692	38.462
20	2	7	15.385	53.846
23	1	8	7.692	61.538
23.5	1	9	7.692	69.231
25	2	11	15.385	84.615
34	1	12	7.692	92.308
35	1	13	7.692	100.000

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DATA FROM TASK TWO/SETTING FIVE

FRIDAY, AUGUST 9, 1985 38

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V
MMV	13	1.0000000	2.82842712	0.0000000	10.0000000	0.78446454	13.0000000	8.0000000	282.843
MMVI	13	3.4000000	3.44068792	0.0000000	13.0000000	0.95427513	44.2000000	11.83833333	101.197
MMVII	13	7.40769231	4.96428268	0.3000000	16.5000000	1.37684429	96.3000000	24.64410256	67.015
MMVIII	13	13.61538462	6.80598342	0.8000000	24.0000000	1.88764017	177.0000000	46.32141026	49.987
MMIX	13	19.92307692	9.71005955	1.0000000	35.0000000	2.69308597	259.0000000	94.28525641	48.718

DATA FROM TASK TWO/SETTING SIX

FRIDAY, AUGUST 9, 1985 39

DBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.00	0.5	1.0	1.5
2	ASAKURA	0.0	2.00	7.0	17.5	24.0
3	BACHRACH	0.0	2.00	6.5	17.5	25.0
4	BLANK	5.0	10.00	15.0	19.0	23.0
5	BRAUN	0.0	0.00	2.5	19.0	35.0
6	GALLAGH					
7	JOHNSON	0.0	2.00	9.0	17.5	22.5
8	JOHNSONR	0.0	0.25	0.5	3.5	10.0
9	MASRI	0.0	2.50	7.5	12.5	17.0
10	MTINGLY	1.5	6.50	11.0	16.0	20.0
11	SANDY	0.0	0.00	3.0	5.0	6.0
12	WALLACE	0.0	1.00	5.0	10.0	15.0
13	WHITMAN	0.0	1.00	4.5	12.0	22.0

DATA FROM TASK TWO/SETTING SIX

FRIDAY, AUGUST 9, 1985 40

MMV	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	10	10	83.333	83.333
1.5	1	11	8.333	91.667
5	1	12	8.333	100.000

MMVI	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0	3	3	25.000	25.000
0.25	1	4	8.333	33.333
1	2	6	16.667	50.000
2	3	9	25.000	75.000
2.5	1	10	8.333	83.333
6.5	1	11	8.333	91.667
10	1	12	8.333	100.000

MMVII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
0.5	2	2	16.667	16.667
2.5	1	3	8.333	25.000
3	1	4	8.333	33.333
4.5	1	5	8.333	41.667
5	1	6	8.333	50.000
6.5	1	7	8.333	58.333
7	1	8	8.333	66.667
7.5	1	9	8.333	75.000
9	1	10	8.333	83.333
11	1	11	8.333	91.667
15	1	12	8.333	100.000

MMVIII	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.	.	.
1	1	1	8.333	8.333
3.5	1	2	8.333	16.667
5	1	3	8.333	25.000
10	1	4	8.333	33.333
12	1	5	8.333	41.667
12.5	1	6	8.333	50.000
16	1	7	8.333	58.333
17.5	3	10	25.000	83.333
19	2	12	16.667	100.000

1.1

DATA FROM TASK TWO/SETTING SIX

FRIDAY, AUGUST 9, 1985 41

MMIX	FREQUENCY	CUM FREQ	PERCENT	CUM PERCENT
.	1	.		
1.5	1	1	8.333	8.333
6	1	2	8.333	16.667
10	1	3	8.333	25.000
15	1	4	8.333	33.333
17	1	5	8.333	41.667
20	1	6	8.333	50.000
22	1	7	8.333	58.333
22.5	1	8	8.333	66.667
23	1	9	8.333	75.000
24	1	10	8.333	83.333
25	1	11	8.333	91.667
35	1	12	8.333	100.000

DATA FROM TASK TWO SETTING SIX

FRIDAY, AUGUST 9, 1985 42

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C. V
MMV	12	0.54166667	1.46873993	0.00000000	5.00000000	0.42398870	6.50000000	2.15719697	271.152
MMVI	12	2.27083333	3.02538189	0.00000000	10.00000000	0.87335253	27.25000000	9.15293561	133.228
MMVII	12	6.00000000	4.30116263	0.50000000	15.00000000	1.24163870	72.00000000	18.50000000	71.686
MMVIII	12	12.54166667	6.37986297	1.00000000	19.00000000	1.84170780	150.50000000	40.70265152	50.869
MMIX	12	18.41666667	9.18208070	1.50000000	35.00000000	2.65063838	221.00000000	84.31060606	49.857

A P P E N D I X A 4

Damage curves -- cycle III

DATA FROM TASK ONE/SETTING ONE

CBS	ID	MRV	MRVI	MRVII	MRVIII	MRIX	C
1	BLANCK	22.5	26.00	22.5	26.0	26.0	0
2	ARNOLD	0.0	0.50	2.5	3.0	11.5	0
3	ASAKURA	0.0	0.50	3.0	3.0	19.0	0
4	BACHRACH	0.0	1.00	3.0	3.3	13.0	0
5	BRAUN	0.0	1.50	5.0	16.0	21.5	0
6	GALLAGH	0.0	0.45	2.5	3.0	11.5	0
7	JOHNSON	0.0	0.50	3.0	3.5	12.5	0
8	JOHNSON	0.0	0.45	2.5	3.0	13.5	0
9	KOCKELMAN	0.0	1.00	3.0	10.0	16.5	0
10	MASRI	0.0	5.00	10.0	15.0	20.0	0
11	MITTINGLY	0.0	2.50	6.5	11.5	15.0	0
12	SANDY	0.0	0.45	3.0	10.0	15.5	1
13	MALLACE	0.0	0.45	2.5	10.0	15.0	0
14	WHITMAN	0.0	0.00	1.0	3.0	6.0	1

DATA FROM TASK ONE/SETTING ONE

2

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	1.1000000	1.3332818	0.0000000	8.0000000	0.36877124	14.3000000	1.7775000	121.203
MMVII	13	3.65384615	2.32185823	1.0000000	10.0000000	0.64386761	47.5000000	5.38102564	63.848
MMVIII	13	8.86823077	3.19853733	3.0000000	15.0000000	0.88711464	128.3000000	10.23064103	32.408
MMIX	13	14.21538462	3.89162480	6.0000000	21.5000000	1.07834252	184.8000000	15.14474358	27.376

DATA FROM TASK ONE/SETTING TWO

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	12	16.0	21.0	28.5	30.5	0
2	ASAKURA	0	3.0	10.0	17.5	24.0	0
3	ARNOLD	0	2.0	7.5	14.0	20.0	0
4	BACHRACH	0	6.0	12.5	20.5	27.0	0
5	BRAUN	0	3.5	11.0	20.5	28.0	0
6	GALLAGH	0	2.3	7.5	14.0	20.5	0
7	JOHNSON	1	3.0	10.0	17.0	28.0	0
8	JOHNSONR	0	2.3	7.5	13.5	22.3	0
9	KOCKELMN	0	3.5	8.5	15.5	22.0	0
10	MASRI	0	6.5	12.5	18.5	25.0	0
11	MTINGLY	5	10.0	17.5	23.0	27.5	0
12	SANDY	0	2.3	8.0	16.2	23.0	1
13	WALLACE	0	2.3	10.0	20.0	30.0	0
14	WHITMAN	0	0.5	2.5	6.0	8.0	0

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VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C. V.
MMV	12	0.46153846	1.28196891	0.00000000	5.00000000	0.34588993	6.00000000	1.89469714	901.468
MMVI	12	3.63076923	2.49649803	0.80000000	10.00000000	0.68289316	47.20000000	6.23280768	68.768
MMVII	12	8.69230769	3.50614566	2.90000000	17.50000000	0.87326187	126.00000000	12.31410296	26.205
MMVIII	12	15.63076923	4.32403835	6.00000000	23.00000000	1.19827248	216.20000000	18.69730769	26.000
MMIX	12	23.40769231	5.32828368	8.00000000	30.00000000	1.47810783	304.20000000	28.40243590	23.768

DATA FROM TASK DMG/SETTING THREE

OBS	ID	MRV	MRVI	MRVII	MRVIII	MRIX	C
1	BLANCK	14.5	17.00	22.5	25.5	28.0	0
2	ASAKURA	0.0	0.00	4.0	10.0	14.0	0
3	ARNOLD	0.0	0.80	3.0	8.0	13.4	1
4	BACHRACH	0.0	2.20	5.0	11.0	14.0	0
5	BRUN	0.0	1.50	5.0	14.5	20.0	0
6	GALLAGH	0.0	1.00	3.5	8.5	12.0	0
7	JOHNSON	0.0	2.00	7.0	15.0	23.0	0
8	JOHNSON	0.0	0.80	2.0	8.5	14.2	0
9	KOCKELM	0.0	1.50	8.0	12.0	16.0	0
10	MASHI	0.0	0.80	3.0	12.2	20.0	0
11	MITTINGLY	0.0	3.90	6.0	11.0	16.0	0
12	SANDY	0.0	1.00	4.5	12.0	18.5	1
13	WALLACE	0.0	0.80	5.0	15.0	20.0	0
14	WHITMAN	0.0	0.50	1.5	4.5	6.0	0

DATA FROM TASK ONE/SETTING THREE

6

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MNV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MNVI	13	1.26846154	0.88934288	0.0000000	3.5000000	0.24665834	16.4900000	0.78083077	70.112
MNVII	13	4.34615385	1.53276188	1.5000000	7.0000000	0.42511166	56.5000000	2.34935897	35.267
MNVIII	13	11.08230769	2.93441122	4.5000000	15.0000000	0.81385924	144.2000000	8.61078923	26.454
MNIX	13	16.00769231	3.87566371	9.0000000	23.0000000	1.07491571	208.1000000	15.02076923	24.211

DATA FROM TASK ONE/SETTING FOUR

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	21	25.50	27.0	30.0	34.0	0
2	ASAKURA	0	1.75	5.0	12.0	17.0	0
3	ARNOLD	0	0.00	4.0	10.0	14.0	0
4	BACHRACH	0	2.00	5.5	11.5	16.0	0
5	BRAUN	0	2.00	5.0	12.5	19.0	0
6	GALLAGH	0	1.50	4.0	10.0	14.2	0
7	JOHNSON	0	2.50	6.0	12.5	17.5	0
8	JOHNSONR	0	1.50	4.0	10.0	18.5	0
9	KOCKEL MW	0	1.00	4.5	10.0	16.0	0
10	MASRI	0	2.00	6.0	12.0	16.0	0
11	MITTINGLY	0	3.50	6.0	10.5	14.0	0
12	SANDY	0	1.50	4.0	10.0	15.0	1
13	WALLAGE	0	1.50	10.0	20.0	26.5	0
14	WHITMAN	0	0.50	1.5	3.5	8.0	0

DATA FROM TASK ONE/SETTING FOUR

8

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
MMVI	13	1.63481538	0.86971848	0.0000000	3.5000000	0.24121851	21.2500000	0.75641028	53.208
MMVII	13	5.03846154	1.93069006	1.5000000	10.0000000	0.53547708	65.5000000	3.72796410	38.318
MMVIII	13	11.11538462	3.52463855	3.5000000	20.0000000	0.97755885	144.5000000	12.42307692	31.710
MMIX	13	16.28461538	3.75207635	9.0000000	25.5000000	1.04063874	211.7000000	14.07807692	23.041

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DATA FROM TASK ONE/SETTING FIVE

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	10	12.00	16.5	20.5	25.0	0
2	ARNOLD	0	2.50	7.0	12.5	18.0	0
3	ASAKURA	0	2.75	8.0	15.5	22.5	0
4	BACHRACH	0	3.25	8.5	15.0	20.0	0
5	BRAUN	0	3.50	10.0	18.0	25.0	0
6	GALLAGH	0	2.50	7.0	13.0	18.0	0
7	JOHNSON	0	3.50	10.0	20.0	27.0	0
8	JOHNSONR	0	2.50	7.0	12.5	20.0	0
9	KOCKELMN	0	2.50	7.5	14.0	20.0	0
10	MASRI	0	6.00	11.5	17.0	22.0	0
11	MTINGLY	0	5.00	9.0	14.5	20.0	0
12	SANDY	0	2.50	7.0	14.0	20.0	1
13	WALLACE	0	2.50	10.0	20.0	25.0	0
14	WHITMAN	0	0.50	2.5	4.5	7.0	0

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DATA FROM TASK ONE/SETTING FIVE

10

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	.
MMVI	13	3.02307692	1.34189180	0.5000000	6.0000000	0.37217377	39.30000000	1.80087308	44.388
MMVII	13	8.07692308	2.23463414	2.5000000	11.5000000	0.61977600	105.00000000	4.99358974	27.667
MMVIII	13	14.67692308	3.97369556	4.5000000	20.0000000	1.10210485	190.80000000	15.79025641	27.074
MMIX	13	21.11538462	6.25883990	7.0000000	35.0000000	1.73588986	274.50000000	39.17307692	29.641

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DATA FROM TASK ONE/SETTING SIX

OBS	ID	MMIV	MMVI	MMVII	MMVIII	MMIX	C
1	BLANCK	26.5	28.0	32.0	35.0	35.0	0
2	ARNOLD	0.0	1.0	2.8	8.7	14.0	0
3	ASAKURA	0.0	1.0	4.5	10.5	15.0	0
4	BACHRACH	0.0	1.5	5.0	12.0	16.5	0
5	BRAUN	0.0	2.0	6.5	14.0	20.0	0
6	GALLAGH	0.0	1.0	4.0	10.0	13.0	0
7	JOHNSON	0.0	2.5	10.0	20.0	25.0	0
8	JOHNSONR	0.0	1.0	4.0	9.5	17.5	0
9	KOCKELMN	0.0	1.0	4.0	10.0	15.0	0
10	MASRI	0.0	4.5	9.0	15.0	20.0	0
11	MITTINGLY	2.0	6.5	10.5	16.0	20.0	0
12	SANDY	0.0	1.0	4.5	12.0	20.0	1
13	WALLACE	0.0	1.0	7.5	18.0	20.0	0

DATA FROM TASK ONE/SETTING SIX

12

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
NNV	12	0.1666667	0.57735027	0.0000000	2.0000000	0.1666667	2.0000000	0.3333333	348.410
NNVI	12	2.0000000	1.75809818	1.0000000	6.9000000	0.60751822	24.0000000	3.08080808	87.808
NNVII	12	6.1166667	2.51353910	3.9000000	10.9000000	0.7259624	73.4000000	6.31787879	41.083
NNVIII	12	12.8083333	3.24666496	9.5000000	20.0000000	0.93723144	153.7000000	10.54083331	25.348
NNIX	12	18.0000000	3.44436298	13.0000000	25.0000000	0.99430185	216.0000000	11.86363636	19.135

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DATA FROM TASK TWO/SETTING ONE

13

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.30	0.8	1.2
2	ASAKURA	0.0	1.2	8.00	18.0	21.0
3	BACHRACH	0.0	2.8	8.00	11.0	18.0
4	BLANCK	8.0	7.0	11.00	14.0	16.0
5	BRAUN	0.0	0.0	2.50	18.0	13.0
6	GALLAGH					
7	JOHNSON	0.0	0.8	9.50	9.0	12.8
8	JOHNSON	0.0	0.8	0.50	1.5	8.0
9	KOCKELM	0.0	1.0	4.50	12.8	21.0
10	MASRI	0.0	3.0	6.50	10.0	15.0
11	MTINGLY	2.8	8.0	8.50	14.0	17.8
12	SANDY	0.0	0.0	0.25	3.5	6.0
13	WALLACE	0.0	0.8	2.80	8.0	10.0
14	WHITMAN	0.0	0.8	2.50	8.0	18.8

591

DATA FROM TASK TWO/SETTING ONE

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C. V.
MMV	12	0.57692308	1.49786172	0.00000000	5.00000000	0.41943210	7.50000000	2.24358974	259.628
MMVI	12	1.74815385	2.31034463	0.00000000	7.00000000	0.54077431	22.70000000	5.33769231	132.310
MMVII	13	4.18615385	3.38196975	0.25000000	11.00000000	0.93788853	54.55000000	11.43769231	80.597
MMVIII	13	9.49230769	5.39451288	0.90000000	18.00000000	1.49616868	123.40000000	29.10076923	56.830
MMIX	13	12.82307692	5.96226381	1.20000000	21.00000000	1.65363445	166.70000000	35.54858974	46.496

DATA FROM TASK TWO/SETTING TWO

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.5	1.2	2.8
2	ASAKURA	0.0	6.0	15.0	27.5	40.0
3	BACHRACH	0.0	8.5	17.5	25.0	31.0
4	BLANK	10.0	13.5	18.0	23.0	29.0
5	BRAUN	0.0	0.0	3.0	23.0	45.0
6	GALLACH					
7	JOHNSONG	0.0	2.5	8.0	15.5	22.5
8	JOHNSONR	0.0	0.5	2.0	4.5	10.0
9	KOCKELMN	0.0	6.0	13.0	20.0	26.0
10	MASRI	0.0	4.5	10.0	16.0	22.0
11	MITTINGLY	7.5	13.0	17.5	23.5	29.0
12	SANDY	0.0	1.0	5.0	14.0	21.0
13	WALLACE	0.0	2.0	8.0	10.0	15.0
14	WHITMAN	0.0	2.0	8.5	16.0	25.0

DATA FROM TASK 1BU/SETTING 1BU

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V. %
MMV	13	1.34615385	3.32531066	0.00000000	10.00000000	0.92227528	17.50000000	11.05769231	247.023
MMVI	13	4.65384615	4.72310198	0.00000000	13.50000000	1.30995280	60.50000000	32.30769231	101.488
MMVII	13	9.50000000	6.36238949	0.30000000	19.00000000	1.76460935	125.80000000	40.48000000	66.275
MMVIII	13	16.86153846	7.88628869	1.20000000	27.50000000	2.21502846	219.20000000	63.78754410	47.365
MMIX	13	24.53846154	11.38248204	2.50000000	45.00000000	3.15693251	319.00000000	128.56089744	46.366

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DATA FROM TASK TWO/SETTING THREE

17

OBS	ID	NRV	NRVI	NRVII	NRVIII	NRXIX
1	ARNOLD	0.0	0.0	0.8	1.0	2.0
2	ASAKURA	0.0	2.8	7.8	14.0	22.8
3	BACHRACH	1.0	3.8	8.0	13.8	17.0
4	BLANCK	6.8	8.0	11.0	15.8	19.0
5	BRAUN	0.0	0.0	3.0	18.0	35.0
6	GALLAGH					
7	JOHNSON	0.0	1.8	5.8	9.0	21.8
8	JOHNSON	0.0	0.8	1.0	2.8	8.0
9	KOCKELMN	0.0	3.8	8.0	18.0	21.0
10	MASRI	0.0	0.8	3.0	9.8	17.0
11	MTINGLY	1.0	3.8	6.0	11.8	16.0
12	SANDY	0.0	0.0	0.0	2.0	4.8
13	WALLACE	0.0	0.8	8.0	8.0	11.0
14	WHITMAN	0.0	1.0	4.0	14.0	28.8

DATA FROM TASK TWO/SETTING THREE

18

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	0.65384615	1.78565001	0.00000000	8.80000000	0.48802371	8.80000000	9.22435887	274.629
MMVI	13	1.84615385	2.27582051	0.00000000	8.00000000	0.63118804	28.30000000	8.17835887	116.939
MMVII	13	4.73076923	3.43763112	0.00000000	11.00000000	0.95342733	61.50000000	11.81730769	72.665
MMVIII	13	10.34615385	5.68398560	1.00000000	18.00000000	1.57645398	134.50000000	32.30769231	54.938
MMIX	13	17.23076923	9.35105917	2.00000000	35.00000000	2.59351718	224.00000000	87.44230769	54.270

DATA FROM TASK TWO/SETTING FOUR

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.0	0.8	1.0	2.0
2	ASAKURA	0.0	2.8	6.8	18.0	30.0
3	BACHRACH	0.0	8.8	8.0	14.8	20.0
4	BLANCK	16.0	18.8	23.0	26.0	30.0
5	BRAUN	0.0	0.0	2.8	13.0	40.0
6	GALLAGH					
7	JOHNSONS	0.0	2.0	4.8	10.0	18.0
8	JOHNSON	0.0	0.8	1.0	3.8	10.0
9	KOCKELM	0.0	2.8	6.8	12.8	18.0
10	MASRI	0.0	3.0	7.8	13.8	19.0
11	MITTINGLY	2.8	5.8	10.0	14.0	20.0
12	SANDY	0.0	1.0	1.8	8.0	8.0
13	WALLACE	0.0	1.8	8.0	18.0	20.0
14	WHITMAN	0.0	1.2	4.8	10.2	18.8

DATA FROM TASK TWO/SETTING FOUR

20

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	13	1.42307482	4.43398877	0.00000000	18.00000000	1.22976722	18.50000000	19.66025641	311.878
MMVI	13	3.28461538	5.10688988	0.00000000	18.00000000	1.41833938	42.70000000	26.07607692	188.472
MMVII	13	6.53846154	5.90360603	0.50000000	23.00000000	1.63736571	85.00000000	34.85256410	80.290
MMVIII	13	12.08230769	6.56460986	1.00000000	26.00000000	1.82069519	157.20000000	43.08410256	54.287
MMIX	13	18.96153846	10.21938631	2.00000000	40.00000000	2.83434835	246.50000000	104.43589744	53.695

DATA FROM TASK TWO/SETTING FIVE

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0	0.0	0.9	0.8	1.0
2	ASAKURA	0	4.0	14.0	24.0	34.0
3	BACHRACH	0	8.0	11.0	18.8	23.0
4	BLANCK	10	13.0	16.5	20.5	25.0
5	BRAUN	0	0.0	2.5	18.5	35.0
6	GALLAGH	0	8.0	8.0	17.0	25.0
7	JOHNSON	0	0.5	1.5	8.0	12.5
8	JOHNSON	0	4.5	8.0	16.0	23.5
9	KOCKELMAN	0	4.5	9.5	14.5	19.0
10	MASRI	0	4.7	9.0	14.2	20.0
11	MITTINGLY	3	0.5	2.0	4.5	8.0
12	SANDY	0	2.8	5.0	10.0	15.0
13	WALLACE	0	2.0	7.0	19.5	30.0
14	WHITMAN	0	2.0	7.0	19.5	30.0

DATA FROM TMSA (BU) BEILING FIVE

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C. V.
MMV	13	1.0000000	2.8262712	0.0000000	10.0000000	0.7844644	13.0000000	8.0000000	282.843
MMVI	13	3.4000000	3.4406792	0.0000000	13.0000000	0.9542751	44.2000000	11.8383333	101.187
MMVII	13	7.4076923	4.9642826	0.3000000	16.5000000	1.3768479	96.3000000	24.8441026	67.015
MMVIII	13	13.6153846	6.8058342	0.8000000	24.0000000	1.8876401	177.0000000	48.3214102	49.987
MMIX	13	19.9230769	9.7100595	1.0000000	35.0000000	2.6930897	259.0000000	94.2852561	48.738

DATA FROM TASK TWO/SETTING SIX

23

OBS	ID	MMV	MMVI	MMVII	MMVIII	MMIX
1	ARNOLD	0.0	0.00	0.8	1.0	1.9
2	ASAKURA	0.0	2.00	7.0	17.8	24.0
3	BACHRACH	0.0	2.00	6.5	17.5	25.0
4	BLANCK	5.0	10.00	15.0	19.0	23.0
5	BRAUN	0.0	0.00	2.5	19.0	35.0
6	GALLAGH					
7	JOHNSONG	0.0	2.00	8.0	17.5	22.5
8	JOHNSONR	0.0	0.25	0.5	2.5	10.0
9	MASRI	0.0	2.50	7.5	12.5	17.0
10	MTINGLY	1.5	6.50	11.0	16.0	20.0
11	SANDY	0.0	0.00	3.0	5.0	6.0
12	MALLACE	0.0	1.00	5.0	10.0	15.0
13	WHITMAN	0.0	1.00	4.5	12.0	22.0

VARIABLE	N	MEAN	STANDARD DEVIATION	MINIMUM VALUE	MAXIMUM VALUE	STD ERROR OF MEAN	SUM	VARIANCE	C.V.
MMV	12	0.5416667	1.4687399	0.0000000	5.0000000	0.4238870	6.5000000	2.1571869	271.152
MMVI	12	2.2708333	3.0253818	0.0000000	10.0000000	0.8733525	27.2500000	9.1529356	133.228
MMVII	12	6.0000000	4.3011626	0.5000000	15.0000000	1.2416387	72.0000000	18.5000000	71.686
MMVIII	12	12.5416667	6.3798629	1.0000000	19.0000000	1.8417078	150.5000000	40.7026515	50.869
MMIX	12	18.4166667	8.1820870	1.5000000	35.0000000	2.6506388	221.0000000	84.3106068	49.857

A P P E N D I X B

Damage curve shifts to reflect policy mitigation effects

FIGURE B.1: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
LOW-DENSITY RESIDENTIAL NEIGHBORHOOD

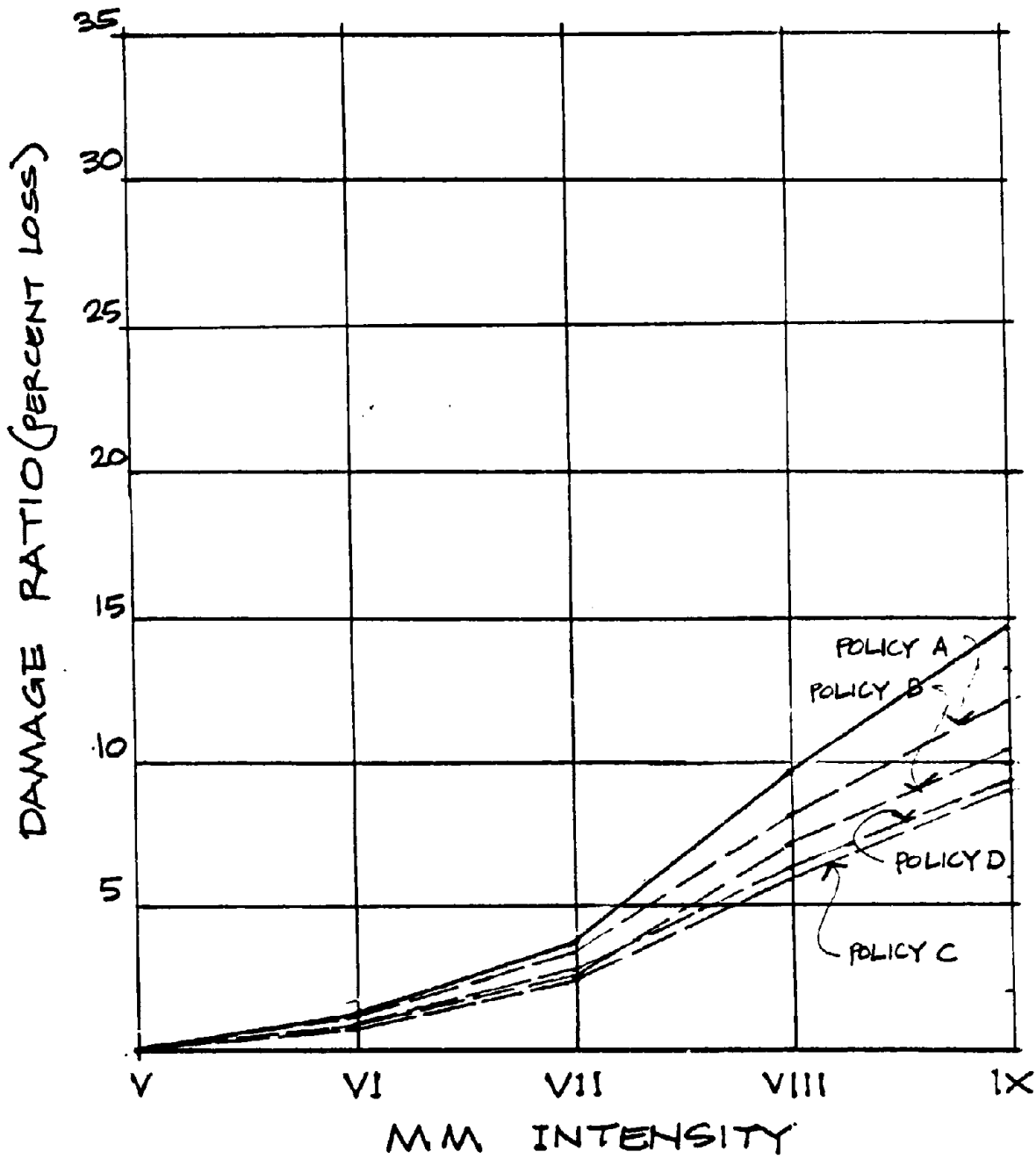


FIGURE B.2: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
MEDIUM-DENSITY, BUILT-UP MIXED-USE DISTRICT

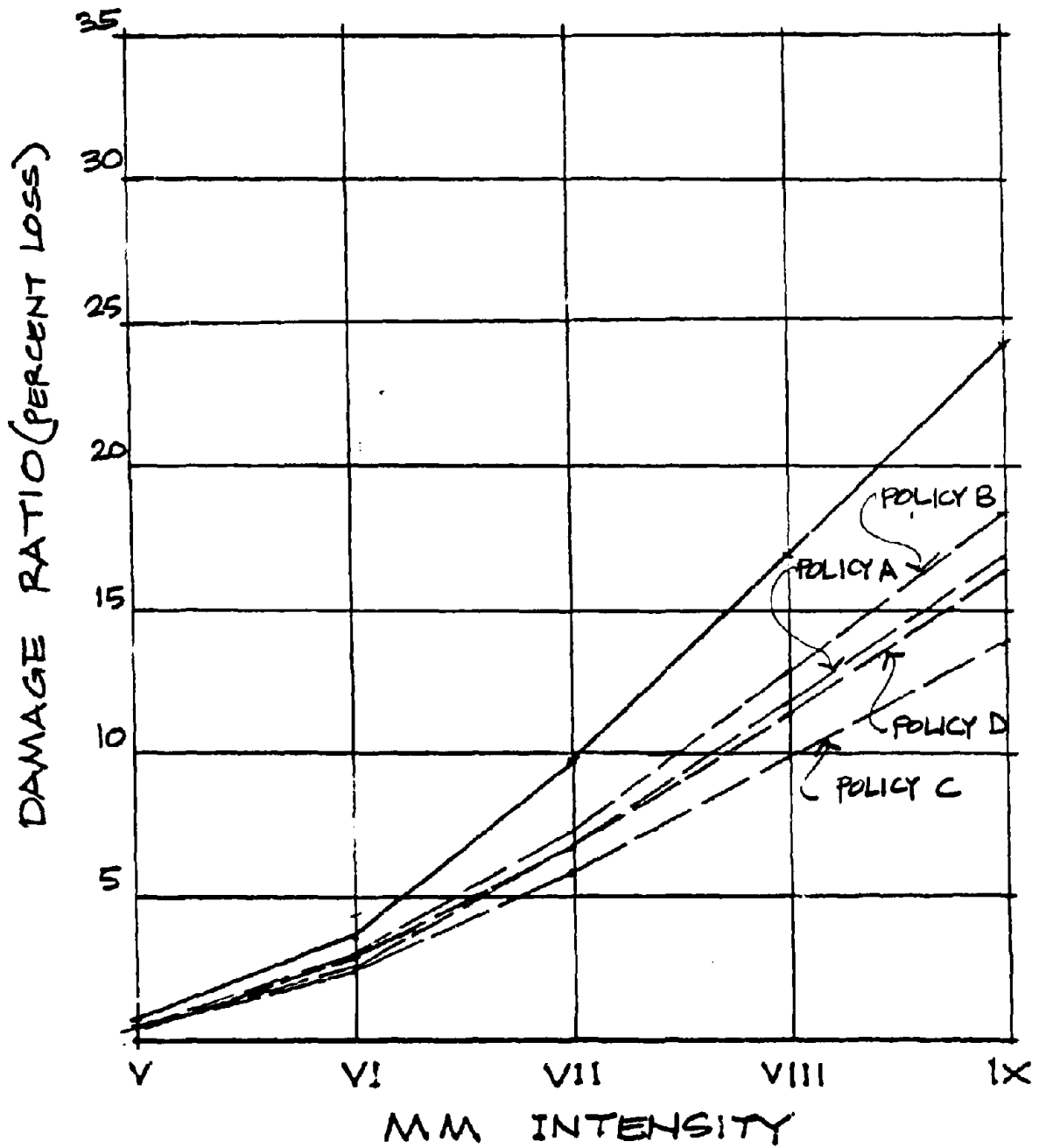


FIGURE B.3: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
OLDER HIGH-DENSITY TRANSITIONAL NEIGHBORHOOD

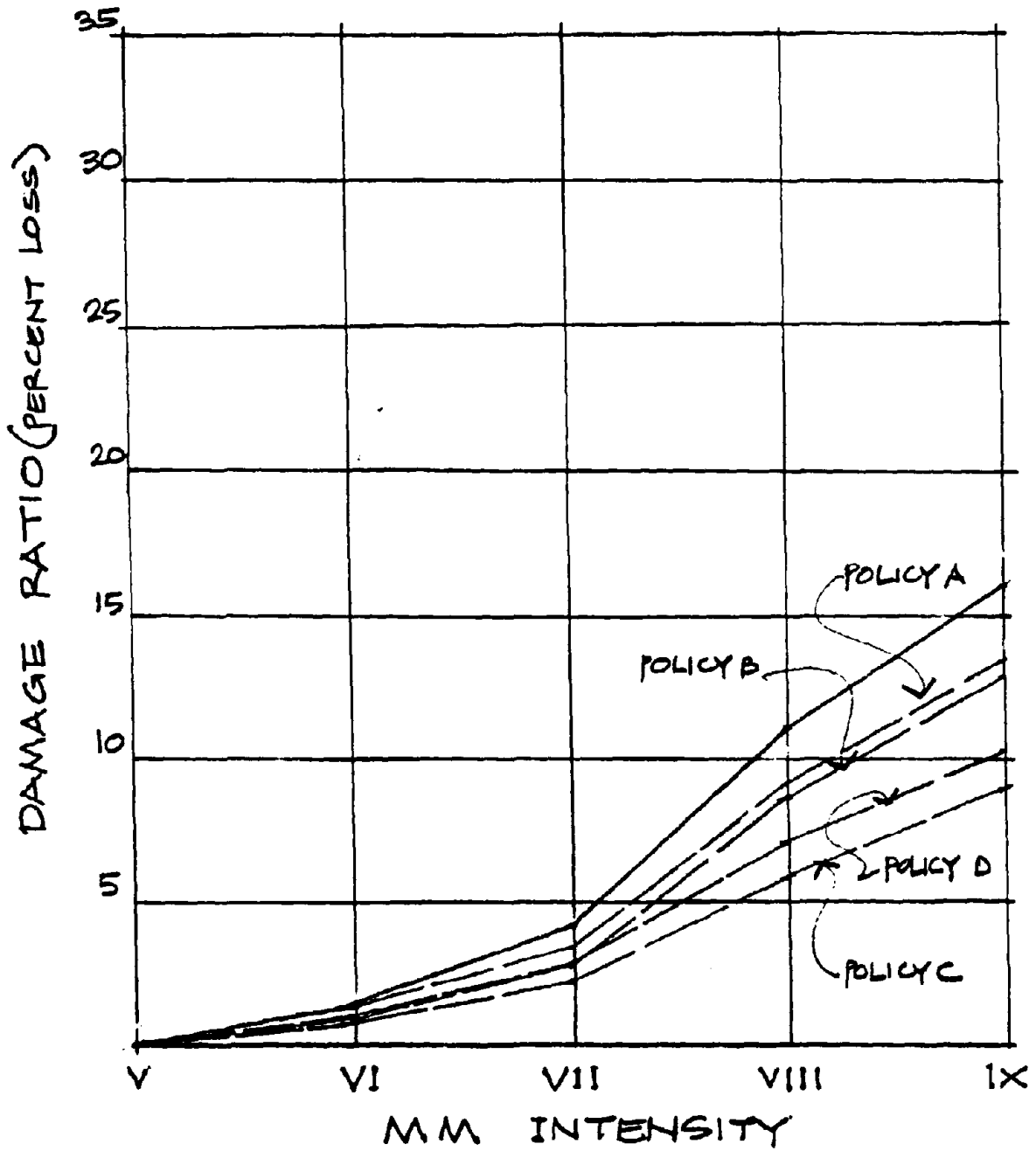


FIGURE B.4: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
HIGH-RISE HIGH-VALUE BUSINESS DISTRICT

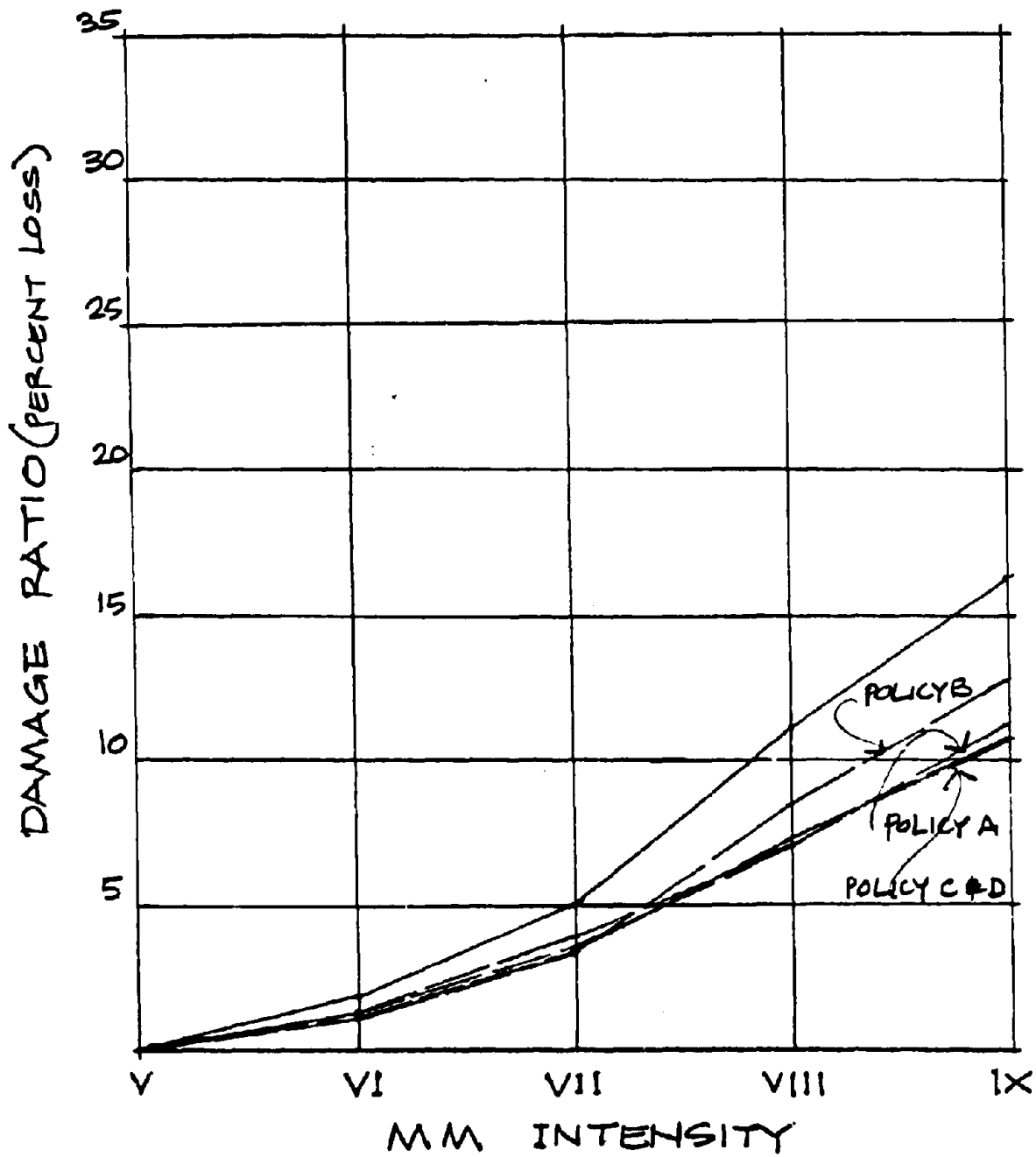


FIGURE B.5: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
MIXED INDUSTRIAL-COMMERCIAL DISTRICT

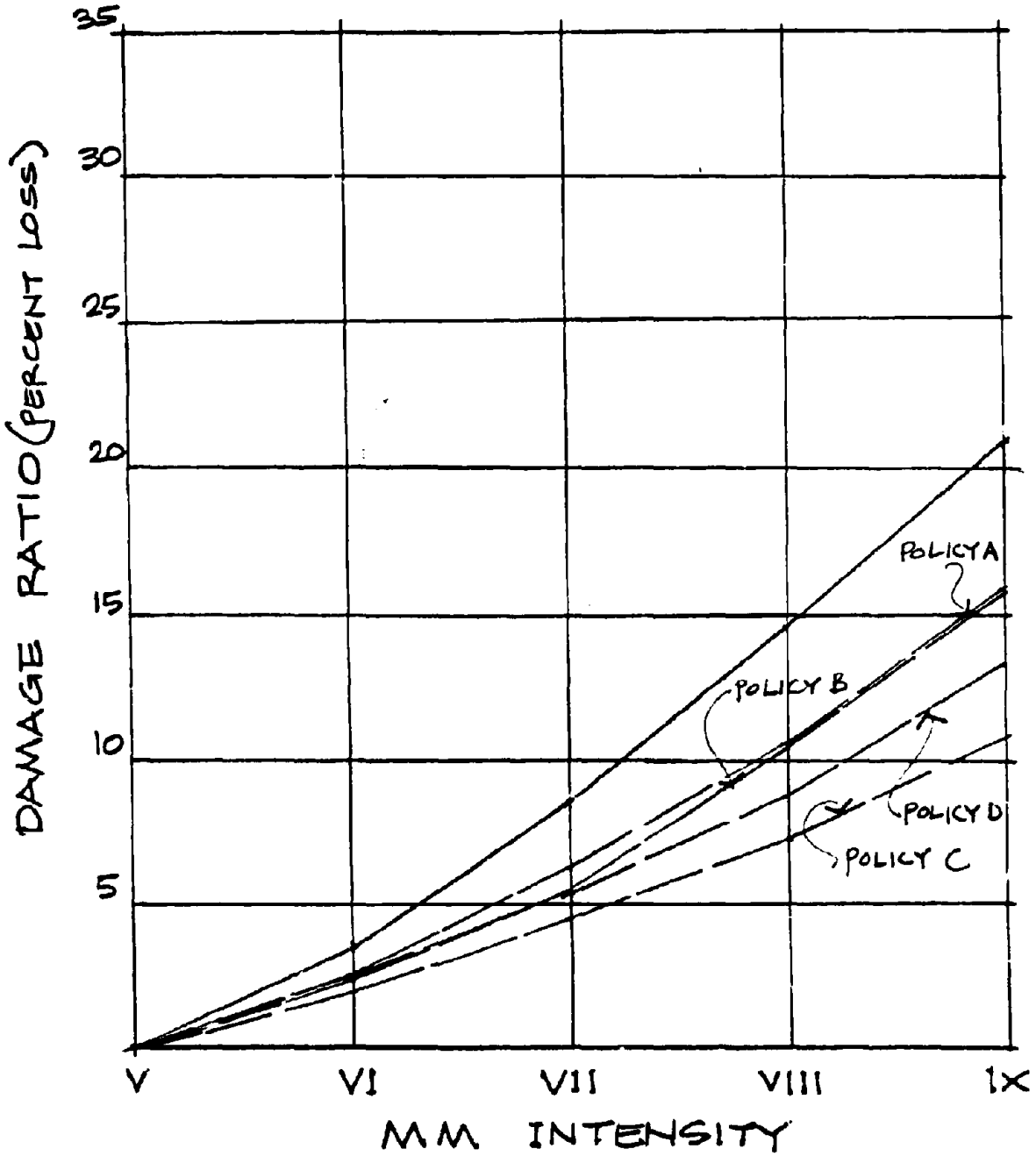


FIGURE B.6: POLICY EFFECTS ON BUILDING DAMAGE INCLUDING INTERACTION
OLDER MEDIUM-DENSITY MIXED NEIGHBORHOOD

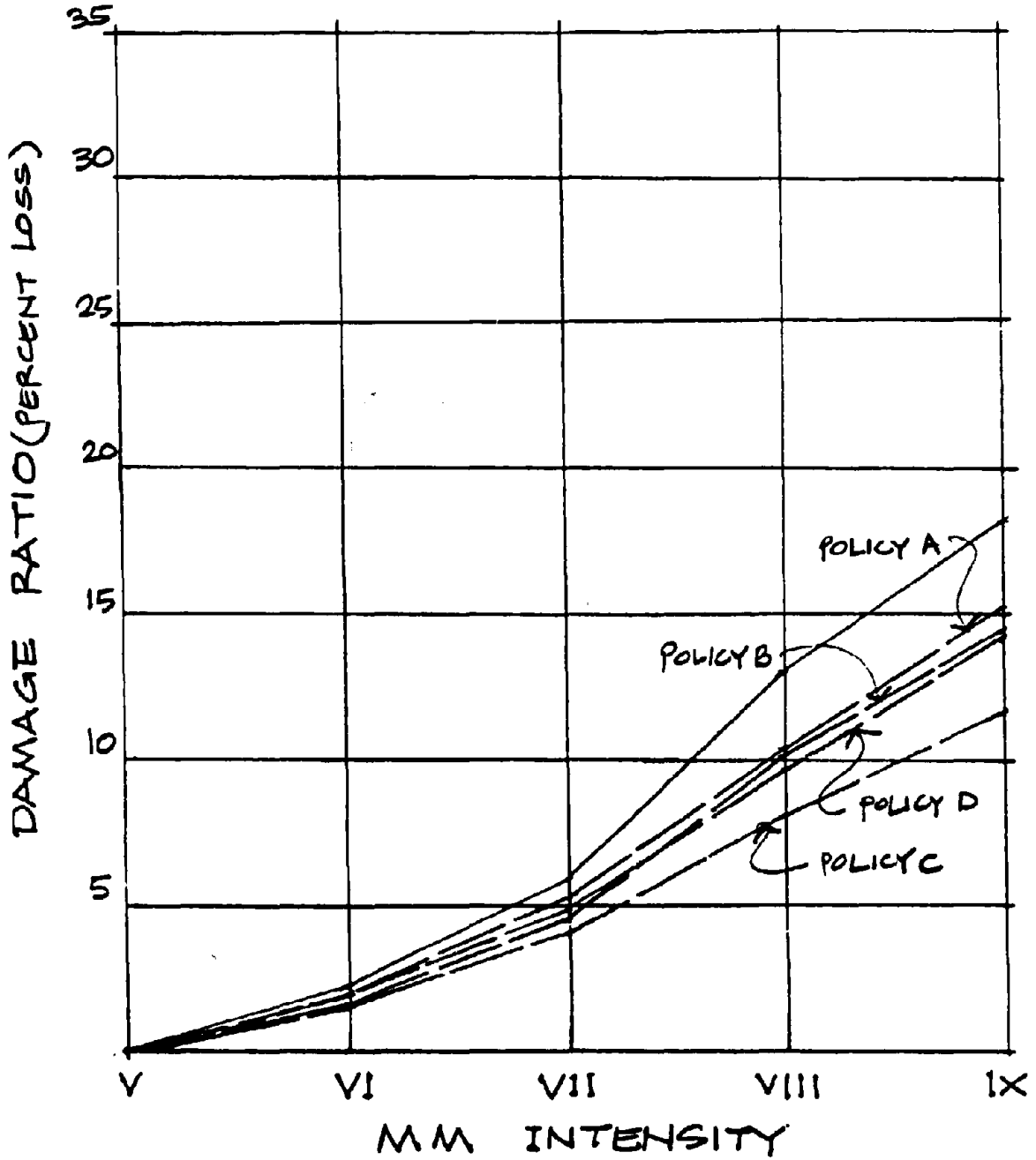


FIGURE B.7: POLICY EFFECTS ON NON-BUILDING STOCK DAMAGE
LOW-DENSITY RESIDENTIAL NEIGHBORHOOD

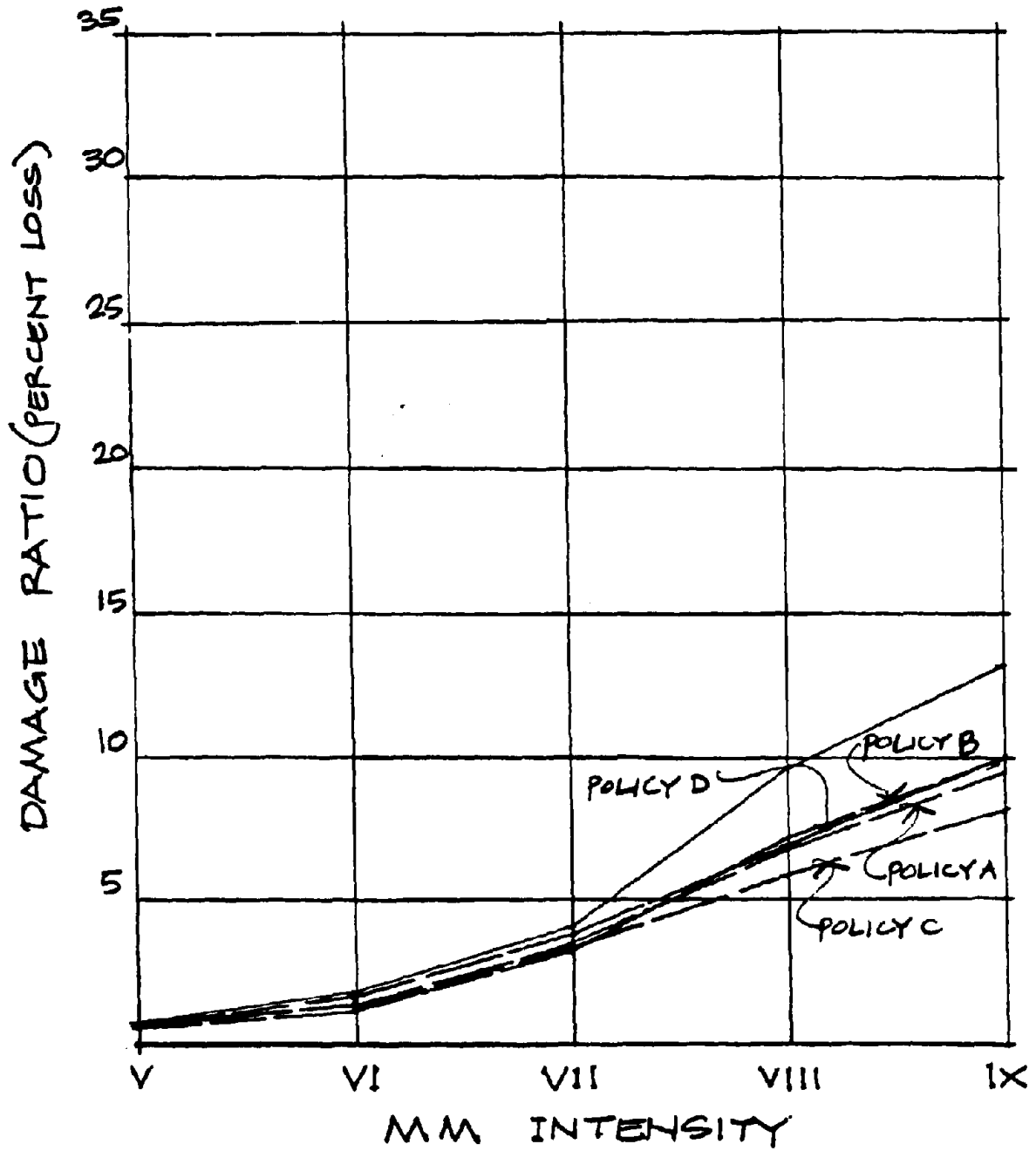


FIGURE B.8: POLICY EFFECTS ON NON-BUILDING STOCK DAMAGE
MEDIUM-DENSITY BUILT-UP MIXED-USE DISTRICT

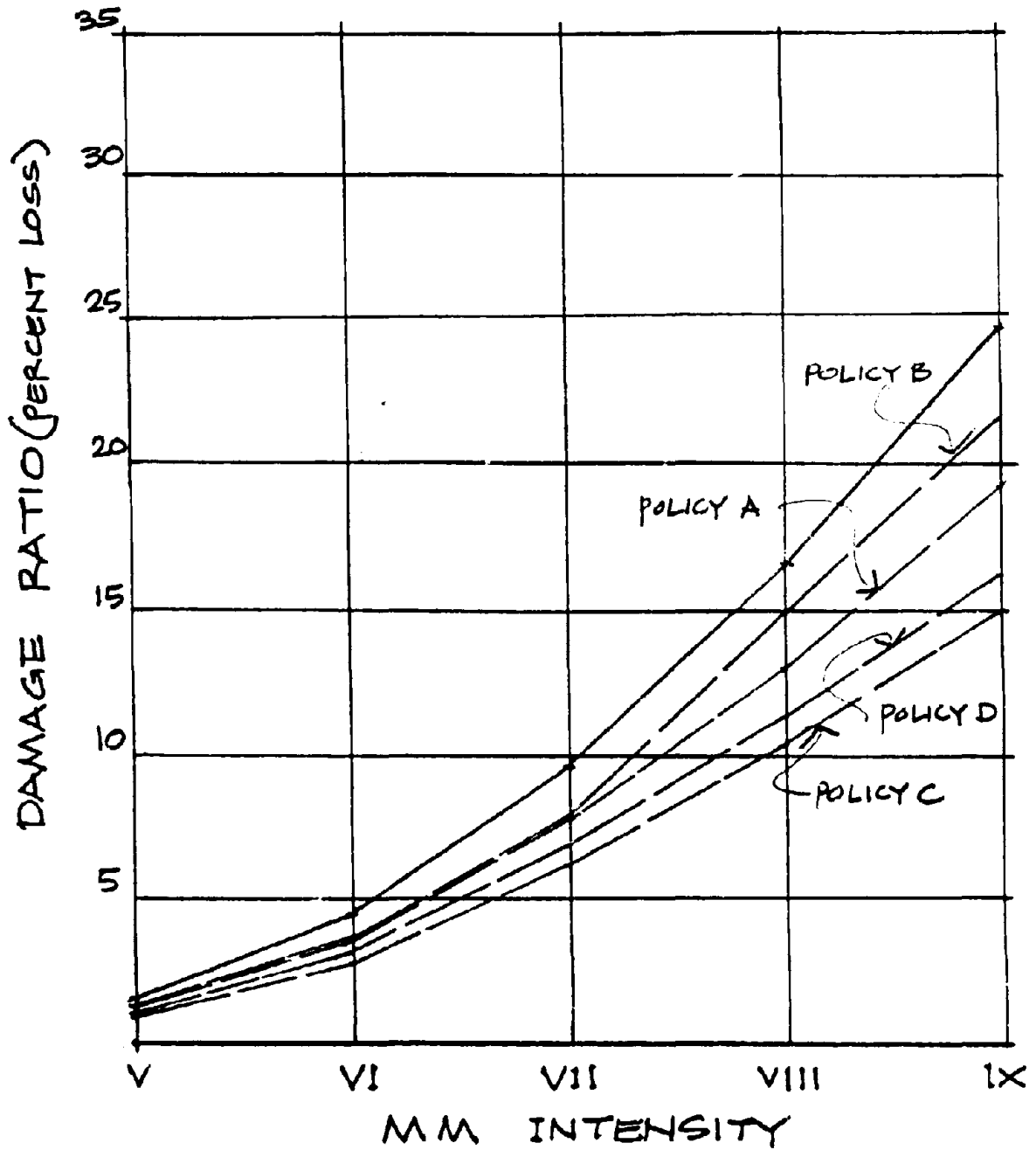


FIGURE B.9: POLICY EFFECTS ON NON-BUILDING STOCK DAMAGE
OLDER HIGH-DENSITY TRANSITIONAL NEIGHBORHOOD

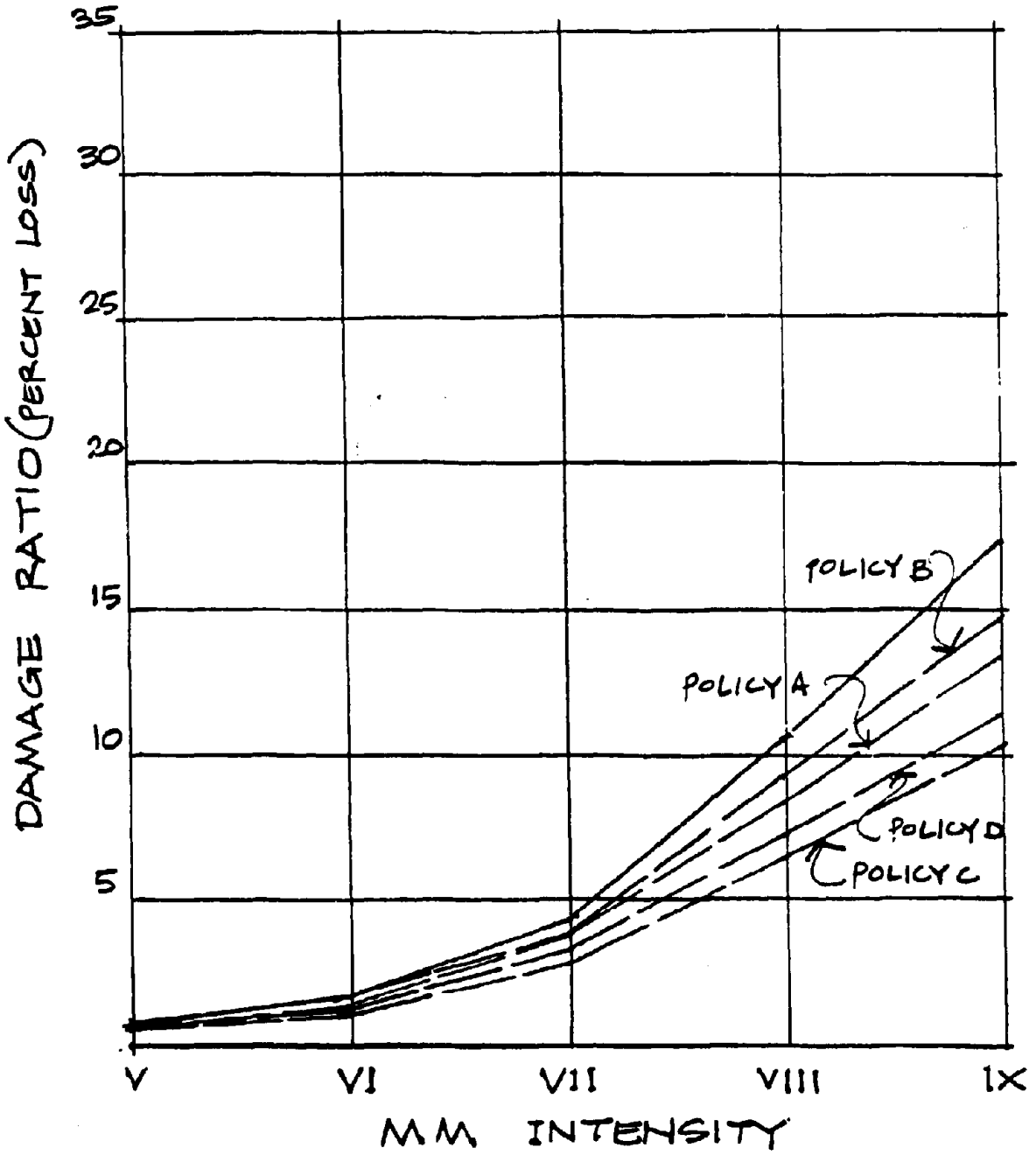


FIGURE B.10: POLICY EFFECTS ON NON-BUILDING STOCK DAMAGE
HIGH-RISE HIGH-VALUE BUSINESS DISTRICT

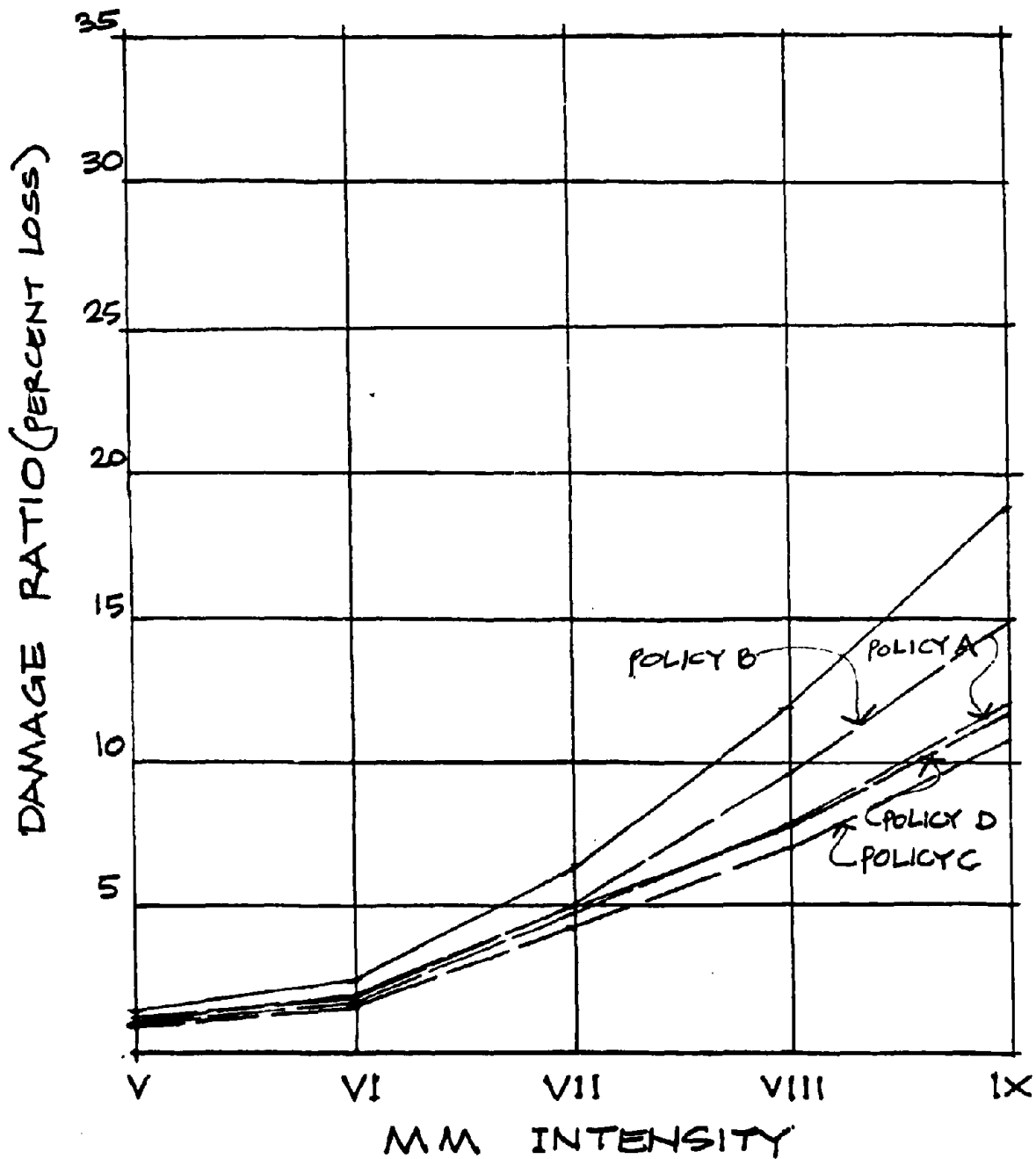


FIGURE B.11: POLICY EFFECTS ON NON-BUILDING STOCK DAMAGE
MIXED-USE INDUSTRIAL-COMMERCIAL DISTRICT

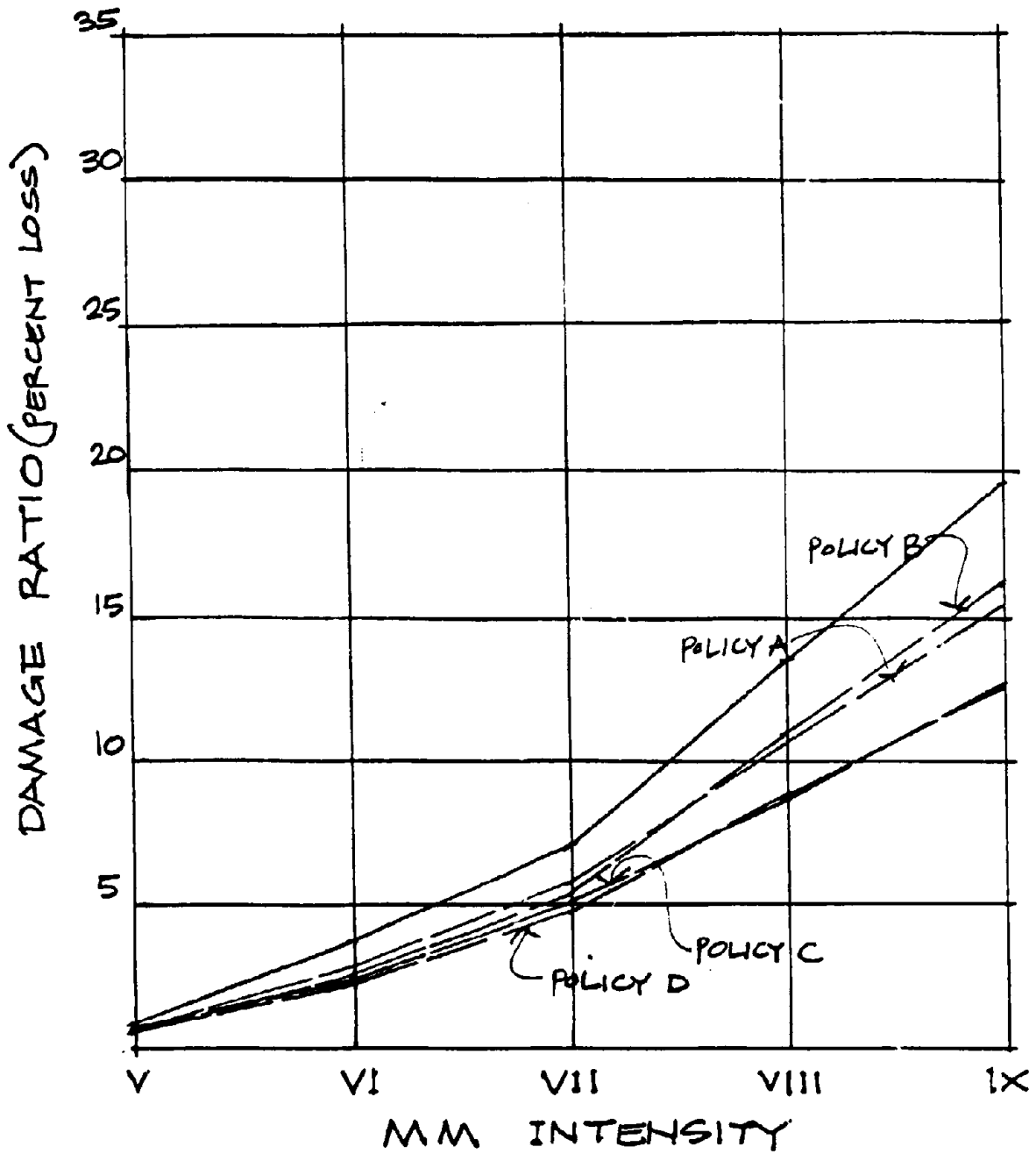
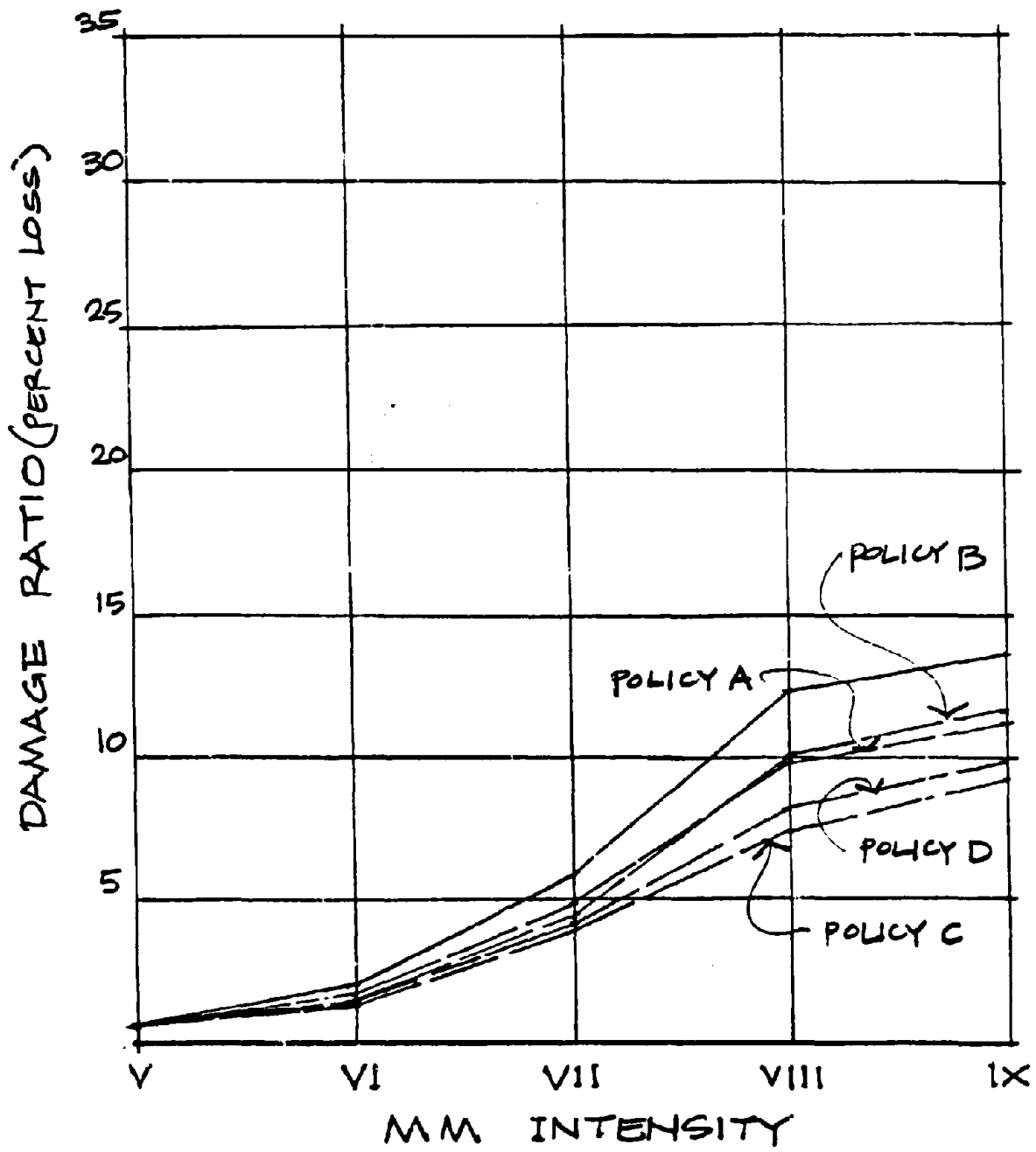


FIGURE B.12: POLICY EFFECTS OF NON-BUILDING STOCK DAMAGE
OLDER MEDIUM-DENSITY MIXED NEIGHBORHOOD



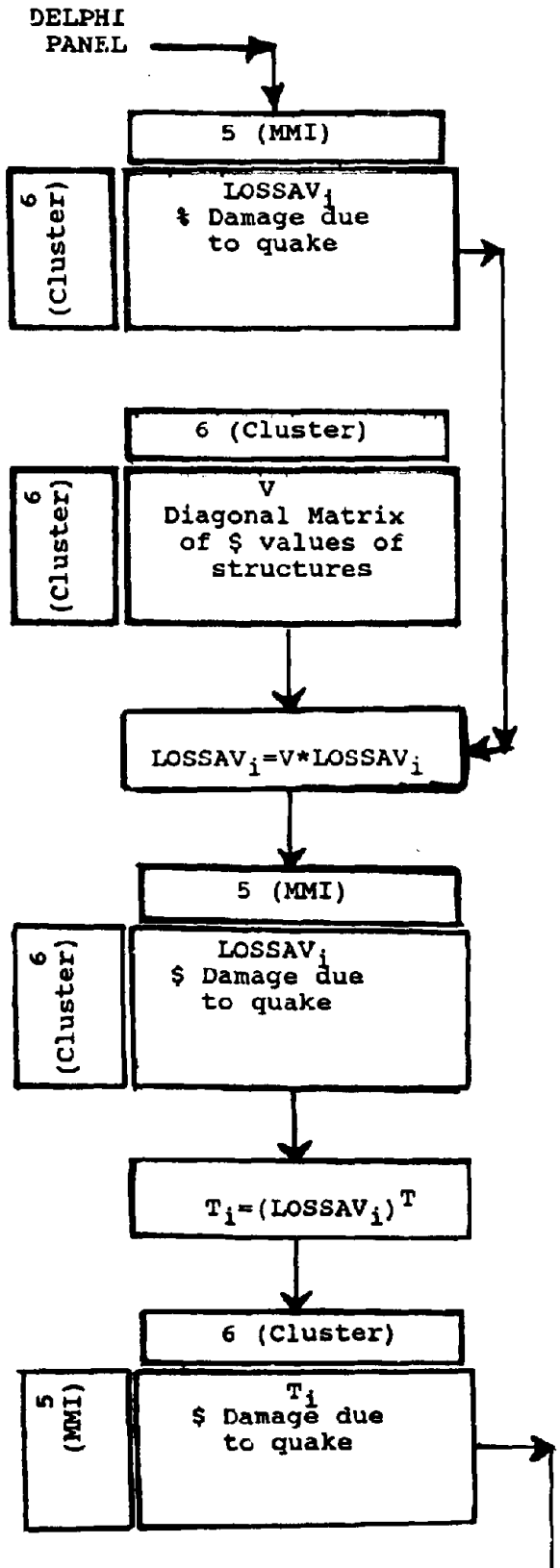
A P P E N D I X C

Flow-chart of the model

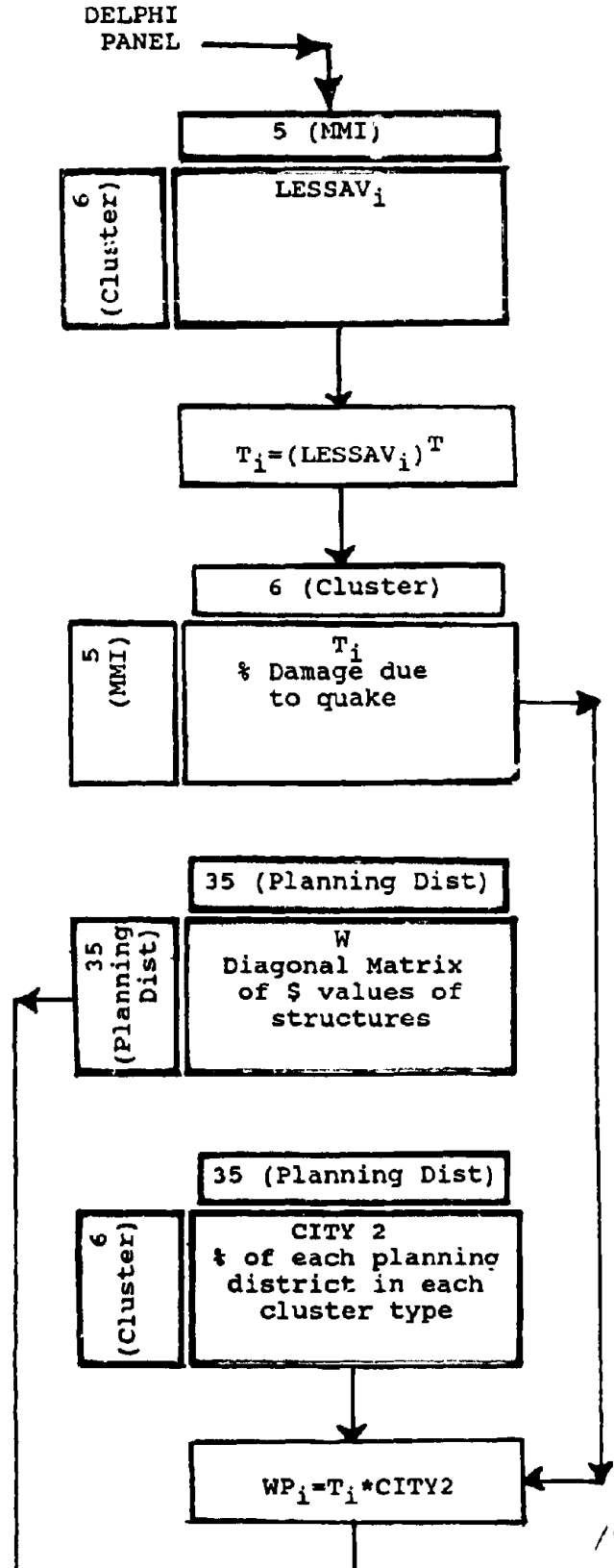
pink boxes -- inputs
small blue boxes -- computation
large blue boxes -- outputs
white boxes -- panel inputs

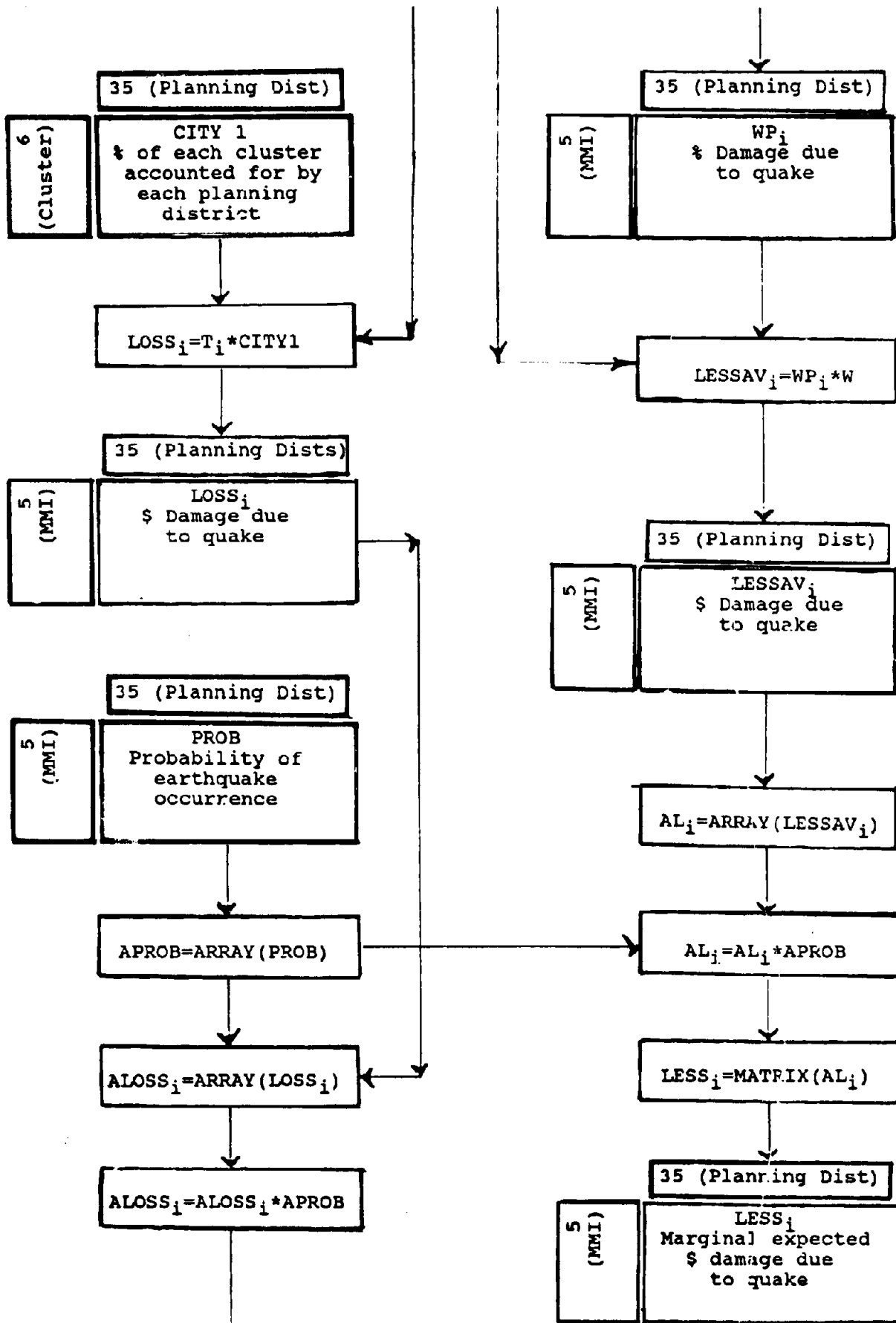
(i = 0,A,B,C,D)

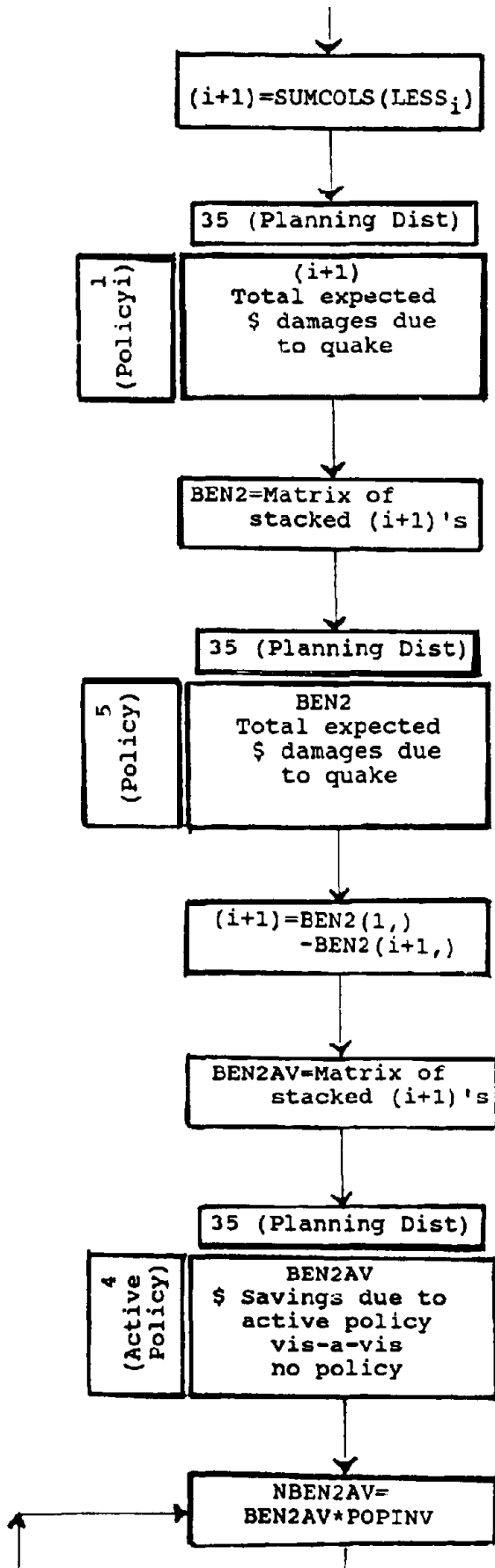
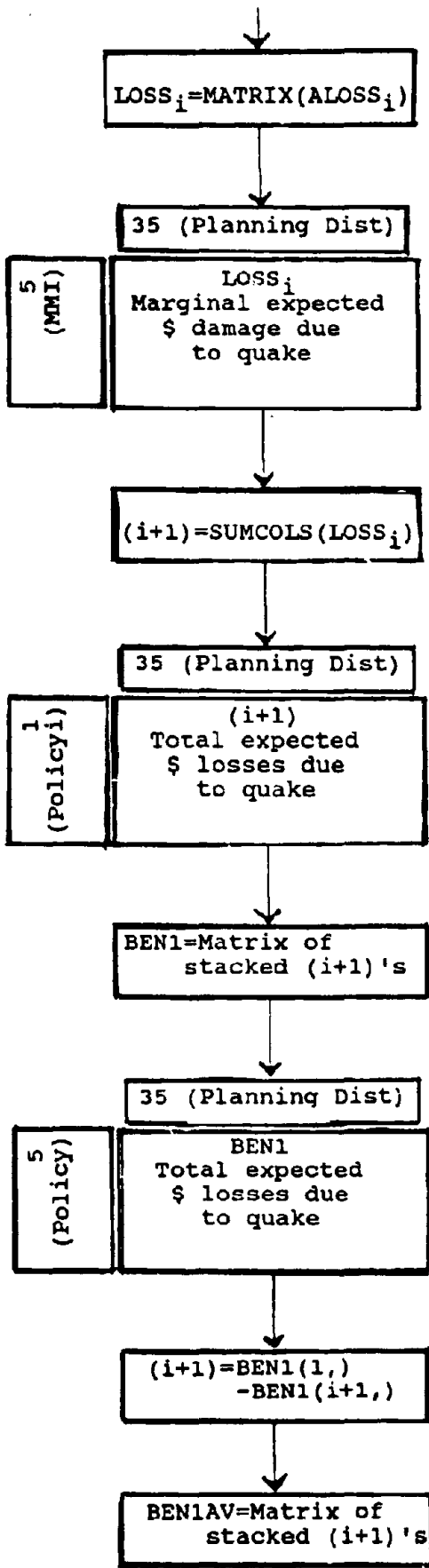
PRIVATE STRUCTURES

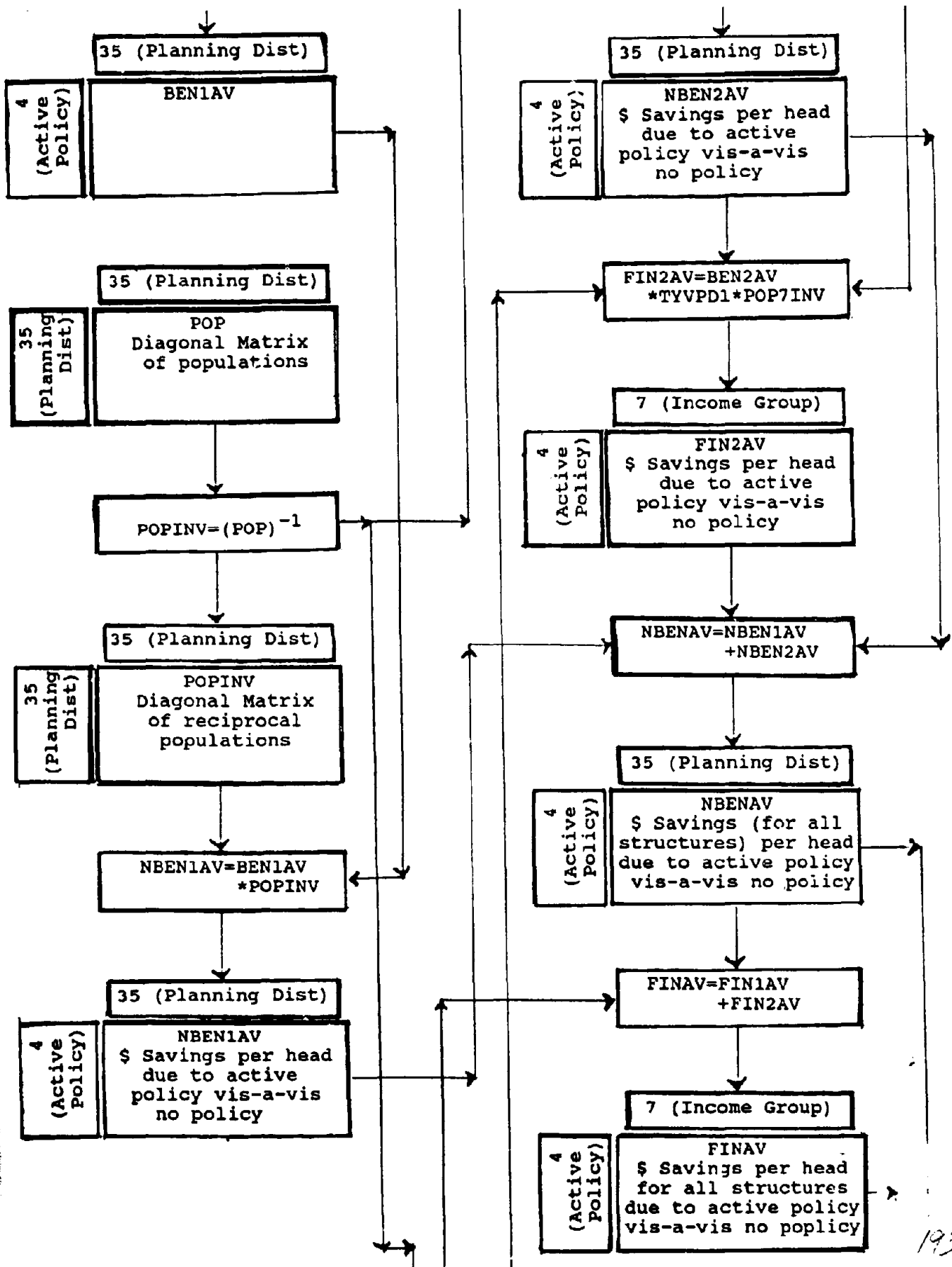


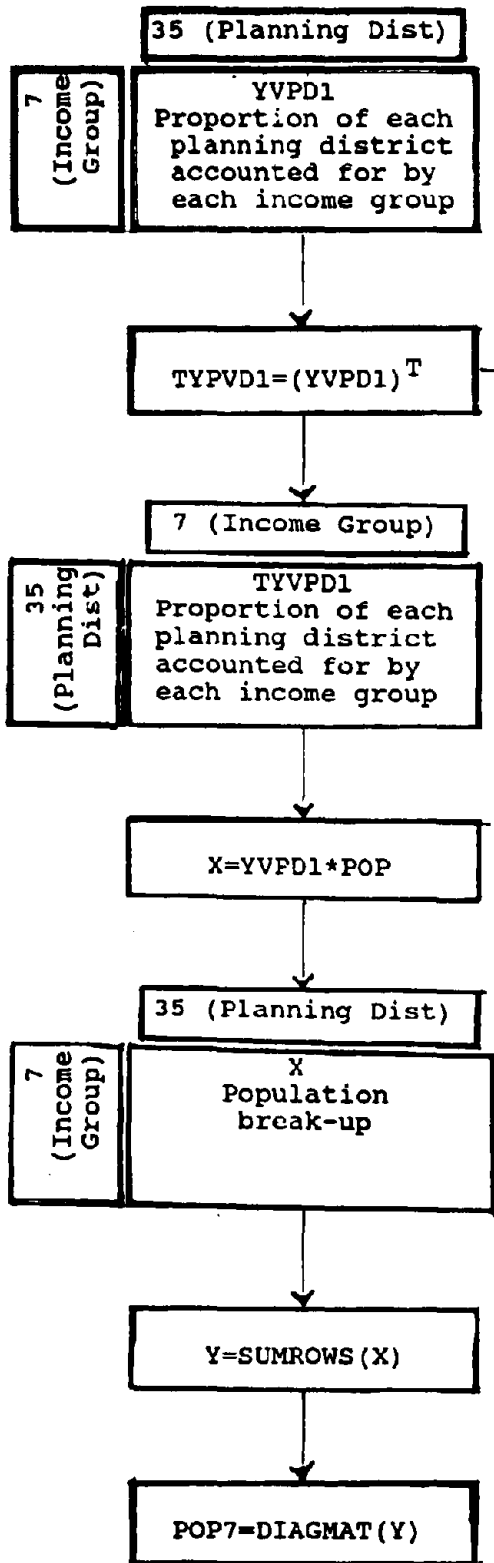
NON-PRIVATE STRUCTURES

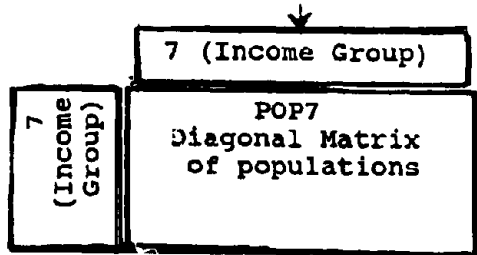




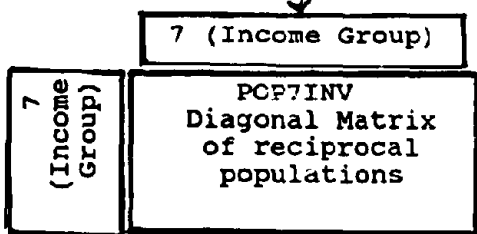




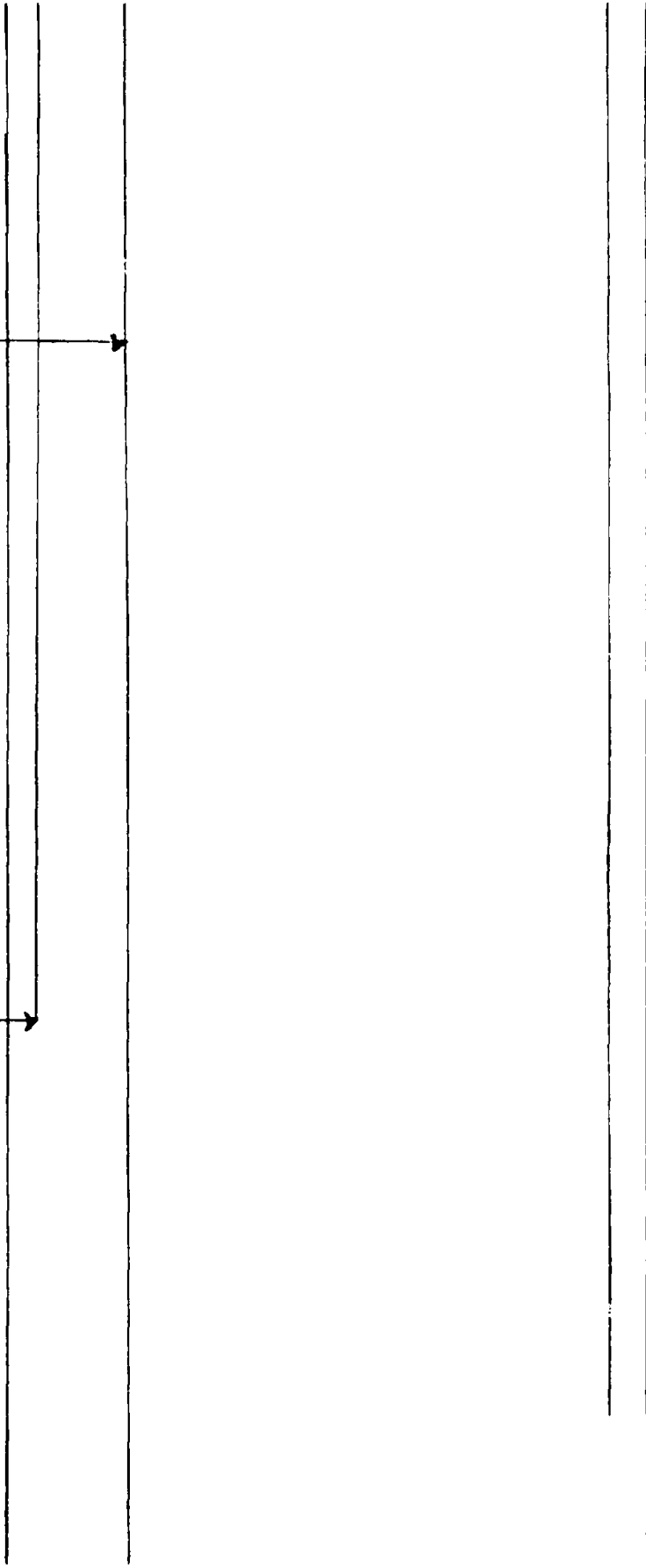
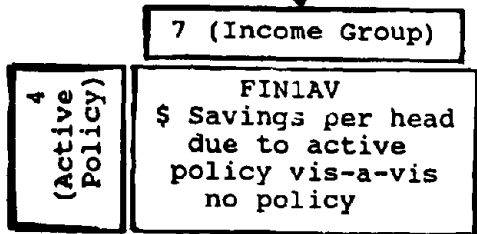


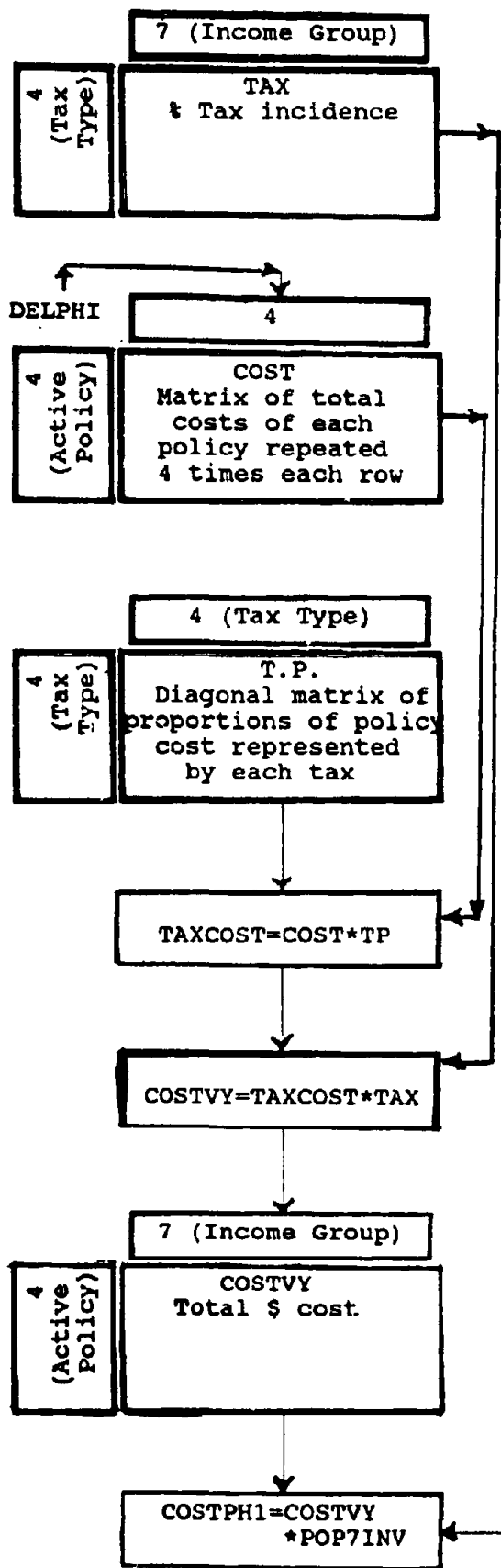


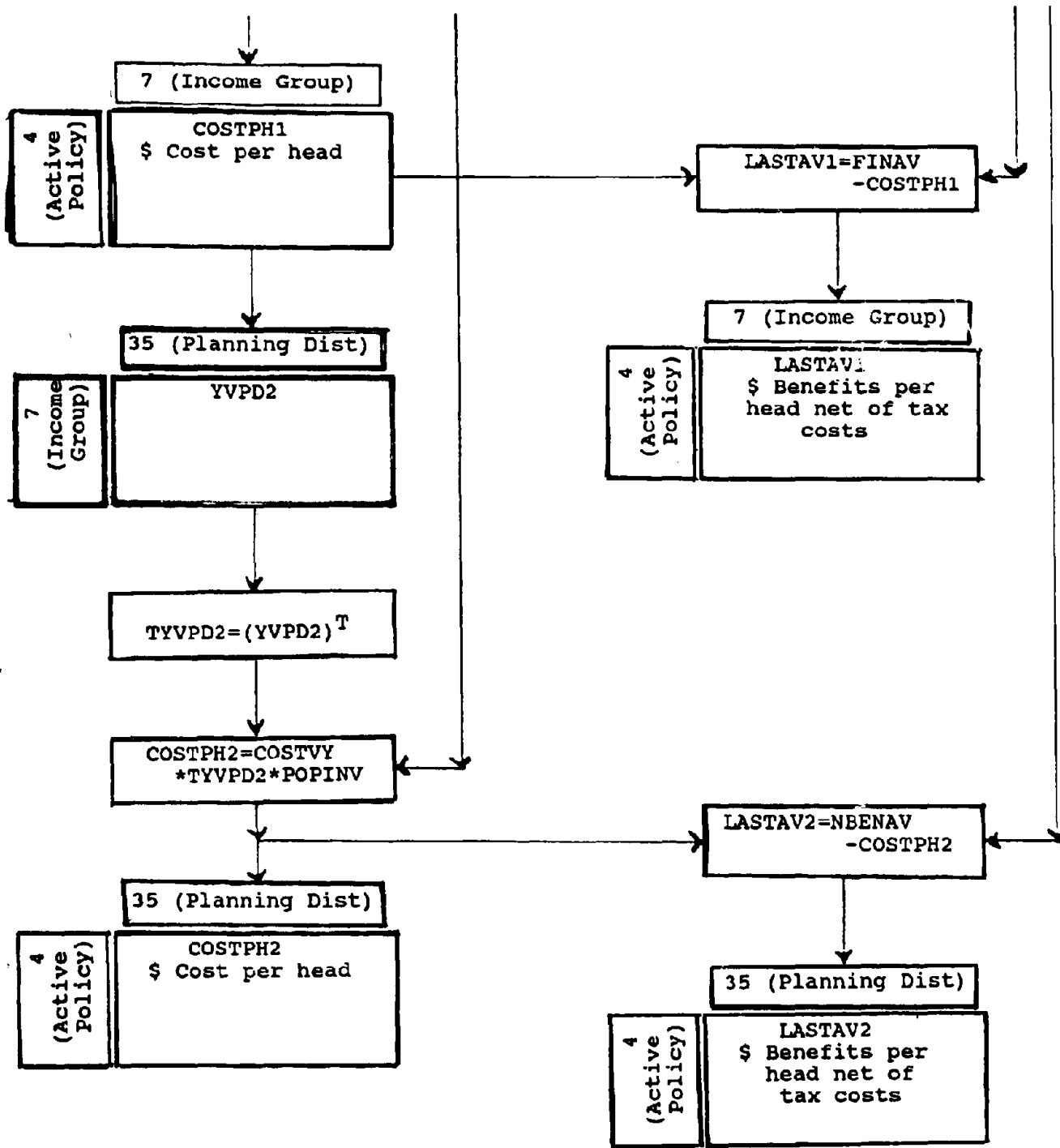
$POP7INV = (POP7)^{-1}$



$FIN1AV = BEN1AV * TYVPD1 * POP7INV$







A P P E N D I X D

Computer program (SPEAKEZ)

:_edit project2
EDIT COMMAND MODE

:X1

EDITING PROJECTS

```
1 PROGRAM
2 JOURNAL(OH)
3 LOSSAVO=KEPT(LOSSAVO); LOSSAVA=KEPT(LOSSAVA); LOSSAVB=KEPT(LOSSAVB)
4 LOSSAVC=KEPT(LOSSAVC); LOSSAVD=KEPT(LOSSAVD)
5 LOSSAVO; LOSSAVA; LOSSAVB; LOSSAVC; LOSSAVD
6 LOSSCONO=KEPT(LOSSCONO); LOSSCONA=KEPT(LOSSCONA); LOSSCONB=KEPT(LOSSCONB)
7 LOSSCONC=KEPT(LOSSCONC); LOSSCOND=KEPT(LOSSCOND)
8 LOSSCONO; LOSSCONA; LOSSCONB; LOSSCONC; LOSSCOND
9 V=KEPT(V)
10 V
11 LOSSAVO=V*LOSSAVO; LOSSAVA=V*LOSSAVA; LOSSAVB=V*LOSSAVB
12 LOSSAVC=V*LOSSAVC; LOSSAVD=V*LOSSAVD
13 TO=TRANPOSE(LOSSAVO); TA=TRANPOSE(LOSSAVA); TB=TRANPOSE(LOSSAVB)
14 TC=TRANPOSE(LOSSAVC); TD=TRANPOSE(LOSSAVD)
15 CITY1=KEPT(CITY1)
16 CITY1
17 LOSSO=TO*CITY1; LOSSA=TA*CITY1; LOSSB=TB*CITY1; LOSSC=TC*CITY1; LOSSD=TD*CITY1
18 PROB=KEPT(PROB)
19 PROB
20 APROB=ARRAY(175;PROB)
21 ALOSSO=ARRAY(175;LOSSO); ALOSSA=ARRAY(175;LOSSA); ALOSSB=ARRAY(175;LOSSB)
22 ALOSSC=ARRAY(175;LOSSC); ALOSSD=ARRAY(175;LOSSD)
23 ALOSSO=ALOSSO*APROB; ALOSSA=ALOSSA*APROB; ALOSSB=ALOSSB*APROB
24 ALOSSC=ALOSSC*APROB; ALOSSD=ALOSSD*APROB
25 LOSSO=MAT(5,35;ALOSSO); LOSSA=MAT(5,35;ALOSSA); LOSSB=MAT(5,35;ALOSSB)
26 LOSSC=MAT(5,35;ALOSSC); LOSSD=MAT(5,35;ALOSSD)
27 A=SUMCOLS(LOSSO); B=SUMCOLS(LOSSA); C=SUMCOLS(LOSSB); D=SUMCOLS(LOSSC)
28 E=SUMCOLS(LOSSD)
29 BENI=MAT(5,35:A B C D E)
30 A=BENI(1,)-BENI(2,); B=BENI(1,)-BENI(3,); C=BENI(1,)-BENI(4,);
31 D=BENI(1,)-BENI(5,);
32 BENIAY=MAT(4,35:A B C D)
33 POP=KEPT(POP)
34 POP
35 POPINV=INVERSE(POP)
36 MBENIAY=BENIAY*POPINV
37 MBENIAY
38 YVPD1=KEPT(YVPD1)
39 YVPD1
40 TYVPD1=TRANPOSE(YVPD1)
41 X=YVPD1*POP
42 Y=SUMROWS(X)
43 POP7=DIAGMAT(Y)
44 POP7
45 POP7INV=INVERSE(POP7)
46 FINIAY=BENIAY*TYVPD1*POP7INV
47 FINIAY
48 LESSAVO=KEPT(LESSAVO); LESSAVA=KEPT(LESSAVA); LESSAVB=KEPT(LESSAVB)
49 LESSAVC=KEPT(LESSAVC); LESSAVD=KEPT(LESSAVD)
50 LESSAVO; LESSAVA; LESSAVB; LESSAVC; LESSAVD
51 LESSCONO=KEPT(LESSCONO); LESSCONA=KEPT(LESSCONA); LESSCONB=KEPT(LESSCONB)
52 LESSCONC=KEPT(LESSCONC); LESSCOND=KEPT(LESSCOND)
53 LESSCONO; LESSCONA; LESSCONB; LESSCONC; LESSCOND
54 TO=TRANPOSE(LESSAVO); TA=TRANPOSE(LESSAVA); TB=TRANPOSE(LESSAVB)
55 TC=TRANPOSE(LESSAVC); TD=TRANPOSE(LESSAVD)
56 W=KEPT(W)
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57 W
58 CITY2=KEPT(CITY2)
59 CITY2
60 WPO=TO*CITY2;WPA=TA*CITY2;WPB=TB*CITY2;WPC=TC*CITY2;WPD=TD*CITY2
61 LESSAVO=WPO*W;LESSAVA=WPA*W;LESSAVB=WPB*W;LESSAVC=WPC*W;LESSAVD=WPD*W
62 ALO=ARRAY(175:LESSAVO);ALA=ARRAY(175:LESSAVA);ALB=ARRAY(175:LESSAVB)
63 ALC=ARRAY(175:LESSAVC);ALD=ARRAY(175:LESSAVD)
64 ALO=ALO*APROB;ALA=ALA*APROB;ALB=ALB*APROB;ALC=ALC*APROB;ALD=ALD*APROB
65 LESSO=MAT(5,35:ALO);LESSA=MAT(5,35:ALA);LESSB=MAT(5,35:ALB)
66 LESSC=MAT(5,35:ALC);LESSD=MAT(5,35:ALD)
67 A=SUMCOLS(LESSO);B=SUMCOLS(LESSA);C=SUMCOLS(LESSB);D=SUMCOLS(LESSC)
68 E=SUMCOLS(LESSD)
69 BEN2=MAT(5,35:A B C D E)
70 A=BEN2(1,)-BEN2(2,);B=BEN2(1,)-BEN2(3,);C=BEN2(1,)-BEN2(4,);
71 D=BEN2(1,)-BEN2(5,.)
72 BEN2AV=MAT(4,35:A B C D)
73 NBEN2AV=NBEN2AV*POPINV
74 NBEN2AV
75 FIN2AV=NBEN2AV*TYVPD1*POP7INV
76 FIN2AV
77 NBENAV=NBEN1AV+NBEN2AV
78 NBENAV
79 FINAV=FIN1AV+FIN2AV
80 FINAV
81 LOSSCONO=V*LOSSCONO;LOSSCONA=V*LOSSCONA;LOSSCONB=V*LOSSCONB
82 LOSSCONC=V*LOSSCONC;LOSSCOND=V*LOSSCOND
83 TO=TRANPOSE(LOSSCONO);TA=TRANPOSE(LOSSCONA);TB=TRANPOSE(LOSSCONB)
84 TC=TRANPOSE(LOSSCONC);TD=TRANPOSE(LOSSCOND)
85 LOSSO=TO*CITY1;LOSSA=TA*CITY1;LOSSB=TB*CITY1;LOSSC=TC*CITY1;LOSSD=TD*CITY1
86 ALOSSO=ARRAY(175:LOSSO);ALOSSA=ARRAY(175:LOSSA);ALOSSB=ARRAY(175:LOSSB)
87 ALOSSC=ARRAY(175:LOSSC);ALOSSD=ARRAY(175:LOSSD)
88 ALOSSO=ALOSSO*APROB;ALOSSA=ALOSSA*APROB;ALOSSB=ALOSSB*APROB
89 ALOSSC=ALOSSC*APROB;ALOSSD=ALOSSD*APROB
90 LOSSO=MAT(5,35:ALOSSO);LOSSA=MAT(5,35:ALOSSA);LOSSB=MAT(5,35:ALOSSB)
91 LOSSC=MAT(5,35:ALOSSC);LOSSD=MAT(5,35:ALOSSD)
92 A=SUMCOLS(LOSSO);B=SUMCOLS(LOSSA);C=SUMCOLS(LOSSB);D=SUMCOLS(LOSSC)
93 E=SUMCOLS(LOSSD)
94 BEN1=MAT(5,35:A B C D E)
95 A=BEN1(1,)-BEN1(2,);B=BEN1(1,)-BEN1(3,);C=BEN1(1,)-BEN1(4,);
96 D=BEN1(1,)-BEN1(5,.)
97 BEN1CON=MAT(4,35:A B C D)
98 NBEN1CON=NBEN1CON*POPINV
99 NBEN1CON
100 FIN1CON=NBEN1CON*TYVPD1*POP7INV
101 FIN1CON
102 TO=TRANPOSE(LESSCONO);TA=TRANPOSE(LESSCONA);TB=TRANPOSE(LESSCONB)
103 TC=TRANPOSE(LESSCONC);TD=TRANPOSE(LESSCOND)
104 WPO=TO*CITY2;WPA=TA*CITY2;WPB=TB*CITY2;WPC=TC*CITY2;WPD=TD*CITY2
105 LESSCONO=WPO*W;LESSCONA=WPA*W;LESSCONB=WPB*W;LESSCONC=WPC*W;LESSCOND=WPD*W
106 ALO=ARRAY(175:LESSCONO);ALA=ARRAY(175:LESSCONA);ALB=ARRAY(175:LESSCONB)
107 ALC=ARRAY(175:LESSCONC);ALD=ARRAY(175:LESSCOND)
108 ALO=ALO*APROB;ALA=ALA*APROB;ALB=ALB*APROB;ALC=ALC*APROB;ALD=ALD*APROB
109 LESSO=MAT(5,35:ALO);LESSA=MAT(5,35:ALA);LESSB=MAT(5,35:ALB)
110 LESSC=MAT(5,35:ALC);LESSD=MAT(5,35:ALD)
111 A=SUMCOLS(LESSO);B=SUMCOLS(LESSA);C=SUMCOLS(LESSB);D=SUMCOLS(LESSC)
112 E=SUMCOLS(LESSD)
113 BEN2=MAT(5,35:A B C D E)
114 A=BEN2(1,)-BEN2(2,);B=BEN2(1,)-BEN2(3,);C=BEN2(1,)-BEN2(4,);
115 D=BEN2(1,)-BEN2(5,.)
116 BEN2CON=MAT(4,35:A B C D)
117 NBEN2CON=NBEN2CON*POPINV

```

```

118 NBEN2CON
119 FIN2CON=NBEN2CON*TYVPD1*POP7INV
120 FIN2CON
121 NBENCON=NBEN1CON+NBEN2CON
122 NBENCON
123 FINCON=FIN1CON+FIN2CON
124 FINCON
125 TAX=KEPT(TAX)
126 TAX
127 COST=KEPT(COST)
128 COST
129 TP=KEPT(TP)
130 TP
131 TAXCOST=COST*TP
132 COSTVY=TAXCOST*TAX
133 COSTPH1=COSTVY*POP7INV
134 COSTPH1
135 YVPD2=KEPT(YVPD2)
136 YVPD2
137 TYVPD2=TRANSPOSE(YVPD2)
138 COSTPH2=COSTVY*TYVPD2*POPINV
139 COSTPH2
140 LASTAV1=PINAV-COSTPH1
141 LASTAV1
142 LASTAV2=NBENAV-COSTPH2
143 LASTAV2
144 LASTCON1=FINCON-COSTPH1
145 LASTCON1
146 LASTCON2=NBENCON-COSTPH2
147 LASTCON2
*148 JOURNAL(OFF)
:End
MANUAL MODE
:_journal(off)

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PRINTOUT

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A P P E N D I X E

Complete model outputs

LOSSAVO (A 6 BY 5 MATRIX)

O	.017	.038	.096	.141
.01	.037	.045	.167	.136
O	.017	.043	.113	.162
O	.02	.05	.11	.163
O	.034	.033	.146	.21
O	.023	.06	.13	.18

LOSSAVA (A 6 BY 5 MATRIX)

O	.0122	.0342	.0612	.1215
.0053	.0265	.0655	.1166	.1679
O	.0136	.0355	.0915	.145
O	.0135	.0357	.0693	.1128
3E-4	.0265	.0629	.1063	.1605
O	.0181	.0626	.1052	.152

LOSSAVB (A 6 BY 5 MATRIX)

O	.0098	.0269	.0702	.1038
.0054	.0303	.0718	.1291	.1844
O	.0134	.0295	.0861	.1286
O	.0131	.0335	.0856	.1277
O	.0231	.0643	.1091	.1588
O	.016	.0447	.1025	.1453

LOSSAVC (A 6 BY 5 MATRIX)

O	.0088	.0246	.0587	.0902
.0054	.0247	.0686	.0985	.1377
O	.0085	.0222	.0579	.09
O	.014	.0354	.0733	.108
O	.0196	.0455	.0725	.1096
O	.0146	.0417	.0817	.1166

LOSSAVD (A 6 BY 5 MATRIX)

O	.0105	.0275	.0638	.0932
.0088	.0285	.0688	.1148	.163
O	.0108	.0288	.0702	.1027
O	.0143	.0392	.0724	.11
O	.0235	.0541	.0879	.1333
O	.0179	.0482	.0987	.142

LOSSAVO (A 6 BY 5 MATRIX)

O	.017	.038	.096	.141
.01	.037	.045	.167	.136
O	.017	.043	.113	.162
O	.02	.05	.11	.163
O	.035	.033	.146	.21
O	.023	.06	.13	.18

LOSSAVO (A 6 BY 5 MATRIX)

O	.0147	.038	.0933	.1407
.0064	.037	.045	.1656	.136
O	.0185	.0428	.1017	.1532
O	.02	.05	.1008	.1579
O	.035	.033	.146	.21
O	.022	.06	.1235	.1795

LOSSAVO (A 6 BY 5 MATRIX)

O	.0164	.038	.098	.141
.0068	.037	.045	.167	.136

0	.017	.043	.113	.162
0	.02	.05	.11	.163
0	.035	.033	.146	.21
0	.023	.06	.13	.18

LOSSCOND (A 6 BY 6 MATRIX)

0	.0137	.038	.086	.141
.0087	.0348	.045	.1405	.136
0	.0135	.0336	.0863	.1343
0	.02	.0496	.1031	.1555
0	.0317	.033	.1178	.1788
0	.0201	.058	.1128	.1642

LOSSCOND (A 6 BY 5 MATRIX)

0	.0143	.038	.086	.141
.0095	.037	.045	.1638	.136
0	.0163	.0421	.1033	.1528
0	.0194	.0487	.0863	.145
0	.0331	.033	.1242	.1804
0	.0228	.0584	.1255	.1765

V (A 6 BY 6 MATRIX)

1.8871E8	0	0	0	0	0
0	2.0098E8	0	0	0	0
0	0	3.3019E8	0	0	0
0	0	0	5.6396E8	0	0
0	0	0	0	2.3599E8	0
0	0	0	0	0	9.7561E8

CITY1 (A 6 BY 35 MATRIX)

ROW 1	.0238	.0736	.0255	.013	.0811	.0057	.043	.0794	.0457	.0392
	.028	0	.023	.054	.027	.012	.0256	.014	.029	.008
	.015	.0289	.055	.0517	.018	.011	.0356	.0056	.0148	.022
	.0273	.0025	.0077	.0198	.055					
ROW 2	0	0	0	0	0	.5754	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	.4246	0	0	0	0	0	0
ROW 3	0	.0171	0	.0183	0	.0272	0	0	0	.1978
	0	0	0	0	0	0	0	0	0	0
	.1994	0	0	0	0	0	0	0	0	.0577
	0	.169	.0233	.28	0	0	0	0	0	0
ROW 4	0	0	0	0	0	1	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
ROW 5	0	0	0	.1356	0	.1894	.0745	0	0	0
	0	.1885	0	0	0	0	.1265	0	0	0
	0	.0887	0	0	0	0	0	0	0	0
	.0629	0	.1639	0	0	0	0	0	0	0
ROW 6	.0044	.003	0	.0373	0	0	0	.0031	0	.0817
	.0225	.004	.0184	.0461	0	.1438	.0038	.0102	0	.0279
	.1883	.113	0	0	0	0	.006	.0164	.0287	.087
	0	.0585	.005b	.1351	.0048	0	0	0	0	0

PROB (A 5 BY 35 MATRIX)

ROW 1	.4508	.4242	.4373	.4465	.4296	.4443
	.4428	.4384	.4495	.4469	.4499	.443
	.448	.4508	.4485	.4207	.4418	.4848
	.4427	.4481	.4472	.4328	.4509	.4483
	.4504	.4112	.4468	.4134	.4263	.4332

205

	.4145	.443	.4291	.4414	.398	
ROW 2	.2606	.19387	.21251	.2328	.2015	.22689
	.22447	.21479	.247	.2342	.2476	.22359
	.242	.2514	.2337	.18931	.22141	.15527
	.23315	.2389	.23536	.20615	.2559	.2777
	.2825	.17844	.2342	.18134	.19618	.20559
	.18223	.22384	.1999	.22005	.16604	
ROW 3	.08863	.06011	.06736	.07606	.06358	.0735
	.07321	.06854	.08257	.076076	.08251	.07208
	.0901	.06488	.07685	.05807	.07148	.04549
	.07167	.07866	.07732	.06438	.08659	.09785
	.0885	.06408	.07635	.05505	.08074	.06451
	.06534	.07218	.06216	.07052	.04842	
ROW 4	.024845	.015787	.017929	.020589	.017224	.019763
	.02043	.018528	.023208	.020912	.022961	.019279
	.022012	.023465	.021362	.014929	.019596	.011122
	.018329	.021481	.021079	.016795	.024302	.028106
	.023397	.013522	.020888	.014018	.019772	.016888
	.01415	.01938	.018242	.01884	.012028	
ROW 5	.006449	.003524	.0041319	.004901	.0040517	.0046389
	.005058	.0043873	.005949	.005039	.005795	.0044219
	.005414	.00604	.00533	.0031872	.0047545	.0020282
	.004586	.0052143	.0051178	.0036384	.00618	.007497
	.006644	.00271	.005099	.002923	.0034278	.003703
	.0028766	.8.8038E-4	.0035804	.0043208	.0023184	

PDP (A 35 BY 35 MATRIX)

ROW 1	69077	0	0	0	0	0	0	0
ROW 2	0	53596	0	0	0	0	0	0
ROW 3	0	0	20201	0	0	0	0	0
ROW 4	0	0	0	81279	0	0	0	0
ROW 5	0	0	0	0	134495	0	0	0
ROW 6	0	0	0	0	0	22828	0	0
ROW 7	0	0	0	0	0	0	67899	0
ROW 8	0	0	0	0	0	0	0	66852
ROW 9	0	0	0	0	0	0	0	0
ROW 9	0	0	0	0	0	0	0	55886

ROW 25	0	0	0	0	0	0	0	0	0
ROW 26	0	0	0	0	0	0	42375	0	0
ROW 27	0	0	0	0	0	0	0	30238	0
ROW 28	0	0	0	0	0	0	0	0	108511
ROW 29	0	0	0	0	0	0	0	0	0
ROW 30	0	0	0	0	0	0	0	0	0
ROW 31	0	0	0	0	0	0	0	0	0
ROW 32	0	0	0	0	0	0	0	0	0
ROW 33	0	0	0	0	0	0	0	0	0
ROW 34	0	0	0	0	0	0	0	0	0
ROW 35	0	0	0	0	0	0	0	0	0

NSBENTAV (A 4 BY 35 MATRIX)

ROW 1	1.5111	4.0416	3.8636	1.8795	1.7555	108.63	2.267	3.7874	3.0228
	2.3637	2.0223	2.7802	1.3165	1.9567	1.7344	2.724	1.4184	.67772
	1.421	1.1384	2.3196	1.5488	3.4544	4.9958	1.6878	.90267	1.2127
	1.0596	1.4095	1.4046	1.7514	2.2942	1.8859	3.5203	1.0633	
ROW 2	2.6966	7.1124	6.8939	3.5972	3.1311	101.45	4.3063	6.7539	5.4242
	3.8348	3.548	8.8224	2.3089	2.7295	3.107	4.638	2.8801	1.1851
	2.3381	1.8502	3.7122	2.7414	6.2088	3.0133	3.0347	1.6026	2.1895
	1.8278	2.4408	2.3364	3.3888	3.9371	4.1478	4.2562	1.8773	
ROW 3	3.3265	8.824	8.3942	5.4763	3.8149	122.2	5.7684	8.246	6.6351
	5.5282	4.4467	17.389	2.8918	3.4277	3.7961	6.0101	4.2798	1.4648
	3.095	2.5091	5.3564	3.6464	7.5925	11.058	3.7167	1.9425	2.6513
	2.3128	3.0859	2.1888	4.5854	6.4774	7.2876	7.0448	2.287	
ROW 4	2.6832	7.1041	6.8338	3.5916	3.1074	98.975	4.421	6.6817	5.4182
	3.3633	3.3521	8.987	2.1788	2.9755	3.0974	3.8987	3.033	1.1044

208

2.5215 1.7238 3.6141 2.4477 6.1996 9.0456 3.0373 1.5765 2.1223
 1.6098 2.1973 2.2139 3.4517 3.7365 4.6911 4.7002 1.8126

YVPO1 (A 7 BY 35 MATRIX)

ROW 1	.0838	.0432	.031	.1894	.0483	.4798	.0618	.0718	.0873	.1844
	.0822	.3248	.0883	.141	.0738	.0688	.1002	.1423	.0956	.165
	.2508	.338	.1158	.0873	.089	.1112	.1103	.1266	.1002	.1852
	.0852	.2771	.1261	.1877	.1312					
ROW 2	.1714	.0745	.0369	.2685	.072	.3372	.0726	.0991	.0734	.2102
	.1193	.2984	.1817	.1819	.0945	.2308	.1416	.1819	.1148	.1962
	.2376	.2853	.1816	.1187	.094	.1377	.1853	.1667	.1451	.1919
	.0832	.287	.1089	.2058	.1477					
ROW 3	.1384	.0871	.0325	.2	.0884	.0333	.0655	.0826	.0862	.1733
	.1402	.1473	.1805	.1844	.1046	.2597	.1324	.1486	.1352	.1875
	.1717	.1675	.1572	.1222	.122	.1722	.1702	.1693	.1586	.1794
	.1089	.1851	.1168	.1756	.1677					
ROW 4	.1838	.0986	.0418	.1588	.0993	.0478	.1025	.1143	.0983	.1193
	.1385	.0789	.1453	.1461	.096	.1116	.1271	.1278	.1309	.1341
	.1231	.0887	.1707	.1068	.1274	.1128	.1455	.131	.1318	.1498
	.1182	.1143	.0993	.1256	.1266					
ROW 5	.1514	.0856	.1727	.0947	.1115	.0399	.0997	.0911	.1174	.0872
	.1282	.0507	.1356	.1172	.1028	.0862	.1401	.1117	.1144	.0998
	.0881	.0583	.1858	.1288	.1424	.1308	.1188	.0838	.1114	.1019
	.1301	.0814	.0878	.0848	.1269					
ROW 6	.184	.1416	.0941	.0798	.2121	.0262	.1884	.158	.234	.084
	.2094	.0695	.1294	.1322	.1758	.1135	.1977	.1495	.1373	.1136
	.0858	.0452	.1538	.2045	.2254	.1875	.1477	.1161	.1387	.1121
	.211	.0472	.1362	.0992	.1842					
ROW 7	.1288	.4814	.3812	.0351	.2684	.0396	.3889	.3728	.3224	.1316
	.1742	.0302	.1262	.0872	.3527	.1193	.1608	.1382	.2718	.1038
	.0446	.0238	.0813	.2308	.1988	.1477	.1551	.1947	.2141	.0801
	.2734	.0279	.3242	.1112	.1163					

POP7 (A 7 BY 7 MATRIX)

454288	0	0	0	0	0	0
0	523807	0	0	0	0	0
0	0	484184	0	0	0	0
0	0	0	367376	0	0	0
0	0	0	0	308747	0	0
0	0	0	0	0	387505	0
0	0	0	0	0	0	460885

FINIAY (A 4 BY 7 MATRIX)

4.6195	3.6174	2.2149	2.3521	2.3738	2.2258	2.423
5.7478	4.8399	3.5545	3.7076	3.7838	3.6938	4.0479
7.6077	6.4709	4.8588	4.9889	5.0309	4.8792	5.2793
9.6688	4.787	3.5018	3.6863	3.7434	3.6803	4.0313

LESSAVO (A 6 BY 5 MATRIX)

.01	.02	.04	.095	.13
.017	.043	.096	.168	.24
.01	.02	.043	.107	.17
.014	.025	.068	.12	.188
.01	.048	.072	.135	.194
.008	.02	.06	.126	.138

LESSAVA (A 6 BY 5 MATRIX)

.0061	.0143	.0354	.0669	.0949
.0121	.0259	.0778	.1308	.1816
.0082	.0149	.0288	.0834	.1335
.0102	.0188	.0478	.0782	.1209

209

.007	.0294	.0567	.1062	.1554
.0088	.018	.0485	.0981	.112

LESSAVB (A 6 BY 5 MATRIX)

.0088	.0135	.0327	.0717	.1009
.0123	.0368	.0756	.1488	.2161
.007	.016	.0379	.0838	.1464
.0118	.0184	.0505	.0955	.1486
.0062	.0263	.053	.1107	.1613
.006	.0165	.0443	.1005	.1157

LESSAVC (A 6 BY 5 MATRIX)

.0098	.0142	.0329	.0588	.0816
.0086	.0285	.0614	.1041	.15
.0057	.0115	.0287	.0651	.1033
.0088	.0163	.0426	.0701	.1074
.0081	.0217	.0477	.0874	.1286
.0083	.0141	.0393	.074	.0817

LESSAVD (A 6 BY 5 MATRIX)

.0072	.0169	.0385	.0698	.1001
.0111	.0316	.0687	.1144	.1627
.0088	.0127	.0324	.0716	.1149
.0117	.0185	.0499	.0788	.1173
.0086	.0247	.0509	.0875	.127
.0059	.0147	.0424	.0816	.0993

LESSCONO (A 6 BY 5 MATRIX)

.01	.02	.04	.095	.13
.017	.043	.096	.168	.24
.01	.02	.043	.107	.17
.014	.025	.066	.12	.189
.01	.048	.072	.135	.194
.008	.02	.06	.126	.139

LESSCONA (A 6 BY 5 MATRIX)

.01	.02	.04	.095	.13
.017	.043	.096	.168	.24
.01	.02	.043	.107	.17
.014	.025	.066	.12	.189
.01	.0405	.072	.135	.194
.008	.02	.06	.126	.139

LESSCONB (A 6 BY 5 MATRIX)

.01	.02	.04	.095	.13
.0163	.043	.096	.1647	.24
.01	.02	.043	.107	.17
.014	.025	.066	.12	.189
.0098	.0363	.072	.135	.194
.008	.02	.06	.126	.139

LESSCONC (A 6 BY 5 MATRIX)

.01	.02	.04	.095	.13
.0153	.0418	.0909	.1584	.2331
.0097	.0172	.0427	.0938	.1629
.014	.025	.066	.1158	.181
.0092	.0346	.0703	.1317	.193
.008	.02	.06	.1236	.139

LESSCOND (A 6 BY 5 MATRIX)

.01	.02	.04	.0879	.1221
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012

ROW 12				7538509		
ROW 13					54433771	
ROW 14						1.1626E8
ROW 15	1602372					
ROW 16		55708712				
ROW 17			45189404			
ROW 18				36566855		
ROW 19					38823525	
ROW 20						44963559
ROW 21						1.2962E8
ROW 22						
ROW 23	1.0804E8					
		35875817				

21

ROW 24

ROW 25

ROW 26

ROW 27

ROW 28

ROW 29

ROW 30

ROW 31

ROW 32

ROW 33

ROW 34

ROW 35

25874461

24857545

17737875

84200109

21442309

56807938

88887647

25657679

54210859

20253946

1.322318

35453428

CITY2 (A 6 BY 35 MATRIX)

ROW 1	.96	.979	1	.459	1	.2788	.919	.9902	1	.5656
	.8319	0	.8341	.8232	1	.4211	.767	.8526	1	.5587
	.2348	.4715	1	1	1	1	.966	.5743	.672	.5349
	.8902	.0997	.4011	.2828	.9968					
ROW 2	0	0	0	0	0	.137	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	.032	0					
ROW 3	0	.011	0	.0277	0	.059	0	0	0	.1377
	0	0	0	0	0	0	0	0	0	0
	.1488	0	0	0	0	0	0	0	0	.0659
	0	.3138	.064	.1988	0					
ROW 4	0	0	0	0	0	.1159	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0
ROW 5	0	0	0	.2247	0	.4084	.081	0	0	0
	0	.914	0	0	0	0	.2044	0	0	0
	0	.064	0	0	0	0	0	0	0	0
	.1098	0	.4523	0	0					
ROW 6	.04	.01	0	.2887	0	0	0	.0088	0	.2967
	.1881	.086	.1659	.1768	0	.5788	.0285	.1474	0	.4413
	.6182	.4645	0	0	0	0	.034	.4257	.328	.3992
	0	.582	.0728	.4865	.0434					

NSBEN2AV (A 4 BY 35 MATRIX)

ROW 1	2.6555	2.1069	2.2734	2.6785	2.1955	3.8906	2.5522	2.2952	2.5821
	10.714	2.4421	3.99	2.4101	2.4856	2.4721	1.8951	2.7177	1.8633
	2.3848	2.1883	2.0372	1.8711	2.8464	2.8361	2.7051	1.9619	2.435
	1.7268	1.8131	1.8409	2.1589	1.8766	2.8568	2.038	1.8193	
ROW 2	2.8858	2.2799	2.467	3.0343	2.3793	4.0629	2.7991	2.4904	2.7966
	11.324	2.679	4.7382	2.6447	2.7294	2.6783	1.9187	3.038	1.8234
	2.5657	2.4525	2.1962	2.2346	2.8682	3.0704	2.9295	2.1326	2.6467
	1.9296	2.1236	2.1187	2.3853	1.9059	2.2479	2.1057	1.9834	
ROW 3	3.0738	2.3781	2.5534	3.7725	2.4844	5.5727	3.0113	2.5923	2.8271
	14.431	2.8883	2.0534	2.9385	3.0584	2.784	2.4588	3.421	1.9602
	2.6625	3.0253	3.2027	2.7892	3.0043	3.2407	3.0771	2.1853	2.7907
	2.3227	2.4731	2.6527	2.555	3.0211	3.9946	3.0869	2.0373	
ROW 4	1.8249	1.4208	1.5015	2.9703	1.4537	4.5602	1.9383	1.5314	1.7075
	10.371	1.8929	5.5283	1.8614	1.9438	1.6353	1.8298	2.4556	1.2442
	1.5818	2.1491	2.5648	2.0852	1.7478	1.8745	1.7879	1.2949	1.8614
	1.6388	1.6822	1.9141	1.6972	2.5446	3.324	2.4436	1.2465	

FIN2AV (A 4 BY 7 MATRIX)

	2.8426	2.8344	2.7803	2.7438	2.7054	2.6593	2.7288
	3.0736	3.0671	3.012	2.9729	2.934	2.8889	2.9575
	3.8651	3.8125	3.8838	3.8804	3.476	3.3638	3.397
	2.8137	2.7422	2.6042	2.4941	2.3848	2.2721	2.2833

NSBENAV (A 4 BY 35 MATRIX)

ROW 1	4.1667	6.1485	6.137	4.558	3.951	112.52	4.8193	6.0827	5.6049
	13.098	4.4644	6.7702	3.7267	4.0424	4.2056	4.4181	4.1331	2.341
	3.7889	3.3247	4.3858	3.8199	6.1008	7.8318	4.3928	2.8646	3.6477
	2.7864	3.3225	3.3455	3.9103	4.1708	4.7427	5.5563	2.5828	
ROW 2	5.5824	8.3923	9.361	6.6315	5.5105	105.52	7.1054	9.2444	8.2208
	15.159	6.225	13.261	4.9536	5.4589	5.7852	6.5578	5.9281	3.0094
	5.1047	4.4027	5.9084	4.976	8.0742	12.084	5.9643	3.7352	4.8062
	3.7572	4.8644	4.4581	5.7721	5.4428	7.3857	6.3618	3.8807	
ROW 3	6.4003	11.2	10.848	8.2487	8.2793	127.77	8.7797	10.838	9.5829
	19.989	7.433	23.443	5.8303	6.4841	6.5901	8.4689	7.7008	3.4281

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	5.7575	5.5354	8.5591	6.4357	10.597	14.299	6.7938	4.1279	5.4421
	4.6451	5.559	8.8225	7.1504	6.4985	11.282	10.151	4.3443	
ROW 4	4.4781	8.5249	8.3353	8.5619	4.5611	103.54	6.3594	8.2132	7.1257
	14.335	5.245	15.513	4.0409	4.5191	4.7327	5.7285	5.4886	2.3486
	4.0432	3.8729	6.1789	4.5389	7.9474	10.82	4.8291	2.8714	3.7837
	3.2486	3.8784	4.128	8.1489	6.281	8.0151	7.1438	3.0591	

FINAV (A 4 BY 7 MATRIX)

	7.4621	6.4518	4.9852	5.0959	5.0792	4.8851	5.1518
	8.8215	7.907	6.5664	6.6805	6.7179	6.5807	7.0054
	11.473	10.283	8.5427	8.5893	8.5088	8.2426	8.6752
	8.4828	7.8081	6.108	6.1604	6.128	5.9324	6.2845

NBENICON (A 4 BY 35 MATRIX)

ROW 1	.46284	1.3191	1.2826	.37477	.58167	6.9748	.68474
	1.2376	.97611	.55883	.55608	.10862	.36392	.42308
	.84443	.53806	.37033	.20083	.45867	.25485	.46453
	.33788	1.1128	1.5785	.54053	.30737	.38254	.25168
	.38431	.32252	.53964	.48424	.34058	.56606	.31869
ROW 2	.10209	.30144	.30373	.042159	.13757	.96446	.16096
	.28885	.22869	.057431	.10336	0	.087916	.077542
	.13262	.027086	.06331	.039484	.11071	.028148	.02092
	.035801	.28027	.36713	.12635	.073498	.088204	.031486
	.052383	.033797	.12877	.0068562	.050466	.088733	.079321
ROW 3	.86322	1.8082	1.6705	1.8227	.75666	14.565	1.2698
	1.6467	1.2578	1.7472	1.0113	9.6152	.66065	.78162
	.72939	1.7547	1.0812	.35357	.60892	.67138	1.7732
	1.0136	1.4331	2.0192	.69439	.40424	.53746	.62473
	.79487	.84031	1.0806	2.0288	2.2838	2.1882	.80469
ROW 4	.47819	1.407	1.3468	.72794	.61908	18.059	.97573
	1.3162	1.0291	.84872	.54614	3.4693	.35792	.41421
	.59677	.40284	.75119	.2009	.4982	.22058	.40823
	.35816	1.1726	1.6521	.56813	.33074	.39938	.22013
	.32491	.29422	.82077	.42075	1.2532	.53044	.36867

FINICON (A 4 BY 7 MATRIX)

	.62833	.87383	.50158	.93078	.55812	.58605	.64423
	.088091	.083366	.078441	.08936	.10042	.10801	.13143
	1.6765	1.5199	1.2884	1.2556	1.2238	1.1712	1.2183
	.94608	.78242	.59677	.60821	.83714	.64152	.73682

NBEN2CON (A 4 BY 35 MATRIX)

ROW 1	0	0	0	.23014	0	.48884
	.079986	0	0	0	0	.8991
	0	0	0	0	.19911	0
	0	0	0	.056897	0	0
	0	0	0	0	0	0
ROW 2	.088022	0	.4065	0	0	
	0	0	0	.38255	0	.79783
	.13319	0	0	0	0	1.4976
	0	0	0	0	.33179	0
	0	0	0	.095151	0	0
	0	0	0	0	0	0
ROW 3	.14798	0	.88088	.0069671	0	
	.0013881	.0085473	0	.50938	0	1.1616
	.1691	2.5563E-4	0	.42554	.0053978	1.9009
	.0051412	.0058407	0	.012167	.42178	.002308
	0	.013227	.10585	.13151	0	0
	0	0	3.8827E-4	.0084014	.0072832	.043168
ROW 4	.18787	.18842	.88689	.18148	7.3474E-4	
	.18832	.08748	.10013	.78469	.096575	1.4308

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.30485	.10728	.1324	1.585	.19664	2.2509
.18891	.20665	.12119	.2442	.59005	.09177
.10845	.28895	.46387	.36656	.13782	.16169
.14551	.073638	.1304	.18972	.18322	.26135
.28333	.50779	1.1063	.4433	.075311	

FINCON (A 4 BY 7 MATRIX)

.043874	.036918	.026245	.025526	.025244	.025163	.025295
.074607	.062646	.044411	.043207	.042687	.042443	.042695
.17294	.15188	.1197	.11075	.10165	.093165	.091538
.47428	.44426	.39583	.36791	.34028	.3128	.30046

NSBENCON (A 4 BY 35 MATRIX)

ROW 1	.46284	1.3191	1.2826	.60491	.58167	7.4338	.76472
	1.2378	.97611	.55883	.55608	1.0057	.36392	.42308
	.56443	.53305	.56944	.20083	.46867	.25465	.46453
	.38448	1.1128	1.8788	.64093	.30737	.38254	.29188
ROW 2	.35491	.32252	.62766	.46424	.74708	.56805	.34868
	.10209	.30144	.30373	.42472	.13757	1.7823	.29415
	.28885	.22869	.057431	.10336	1.4576	.067916	.077542
	.13262	.027086	.4151	.039484	.11071	.028149	.018092
	.13106	.26057	.35713	.12625	.073498	.086204	.031456
	.052399	.033787	.27878	.0068562	.73114	.0987	.078321
ROW 3	.66463	1.9108	1.8709	2.132	.78866	18.726	1.4388
	1.647	1.2878	2.1727	1.0167	7.5161	.68579	.78746
	.72939	1.7669	1.503	.35588	.60892	.6847	1.879
	1.1451	1.4331	2.0192	.69439	.40424	.53846	.63313
ROW 4	.80225	.98368	1.2684	2.2144	3.1807	2.3397	.50542
	.83751	1.5044	1.4668	1.5126	.71566	18.49	1.2806
	1.4179	1.1618	2.1307	.74279	5.7202	.64683	.62086
	.71788	.84705	1.3412	.29267	.80665	.50953	.8721
	.72472	1.3104	1.8138	.71364	.40438	.52978	.40985
	.50813	.55557	1.1041	.92855	2.3595	.97374	.44398

FINCON (A 4 BY 7 MATRIX)

.87021	.61078	.52783	.55631	.58136	.48121	.66853
.1827	.14901	.12285	.13217	.14311	.15048	.17412
1.8494	1.6718	1.4081	1.3665	1.3255	1.2643	1.3099
1.4204	1.2267	.9523	.97412	.97743	.95442	1.0973

TAX (A 4 BY 7 MATRIX)

0	.001	.051	.1572	.221	.3481	.2237
.0738	.1424	.1685	.1898	.137	.1787	.132
.1345	.1715	.163	.1439	.1072	.1372	.1527
.0735	.1425	.1674	.1703	.1373	.1765	.1325

COST (A 4 BY 4 MATRIX)

1667247	1667247	1667247	1667247
774.46	774.46	774.46	774.46
2371374	2371374	2371374	2371374
3899420	3899420	3899420	3899420

YP (A 4 BY 4 MATRIX)

.1727	0	0	0
0	.28738	0	0
0	0	.51356	0
0	0	0	.056356

COSTPHI (A 4 BY 7 MATRIX)

.3184	.42311	.82187	.70049	.7356	.7988	.57326
1.4838E-4	1.9854E-4	2.4246E-4	3.2528E-4	3.417E-4	3.7106E-4	2.6629E-4

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.45420	.60181	.74241	.99633	1.0463	1.1362	.61536
.74702	.96958	1.2208	1.6383	1.7204	1.8683	1.3408

YVPD2 (A 35 BY 7 MATRIX)

.0087	.0148	.0132	.0208	.0229	.0229	.013
.0082	.0063	.0108	.0154	.0168	.022	.0383
.0018	.0018	.0018	.0028	.0131	.0088	.0289
.0234	.0284	.0237	.0209	.0175	.0113	.0043
.0075	.0084	.0131	.019	.0258	.0395	.057
.0215	.013	.0014	.0027	.0027	.0014	.0016
.0111	.0085	.0128	.0173	.0203	.031	.0488
.0104	.0123	.0129	.0204	.0198	.0278	.0541
.006	.0068	.0087	.0128	.0183	.0295	.034
.0894	.0871	.0807	.0716	.0632	.055	.0638
.0135	.015	.0197	.0248	.0281	.0371	.0256
.0033	.0026	.0014	.001	8E-4	9E-4	3E-4
.0218	.0343	.0382	.0387	.0447	.0344	.0283
.0501	.0553	.0629	.0644	.0622	.0567	.0346
.0078	.0083	.0103	.0122	.0158	.0218	.0363
.0215	.0817	.0779	.0432	.0449	.0427	.0372
.017	.0218	.022	.0271	.036	.041	.0265
.0194	.0212	.0194	.0218	.0227	.0244	.0188
.0188	.0181	.0252	.0319	.0332	.0321	.0528
.0244	.0248	.0266	.0246	.029	.0203	.0184
.1778	.1438	.1167	.108	.081	.0732	.0316
.0853	.058	.0404	.0308	.0212	.0137	.006
.0085	.0096	.0111	.0156	.0187	.0199	.0137
.0076	.0079	.0091	.0103	.0151	.0192	.0172
.0084	.0087	.0084	.0113	.0182	.0184	.0143
.0058	.0061	.0086	.0072	.0101	.0117	.0077
.0302	.0383	.0447	.0493	.0474	.0488	.0425
.0146	.0161	.0184	.0184	.0159	.0159	.0221
.0174	.0215	.0264	.0284	.0289	.0291	.0372
.0847	.0573	.0802	.0649	.053	.0473	.0282
.0087	.0082	.0107	.015	.02	.0261	.0281
.0876	.081	.0389	.0284	.0181	.0118	.0058
.0107	.008	.0096	.0104	.0112	.0138	.0275
.1104	.1035	.0991	.0915	.0834	.0703	.0654
.0134	.0129	.0165	.0181	.0193	.0227	.012

COSIMP2 (A 4 BY 35 MATRIX)

ROW 1	.41008	.8433	.70884	.38195	.33	.37251
	.86117	.59754	.54174	.65488	.54075	.1705
	.63075	.46491	.52914	.84274	.61624	.56867
	.79435	.48482	.74854	.28658	.38393	.48842
	.48073	.46565	.66702	.80028	.73941	.57748
	.68887	.4944	.64983	.64108	.45543	
ROW 2	1.9048E-4	3.0811E-4	3.2917E-4	1.8813E-4	1.5329E-4	1.7304E-4
	2.6067E-4	2.7758E-4	2.5164E-4	3.041E-4	2.5118E-4	7.9201E-5
	2.9299E-4	2.1596E-4	2.4579E-4	3.9147E-4	2.8625E-4	2.6415E-4
	3.6899E-4	2.2521E-4	3.4771E-4	1.3312E-4	1.8299E-4	2.2688E-4
	2.2331E-4	2.163E-4	3.0984E-4	3.7174E-4	3.4347E-4	2.6825E-4
	2.1102E-4	2.2968E-4	3.0185E-4	2.9778E-4	2.1155E-4	
ROW 3	.58328	.94343	1.0079	.61481	.48937	.52983
	.79817	.84988	.77033	.93114	.76819	.24251
	.89713	.66125	.7526	1.1987	.8765	.80883
	1.1298	.68958	1.0647	.40761	.5603	.69469
	.68376	.6623	.94872	1.1383	1.0517	.82136
	.85334	.7033	.82427	.9118	.64777	
ROW 4	.9591	1.5513	1.8374	.84653	.77182	.87124
	1.3128	1.3878	1.287	1.5311	1.2647	.38878

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..4752	1.0873	1.2376	1.871	1.4413	1.33
1.8579	1.1339	1.7507	.67027	.92134	1.1423
1.1244	1.0881	1.56	1.8717	1.7294	1.3506
1.966	1.1563	1.5198	1.4993	1.0652	

LASTAV1 (A 4 BY 7 MATRIX)

7.1427	8.0288	4.4732	4.3854	4.3436	4.0883	4.5785
8.8213	7.9068	6.5662	6.6802	6.7175	6.5803	7.0051
11.019	9.6818	7.8003	7.573	7.4607	7.1065	7.8609
7.7356	6.5195	4.8852	4.5221	4.4075	4.0642	4.9538

LASTAV2 (A 4 BY 35 MATRIX)

ROW 1	3.7588	3.4852	3.4284	4.1861	3.521	112.15	4.2581	3.4851	3.0832
	12.443	3.9237	6.5997	3.0959	3.5775	3.6775	3.5764	3.5169	1.7724
	2.9916	2.8399	3.6083	3.2333	5.7069	7.3415	3.9121	2.399	2.9807
	1.8881	2.5831	2.768	3.2407	3.8784	4.0929	4.9152	2.4272	
ROW 2	5.5822	8.382	8.2607	8.6313	8.5103	109.52	7.1051	8.2441	8.2208
	18.188	6.2248	13.281	4.8533	8.4587	5.785	8.5574	9.9278	3.0092
	9.1044	4.4023	8.9081	4.9758	9.074	12.084	8.964	3.735	4.8059
	3.7568	4.564	4.4549	5.7717	5.4427	7.3954	6.3616	3.6605	
ROW 3	5.817	10.257	9.9397	8.7339	5.8099	127.24	7.9815	9.9883	8.7917
	19.028	6.6639	23.2	4.9331	5.8228	5.8375	7.2702	6.8243	2.6163
	4.6377	4.8458	7.4944	8.028	10.037	13.604	6.11	3.4658	4.4533
	3.8068	4.8073	8.0011	8.1981	7.7993	10.358	9.2397	3.6966	
ROW 4	3.819	6.9736	6.8778	8.7154	3.7893	102.68	5.0489	6.8156	8.8587
	12.804	3.9803	15.115	2.5657	3.4317	3.4951	3.7575	4.0473	1.0186
	2.2253	2.739	4.4282	3.8666	7.026	9.7777	3.7007	1.7823	2.2236
	1.3769	2.1501	2.7774	3.5829	5.1247	6.4952	5.6445	1.9939	

LASTCON1 (A 4 BY 7 MATRIX)

.35081	.18764	.0058638	-.14418	-.18424	-.20759	.088267
.16255	.14582	.12261	.13284	.14277	.15008	.17385
1.3951	1.07	.66568	.37021	.2792	.12818	.4945
.67335	.2371	-.26849	-.66422	-.74301	-.91386	-.30348

LASTCON2 (A 4 BY 35 MATRIX)

ROW 1	.092784	.65578	.57384	.24296	.25187	.70612	.20355		
	.64004	.43438	-.095835	.015327	.83522	-.26683	-.041833		
	.03529	-.30369	-.046808	-.36784	-.32568	-.23018	-.28401		
	.10787	.71895	1.0911	.059798	-.15828	-.28448	-.5486		
ROW 2	-.38511	-.25496	-.041808	-.030167	.097255	-.075005	-.10674		
	.1019	.30113	.3034	.42455	.13742	1.7621	.28389		
	.28857	.22844	.057127	.10311	1.4875	.067623	.077326		
	.13237	.026694	.41481	.03922	.11034	.027924	.017745		
	.13093	.26039	.3669	.12603	.073282	.085894	.031084		
ROW 3	.052049	.033529	.27645	.0066266	.73084	.095402	.07911		
	.081363	.96736	.68262	1.6172	.28728	15.196	.84067		
	.79707	.4873	1.2416	.2478	7.2736	-.23134	.12621		
	-.023214	.56822	.62848	-.45296	-.82092	-.0048758	.81436		
	.73746	.87284	1.3245	.010625	-.25806	-.41026	-.50513		
	-.24944	.16231	.3161	1.5112	2.2565	1.4279	-.14234		
ROW 4	-.32159	-.046928	-.19045	.66604	-.056165	18.618	-.031908		
	.019848	-.10851	.58958	-.52194	8.3214	-.82838	-.48848		
	-.5186	-.1324	-.10006	-.10374	-.1.2512	-.82438	-.87862		
	.064448	.38905	.87143	-.41072	-.68468	-.1.0303	-.1.4619		
	-1.2212	-.79506	-.46191	-.22778	.83967	-.52559	-.62119		

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A P P E N D I X F

Workshop of January 24, 1986, Participants and Panelists

A number of the panelists met as an Advisory Committee at USC in order to help us evaluate an early draft of this report. Attending were:

Allen Asakura
Chief, Earthquake Safety Division
Department of Building and Safety
City of Los Angeles

Henry Bachrach
Risk Manager, Chief Administrators' Office
County of Los Angeles

Lou Blanck
Engineering Geologist, Building and Safety Department
County of Riverside

David Brauns
Senior Civil Engineer, Bureau of Engineering
City of Los Angeles

Paul Flores
Director, Governor's Office of Emergency Services
Southern California Earthquake Preparedness Project

Lawrence Gallagher
Risk Manager, Metropolitan Water District of Southern California

Jeri Hartman
Disaster Preparedness Operations Officer, Department of Disaster
Preparedness
County of Riverside

Marvin Hopewell
Senior Civil Engineer, Department of Planning and Building
City of Long Beach

Glenn Johnson
Principal City Planner, City Planning Department
City of Los Angeles

Margarita McCoy
Professor, Department Urban and Regional Planning
School of Environmental Design
California State Polytechnic University

Kathleen Tierney
Adjunct Assistant Professor, Institute of Safety Systems Management
University of Southern California

The following panelists did not attend the workshop yet participated in various cycles of the Delphi surveys:

Christopher Arnold
President, Building Systems Development, Inc.
San Mateo

Sharon Frank
Emergency Management Division
Orange County Fire Department

Roy Johnston
Vice-President, Brandow and Johnston and Associates
Los Angeles

Art Jones
City Manager (ret.)
City of El Segundo

Keith Julian
Manager, Economic Development Program
Southern California Association of Governments

William Kockelman
Earth Sciences Applications Planner, U.S. Geological Survey
Menlo Park

George Mader
President, William Spangle and Associates
Portola Valley

Sami Masri
Professor of Civil Engineering
University of Southern California

Shirley Mattingly
Chief Administrative Analyst, Chief Administrator's Office
Los Angeles

William J. Petak
Professor, Institute of Safety and Systems Management
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Haresh Shah
Professor and Chairman, Department of Civil Engineering
Stanford University

Robin Shepherd
Professor, Department of Civil Engineering
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Robert E. Wallace
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R. V. Whitman
Professor, Department of Civil Engineering
Massachusetts Institute of Technology

The investigators express their deep appreciation to the colleagues listed in the previous pages. Their patient help and advice was indispensable. They are, of course, not responsible for any errors in this study.