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Abstract

Major natural disasters in recent years have had high human and economic costs, and triggered record high postdisaster relief from governments and international donors. Given the current economic situation worldwide, selecting the most effective disaster risk reduction (DRR) measures is critical. This is especially the case for low- and middle-income countries, which have suffered disproportionately more economic and human losses from disasters. This article discusses a methodology that makes use of advanced probabilistic catastrophe models to estimate benefits of DRR measures. We apply such newly developed models to generate estimates for hurricane risk on residential structures on the island of St. Lucia, and earthquake risk on residential structures in Istanbul, Turkey, as two illustrative case studies. The costs and economic benefits for selected risk reduction measures are estimated taking account of hazard, exposure, and vulnerability. We conclude by emphasizing the advantages and challenges of catastrophe model-based cost-benefit analyses for DRR in developing countries.

Keywords

Catastrophe modeling, cost-benefit analysis, disaster risk reduction

Disciplines

Emergency and Disaster Management | Other International and Area Studies | Physical and Environmental Geography

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Risk Analysis

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Abstract:

Major natural disasters in recent years have had high human and economic costs, and triggered record high post disaster relief from governments and international donors. Given the current economic situation worldwide, selecting the most effective disaster risk reduction (DRR) measures is critical. This is especially the case for low- and middle-income countries, which have suffered disproportionately more economic and human losses from disasters. This paper discusses a methodology that makes use of advanced probabilistic catastrophe models to estimate benefits of DRR measures. We apply such newly developed models to generate estimates for hurricane risk on residential structures in the island of St. Lucia, and earthquake risk on residential structures in Istanbul, Turkey as two illustrative case studies. The costs and economic benefits for selected risk reduction measures are estimated taking account of hazard, exposure and vulnerability. We conclude by emphasizing the advantages and challenges of catastrophe model-based cost-benefit analyses for DRR in developing countries.

Key words: catastrophe modeling; disaster risk reduction; cost-benefit analysis

1. INTRODUCTION

The 7.0 Haiti earthquake in January 2010 destroyed a large part of the city of Port-au-Prince causing nearly 250,000 fatalities and 300,000 injuries mainly as a result of collapsed buildings. Haiti had no building code in place. International donors pledged more than \$5 billion in disaster relief and rebuilding effort. A few weeks later, populated areas in Chile experienced a much stronger 8.8 earthquake causing only 450 casualties, and many not from the earthquake but from the ensuing tsunami. Chilean building codes are some of the strictest in the world, comparable to the United States and Japan, and are well enforced ⁽¹⁾.

Chile is an exceptional case. Low- and middle-income countries generally suffer greater fatalities overall and losses as a proportion of their GDP from climate and geophysical events. During the period from 1970 to 2008, over 95 percent of recorded deaths from natural disasters occurred in developing countries. For middle-income countries, economic losses amounted to about 1 percent of GDP (1970-2006), which can be compared to losses of less than 0.1 percent of GDP for high income countries. In highly exposed countries, particularly Small Island Developing States, losses have been especially high; in the period 1970-2010 they exceeded one percent of GDP in many cases and eight percent in the most extreme cases ⁽²⁾.

Impacts and losses in developing and developed countries can be reduced substantially with cost-effective investments. The U.S. Federal Emergency Management Agency (FEMA) examined investments in 4,000 disaster risk reduction (DRR) programs, including retrofitting buildings against seismic risk and structural flood defence measures, and found an average benefit-cost ratio of four ⁽³⁾. In developing countries, a review of 21 studies on public and private investments as diverse as planting mangrove forests to protect against tsunamis, relocating schools to non high-hazard areas and strengthening the roots of banana trees to protect against windstorms, demonstrated with few exceptions equally high benefit-cost ratios ⁽⁴⁾.

In spite of potentially high returns, there is limited investment in loss reduction measures by public officials and by individuals residing in hazard-prone areas. In the United States, only about 10 percent of earthquake- and flood-prone households have undertaken cost-effective DRR measures ⁽⁵⁾. A principal reason for this inaction is a focus

on short time horizons: the upfront costs of the investment in DRR loom large relative to the perceived expected benefits from the measures ⁽⁶⁾. For similar reasons, policy makers are also reluctant to commit significant funds to risk reduction. In low-income countries, decision makers not only face fiscal constraints but also may lack adequate information on net economic and social benefits of DRR measures.

The failure to undertake measures to reduce disaster losses places pressure on political leaders to provide assistance after a disaster ⁽⁷⁾. This is true in OECD countries and low-income countries alike. A recent study in the United States shows how federal relief provided after disasters has grown significantly in the past 50 years. In the wake of Hurricane and Flood Diane in 1955, federal relief spending covered only 6.2 percent of total damages, but averaged 69 percent for disasters that occurred between 2005 and 2008 ⁽⁸⁾. In Germany, after the major 2002 Elbe floods, the German government provided the largest amount of public funds ever paid in the country's history to compensate uninsured flood victims, exceeding by far what was paid by insurance companies to insured victims ⁽⁹⁾.

This disaster relief spiral is even more striking in low-income countries. A recent joint report by the World Bank and the United Nations shows that bilateral and multilateral donors currently allocate 99 per cent of their disaster management funds for relief and reconstruction and only one per cent to reduce future loss exposure and vulnerability ⁽¹⁰⁾. To redress this imbalance, the 2005 United Nations World Conference on Disaster Reduction and the resulting Hyogo Framework for Action ⁽¹¹⁾ emphasize the need for pro-active disaster management including cost-effective risk reduction investments and, where this is not possible, risk transfer through insurance and other financial instruments ^{(12) (13)}.

A cornerstone of risk management is access to knowledge on risks and cost effective risk reduction measures. Cost-benefit analysis (CBA) has particular importance in this regard. Since the 1950s, CBA has been standard practice in the United States for the evaluation of risk reduction projects by organizations such as the Federal Emergency Management Agency and the Army Corps of Engineers. In the United Kingdom the Department for Environment, Food and Rural Affairs and the Ministry of Agriculture

also generally advocate the use of CBA for this purpose ⁽¹⁴⁾. France has also used CBA for public investment and transportation infrastructure projects for many years ⁽¹⁵⁾.

CBA for disaster risk management, however, has not been widely applied in developing countries ^{(4), (16), (17)}. When undertaken, the analysis typically uses a *deterministic* approach based on past disasters in the region of concern. For example, a study might estimate the costs and benefits of a specified risk reduction measure if the catastrophe that occurred 10 years ago were to occur again tomorrow -- a form of "what if" scenarios at a given point in time (see, for example, Dixit ⁽¹⁸⁾). A challenge in applying probabilistic CBA for DRR in developing countries is that the local knowledge on the methodology is often limited. As a result, many countries have not yet developed the technical capacity to go beyond a simple deterministic approach.

This paper applies CBA to examine DRR investments by using catastrophe modeling and demonstrates its applicability in developing country contexts. We use a state-of-the-art *probabilistic* disaster model to estimate the benefits of DRR measures for houses in St. Lucia, Caribbean (hurricane risk) and Istanbul, Turkey (earthquake risk). We then determine the impact of different discount rates and time horizons on the benefit/cost ratio of these measures. Finally, we discuss some extensions of this methodology and challenges of carrying out catastrophe model-based CBA on DRR measures in developing countries.

The paper is organized as follows: **Section 2** introduces probabilistic catastrophe modeling as a tool for evaluating the benefits of selected mitigation measures. We then apply this approach to two case studies in St. Lucia and Turkey in **Section 3**. **Section 4** concludes with a discussion of the results.

2. THE USE OF PROBABILISTIC CATASTROPHE MODELS IN COST-BENEFIT ANALYSES

A probabilistic catastrophe modeling approach provides more value than a deterministic approach because it includes all the events that can cause damage, and it generates a detailed analysis of return period based on advanced hazard models ^{(19), (20), (21)} ii A catastrophe model can be represented by four basic components or modules: hazard,

exposure, vulnerability and loss. In the first module, the risk of the *hazard* phenomenon is estimated. This module includes two main parts. The first part addresses the occurrence and frequency of the events. It does this by first developing a *stochastic event set*, a set of simulated events characterizing the observed or scientifically modeled events and their probabilities of occurrence. The second part of the hazard module calculates the severity of the events at every site of the study region.

For example, in the case of earthquake hazard, the first part of the *hazard* module specifies the earthquake size, frequency of occurrence, and location. The second part of the *hazard* module then calculates the amount of ground motion at a particular site for every stochastic event. In case of hurricane hazard, the first part of hazard module generates the storm tracks for each event in the set and the storm parameters such as wind speed and central pressure are estimated. The second part of the hazard module then takes into account the parameters such as surface roughness to calculate the final wind field in the entire study area.

The second module characterizes the exposure of properties at risk. This can be a building of specific interest, a dwelling representative of the average construction type in a given area exposed to the hazard (as we do in this paper and which would be useful to determine cost effective building codes), or an entire portfolio of buildings with different characteristics (for example, an entire city). The characteristics specified in this module include both physical characteristics of the building such as occupancy, construction type (e.g., wood versus concrete), age of the building, and number of stories as well as the replacement value of the property. If the impact of disaster on fatalities and injuries is included, then different scenarios characterizing the number of people in the buildings at different times of the day should also be devised.

The third module quantifies vulnerability, or how this property at risk (exposure) will behave physically under events that the hazard module has generated. Vulnerability functions are the relationships between hazard intensity (e.g. wind speed or ground motion) at the site and the level of damage experienced. Mean (average) damage estimates are expressed as a mean damage ratio, or MDR, which is the measure of the percent of value expected to sustain damage. Certain characteristics of the building affect its vulnerability. Hence it is important that they are specified in the *exposure* module as

much as possible. Vulnerability of buildings also depends on the study region because of different construction practices and enforcement of building codes.

The last module is *loss*. Assume for the moment that the only uncertainty is the occurrence of an event, i.e. the occurrence follows a Poisson process with its annual rate that was specified in the first part of *hazard* module. Given an event has occurred, the second part of the *hazard* module calculates the severity of the event at each site of the study. The *vulnerability* module then calculates the MDR at each location given the characteristics specified in *exposure* module, by taking into account the total value of damage given the replacement value and calculates the total loss at that location. The total *loss* from a single event is then the aggregate of losses in all sites of the study region due to that event.

These losses are then compiled and collected in a table called Event Loss Table (ELT). A typical ELT is shown in Table I below. Each row of ELT corresponds to a catastrophe event taken from a group of credible scenarios (e.g. earthquake, hurricane, flood) with an identification number (Event ID_j), an annual rate of occurrence (λ_j) and resulting loss (L_j) for ID_j.

Table I. Example of An Event Loss Table

Event ID	Annual Rate of Occurrence	Loss
1	λ_1	L_1
2	λ_2	L_2
..
j	λ_j	L_j
..
J	λ_J	L_J

Combining information on frequency and severity of losses, the probabilistic catastrophe model generates the distribution of the expected losses associated with all possible scenarios of disasters. This is often expressed in terms of an aggregate loss exceedance probability (EP) curve. For a given portfolio of structures at risk, an EP curve is a graphical representation of the probability p that a certain level of aggregated loss \$L will be exceeded in a given year. Figure 1 depicts a hypothetical mean EP curve, where the x-axis measures the loss in dollars and the y-axis depicts the annual probability that losses will exceed a particular level. In this figure, the likelihood that losses will exceed L_i is given by p_i .

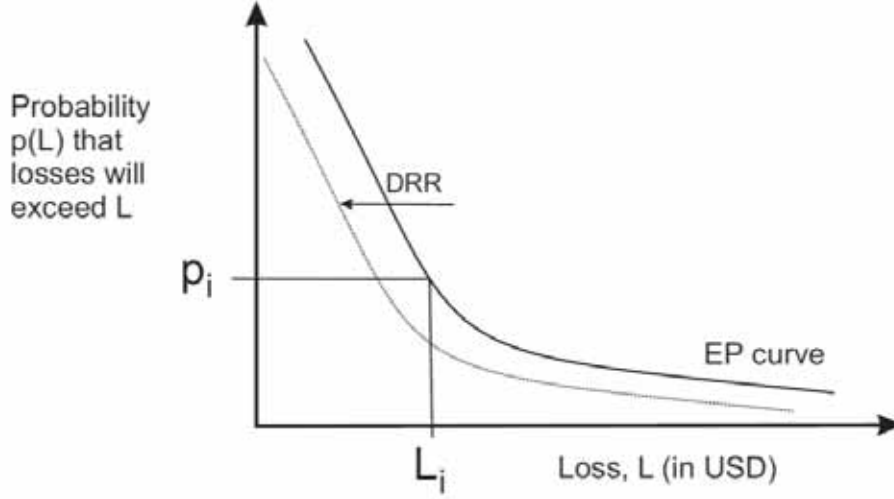


Figure 1: Example of An Exceedance Probability Curve and DRR Effect

To illustrate this methodology with an earthquake example, let's assume that natural disaster events follow a specific distribution, for instance a Poisson distribution with annual rate λ_j . Multiple events occurring at the same time would then be treated as a compound Poisson process with a rate $\lambda = \sum_j \lambda_j$. The corresponding loss distribution is

$$F(L < L_i) = \frac{\sum_{j>i} \lambda_j}{\lambda_j} \quad (1)$$

If n events occur in this compound process then the distribution of any loss L less than a particular value L_i is given as

$$F(L < L_i | n) = F(L_1 + L_2 + \dots + L_n < L_i) = F^{n*}(L < L_i). \quad (2)$$

$F^n(L < L_i)$ is the n^{th} power of F evaluated at loss L_i and $F^{n*}(L < L_i)$ is the n^{th} convolution of F evaluated at that loss. The annual probability p_i that the loss is greater than or equal to a particular value of L_i is:

$$P(L \geq L_i) = 1 - P(L < L_i) = 1 - e^{-\lambda} - \sum_{n=1}^{\infty} \frac{e^{-\lambda} \lambda^n}{n!} F^{n*}(L < L_i) \quad (3)$$

The reciprocal of this annual probability p_i is referred to as the *return period* of loss L_i for this specific exposure in the study area. Equation (3) determines the EP curve when secondary uncertainties around loss calculation for a given event are ignored.ⁱⁱⁱ

While a major advantage of catastrophe models is generating the range of probabilistic losses, the main loss metric used in the CBA of this paper is average annual loss (AAL), the area under the exceedance probability curve. AAL is a single loss metric that takes into account both severity and frequency of all possible events. Beyond expected average annual losses, risk averse policy makers may also be concerned with the right-hand tail of the curve where the largest losses are situated.

As shown in Figure 1, DRR measures would shift the EP curve to the left (the property would be more resistant to the hazard thus the loss for a given catastrophe scenario will be lower, all else being equal). There are many reasons why an EP curve is an improved risk measurement compared to single event or deterministic analysis. As a start, an EP curve provides information of interest to decision makers, including the full range of probabilistic estimates and their summary measures, such as the expected value, the variance and standard deviation ⁽²³⁾. Both frequent and infrequent (high-impact) events are represented, the latter being of special interest. For instance it is possible to generate the probable maximum loss for the exposure under study. This has important economic and social implications: in order for policymakers to establish proper construction codes, they need to know, for example, the possibility that the community or country could be struck by a Category 5 hurricane (measured on the Saffir-Simpson scale), or, as another example, that the most severe hurricane will never exceed Category 2. There is no need to spend scarce funds to make new construction resistant to a massive hurricane if the risk is extremely low. Conversely, if a coastal area is likely to be hit by hurricanes in the Category 3 to Category 5 range in the next 50 years, policymakers may wish to discourage new construction in harm's way. Both decisions have costs and expected benefits and only a detailed probabilistic catastrophe modeling process can provide the information necessary to make fully informed decisions. Similarly, international donors might be more willing to support a large DRR program if they learn from such a model that a large number of people are likely to die because of obsolete construction.

Moreover, since catastrophe models estimate the full range of possible hazard impacts and corresponding probabilities, it is possible to construct portfolios of DRR measures that are tailored to the specific circumstances of the risk at hand. Some DRR measures

may be cost effective or only feasible for high frequency events (e.g., structural mitigation measures), while others may be effective for low frequency events (e.g., insurance or catastrophe bonds)⁽²⁴⁾.

The first step in a catastrophe model-based CBA is thus the construction of an EP curve for a given hazard, exposure and location. This can be a house located in a designated hazard area, as is the case in this paper, or the analysis can focus on measures across a country or economy, for example, through the development of input-output models^{(25), (26)}. The next step is to simulate changes in the EP curve given selected DRR measures. In our cases, which examine earthquake and hurricane risk to representative houses, only the vulnerability module is affected since the likelihood and intensity of the hazard are not impacted by our selected DRR measures. EP curves *with* and *without* DRR measures in place can then be compared (Figure 1). The final step is an estimate of the benefit/cost ratio, which captures the complex interactions of the four main components of the model and the cost of the measure being analyzed. Since the benefits and costs accrue over multiple periods, the estimates require selecting one or a range of discount rates and of time horizons. A risk neutral decision maker is concerned about the reduction of annual average losses within an appropriate time horizon given an appropriate discount rate while a risk averse decision maker will also be concerned with the variability or standard deviation of these losses; value at risk may also be of interest to all decision makers⁽²³⁾.

3. CASE STUDIES IN ST. LUCIA AND ISTANBUL

This section applies a catastrophe model-based cost benefit analysis to two developing country cases: hurricane risk in St. Lucia and earthquake risk in Istanbul. We focus on a single residential structure that is representative of the type of construction in the area and begin by identifying feasible measures and their costs for reducing losses from the respective hazards. By applying a model of the hazard, we construct EP curves for that representative building with and without the selected DRR measures in place. Benefits of the DRR measures are quantified through reductions in the average annual loss (AAL) discounted over the relevant time horizons employing a range of discount rates. Finally,

we compute a benefit-cost ratio for each measure, and undertake a series of sensitivity analyses of the B/C ratio to variations in the discount rate and selected time horizon. The first case is St. Lucia for which we undertake a CBA by focusing only on the reduction in physical damage to the property; in our second case, earthquake risk in Istanbul, we repeat the same analysis but also extend it to include lives saved by different DRR measures.

3.1. Hurricane Risk in St. Lucia

St. Lucia is a small Caribbean island state that is highly exposed to hurricanes (Figure 2). Historically, the frequency and magnitude of the hazard have been greater than what is usual in the region. Hurricanes can have a devastating impact on St. Lucia's economy. For instance, losses from Hurricane Gilbert in 1988 were approximately \$1 billion or more than 350 percent of St. Lucia's GDP that year. By comparison, economic losses from Hurricane Katrina in 2005, the most devastating natural disaster in the history of the United States, represented only 1 percent of the U.S. GDP that year ⁽²⁷⁾.

Hurricanes and storms have caused extensive damage in St. Lucia in large part because the country has not implemented an updated national building code for all types of construction. The current code is based on the Organization of Eastern Caribbean States (OECS) model building code, which uses standards described by CUBiC (Caribbean Uniform Building Code). Unfortunately, CUBiC is now out of date (last updated in 1989), which means that the OECS standards also need to be updated ⁽²⁸⁾ iv



FIGURE 2. ST. LUCIA IN THE CARIBBEAN

The coastline of St. Lucia generally has a sharp topography, and although there are locations that can experience significant flooding, experts agree that storm surge does not create a significant loss potential. Hence, we focus our analysis on wind damage to housing structures.

While a large portion of St. Lucia's population is classified as below the poverty level, there is a rising middle class. This analysis focuses on middle class residential homes. Over 70 percent of residential buildings are constructed using concrete blocks (that is, masonry structures) or have wood outer walls such as plywood and wood/timber walls ⁽²⁹⁾. Two representative houses, one wood frame and the other masonry, were selected for study. It is assumed that the replacement value of the houses is 100,000 USD, which is aligned with current market prices. These representative houses are located in the higher and lower risk cities of Canaries and Patience, respectively.

The hazard for St. Lucia is modeled with the RMS Caribbean Hurricane Model RiskLink© 9.0. The methodology consists of three main steps:

1. Stochastic storm-track generation. A Monte Carlo set of storm tracks, with associated rates of occurrence, is generated using a random-walk technique and calibrated against

historical track data. The hurricane model's representation of hurricane frequency and severity is based on data describing historical storms that have impacted the Atlantic Basin and coastal regions since 1886. Sources for this catalog include data from the US National Oceanic and Atmospheric Administration, the U.S. National Hurricane Center, and the U.S. National Weather Service. The key historical database employed for model development is the HURDAT catalog, a compilation of tropical cyclones data starting from 1945.

2. Adding pressure histories to tracks. This process preserves the large-scale behavior of intensification and decay associated with variations in sea surface temperatures (SSTs) and topography across the basin, and calibrates the pressure distributions at all locations of interest within the basin. After the stochastic track set has been generated, pressure histories are added to the tracks using a second random-walk technique while the storms are over the ocean. The mean and variance of the rate of change of pressure across the simulation area are quantified from historical data.

3. Importance sampling to obtain a manageable number of hurricanes. Finally, the Monte Carlo storm set is importance-sampled to produce a boiled down storm set for loss calculations. Very large numbers of tracks are generated, equivalent to 100,000 years of simulated time. As it is not practical to run loss calculations with this number of tracks, the Monte Carlo event sets are importance-sampled. Tracks with similar paths and intensities at key locations (landfall or bypassing) are identified and grouped together. Most of the tracks are discarded and their rates are passed to the small number of tracks that are retained. Importance sampling is achieved by retaining a greater proportion of intense events than weaker events, which is a standard procedure. Loss convergence, as well as file sizes and run-time issues, are all considered when determining the final number of events retained in the event set. The reduced event set is used by the hurricane hazard module to compute wind speeds.

The above three steps comprise the first part of the hazard module described in section 2 where events are simulated and their rate λ_j is specified. In the second part of the hazard module, once the parameters of each hurricane in the stochastic set are

defined, the wind speed is calculated for each storm at each location in an analyzed portfolio. A hurricane generally begins to lose strength as soon as it makes landfall. The loss of strength after landfall is caused by two factors:

- As a hurricane progresses over land, the pressure at the center of the storm decreases due to a loss of the hurricane's source of energy (the warm ocean waters). Pressure-filling functions are based on meteorological studies and observed historical information;
- Surface friction from man-made and natural barriers reduces wind speeds as the hurricane moves inland. Surface friction characteristics at each site are represented by the eight directional roughness coefficients.

The Caribbean hurricane wind vulnerability model is region-specific to reflect the (mostly man-made) differences, such as construction quality, quality and enforcement of building codes, code-stipulated design wind speeds and the hurricane experience of each island. Within each region, the vulnerability of a building is defined in terms of four primary characteristics and a set of 25 secondary characteristics. The four primary characteristics are occupancy type, construction class, number of stories and year of construction.

Each vulnerability curve in the Caribbean hurricane model represents the average hazard exposure of a portfolio of buildings. The vulnerability of any particular building depends upon its specific characteristics, such as improved roof sheathing or the use of roof anchors. The variation in hurricane performance can be addressed in the RMS hurricane model through the use of secondary modifiers, which quantify the impact of the hurricane-sensitive features of a building on its vulnerability. Secondary modifiers can be used to quantify the impact of structural features or DRR measures.

Once the stochastic event set, associate rate for each event and the key parameters of each event are specified, the wind speed of the event at the location of representative houses is calculated. Specific to the wind speed, the vulnerability model calculates the damage ratio of the building, which is multiplied by the replacement value of the house to yield the total loss of event j at that location. Combining the information on each event

rate λ_j and the loss it creates L_j , the EP curves for the representative buildings can be constructed.

In the absence of DRR measures, the aggregate exceedance probability (AEP) curves for the representative residential buildings in the two cities are shown in Figure 3. These curves show that wood frame housing is more vulnerable than masonry, and there is a significant difference in expected loss between minimum and maximum hazard areas, e.g., AAL for wood frame structures is 4,070 USD in the Canaries (maximum hazard area) but only 830 USD in Patience (minimum hazard area).

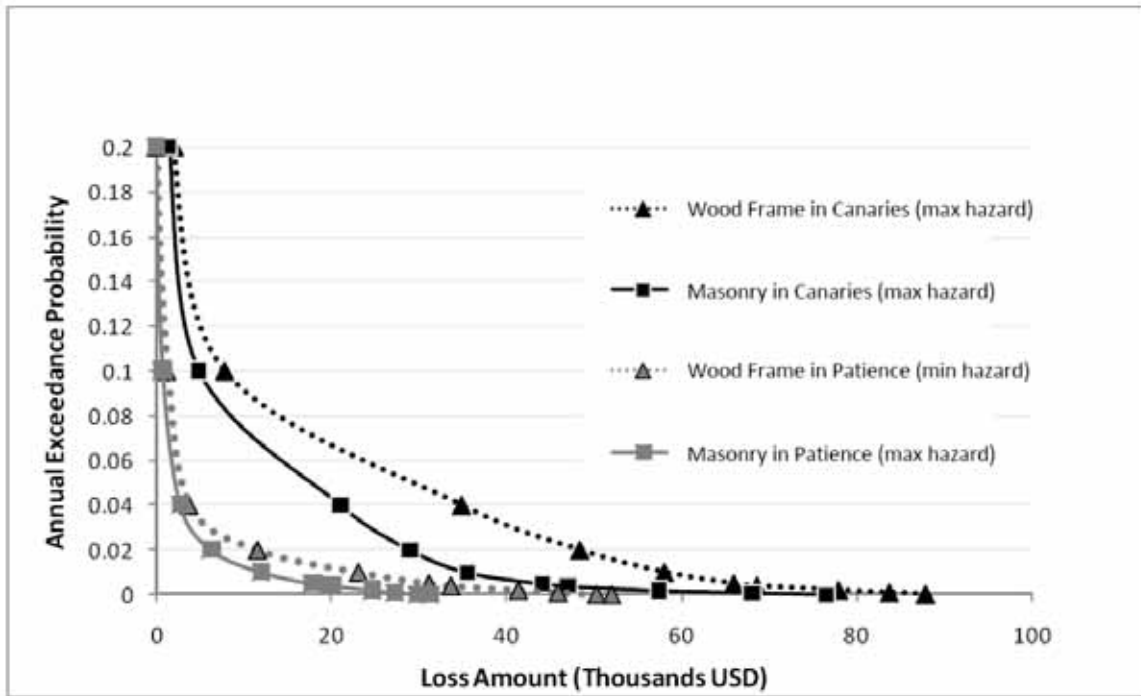


Figure 3. AEP curves with no DRR - Hurricane Risk in St. Lucia

Three DRR measures were examined for reducing hurricane risk to the representative wood frame and masonry houses. The DRR costs for the houses have been developed based on a RMS survey of DRR costs for contractors in Florida in 2009 and from earlier roofing costs reports⁽³⁰⁾, which are aligned with contractors' cost in St. Lucia.

- **Measure 1: Roof Upgrade:** This includes the replacement of the roof material with thicker sheeting and tighter screw spacing, as well the use of roof anchors. The total cost of this measure is estimated to be 9,200 USD.

- **Measure 2: Opening Protection:** This includes strengthening the resistance of windows and doors against wind and heavy pressure. The total cost is estimated to be 6,720 USD.
- **Measure 3: Roof Upgrade and Opening Protection:** Options 1 and 2 can be combined to provide a more comprehensive level of protection for the structure. The cost for both is estimated at 15,920 USD.

Applying CBA

Table II below shows the results of the benefit-cost calculations for the three measures. Not surprisingly, the results are highly sensitive to the choice of the discount rate, the assumed length of life of the residential structure and the hazard level. The results in Table II are based on discount rates of 5 percent and 12 percent (which are typical low and high values used for evaluating development projects, see Mechler ⁽³¹⁾) and an expected life of the structure of 25 and 50 years.

The results show that all three risk-reduction measures for the representative house in Canaries yield a positive net return when the discount rate is 5 percent for both wood and masonry structures that have an expected lifetime of 25 or 50 years. For wood frame with a 12 percent discount rate the B/C ratio is always close or above one, even though not higher than 1.59.

Table II: Summary of selected B/C ratios (ratios above 1 in bold)

DRR measure	Time horizon (years)	Masonry				Wood Frame			
		Canaries (Max Hazard)		Patience (Min Hazard)		Canaries (Max Hazard)		Patience (Min Hazard)	
		Discount rate		Discount rate		Discount rate		Discount Rate	
		5%	12%	5%	12%	5%	12%	5%	12%
1. Roof upgrade	25	1.37	0.76	0.29	0.16	1.73	0.96	0.41	0.23
	50	1.78	0.81	0.37	0.17	2.24	1.02	0.53	0.24
2. Opening	25	1.14	0.63	0.17	0.09	2.70	1.51	0.46	0.26

protection	50	1.47	0.67	0.22	0.10	3.50	1.59	0.60	0.27
3. Combined	25	1.09	0.60	0.20	0.11	1.80	1.00	0.36	0.20
	50	1.41	0.64	0.26	0.12	2.34	1.06	0.46	0.21

The highest B/C ratio is exhibited in the case of the opening protection (measure 2) in the maximum hazard location (Canaries) and for the more vulnerable structure (wood) (B/C ratios of 2.70 for a 25 year lifetime and of 3.50 for a 50 year lifetime; both at 5 percent discount rate). Note however that none of these measures yields a positive B/C ratio for representative houses in Patience, which are exposed to lower risk.

To illustrate the sensitivity of our results to the selected discount rate, in Figure 4 we show how the B/C ratio for strengthening the resistance of windows and doors for a wood frame house in Canaries varies over a range of rates (from 0 percent discount rate to 15 percent (0.15) on the x-axis) and across three different assumed lifetimes of the structure (10, 25 and 50 years shown as separate trend lines). By definition the lower the discount rate and the longer the lifetime of the house, the higher the B/C ratio for a given measure in a given location. For instance, assuming a 50-year lifetime of the building, the B/C ratio is a high 9.7 with a zero percent discount rate (upper curve on Figure 4). In other words, if this DRR measure were put in place this year, the expected benefit over a period of 50 years would be 9.7 times its cost. But it would only be 1.4 times its cost with a 15 percent discount rate. This example illustrates the critical role that the discount rate plays in any DRR cost-benefit analysis.

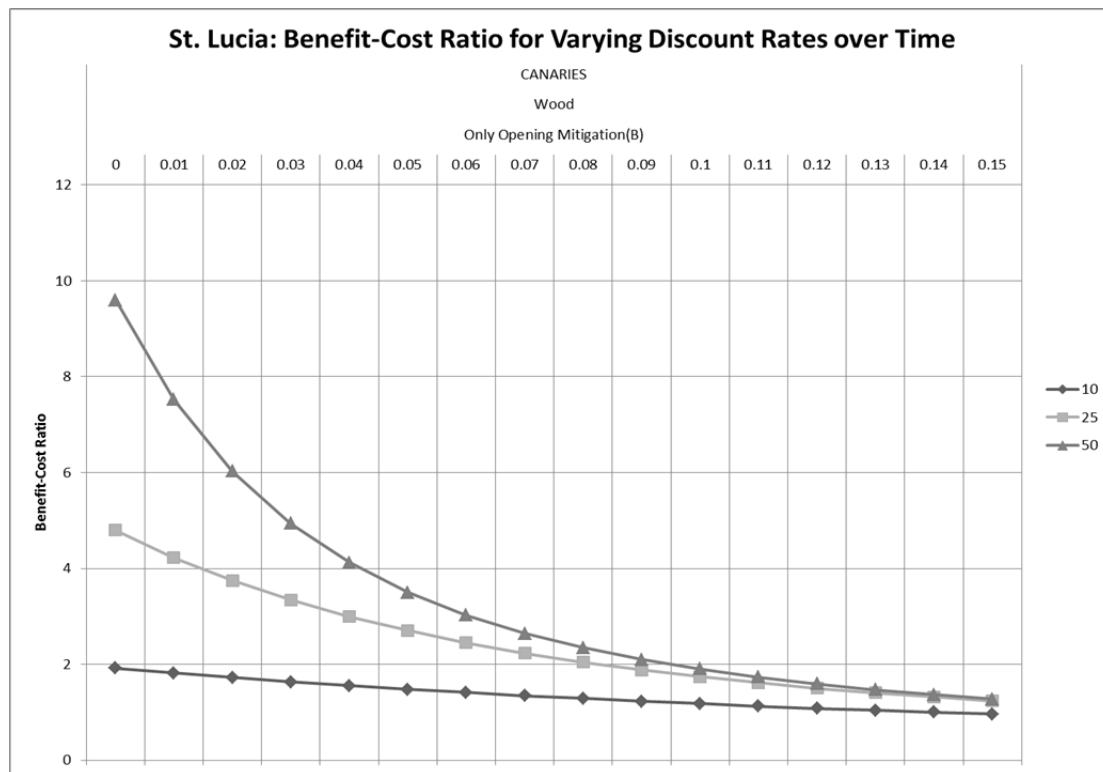


Figure 4. Canaries: B/C Ratios for selected discount rates and time horizons for protecting a wood frame structure by improving window and door protection

3.2. Earthquake Risk in Istanbul (Turkey)

Lying on the North Anatolian and the East Anatolian fault lines, Turkey is a hotspot of seismic activity. Istanbul is a thriving metropolis, with a population that has swelled from about 1 million in 1960, to currently over 11 million. Since Istanbul has not had a severe direct hit by an earthquake since 1766, there are fears that the city is due for a major earthquake. According to experts, there is a 30 to 60 percent chance of a magnitude 7 or greater earthquake impacting Istanbul in the next 25 years ⁽³²⁾. Given Istanbul's large population and its contribution to Turkey's GDP (about 40 percent), there is huge concern about its preparedness ^{(33), (34)}.

EP curve and selected DRR measures

The property loss estimation for Istanbul is based on a catastrophe model approach, namely the RMS[®] Turkey Earthquake Model in RiskLink© 9.0. The probabilistic event set includes more than 9,000 stochastic events from 116 seismic sources where multiple

segment ruptures are included to account for cascading events. A background zone covering the whole country is added to account for the possibility of an earthquake happening outside the geometries of the sources. The seismicity of each seismic source is based on historical events.

Geophysical models are also used to represent the seismicity where data on slip rate is available. Turkey has a very long recorded history of earthquakes that was used in estimating source seismicity. The earthquake catalog was developed in cooperation with experts from the Kandilli Observatory. The seismicity of Turkey is mainly characterized by shallow strike-slip type events that occur along the plate boundary. Deeper events (20-30 kilometers) take place on mainly normal faulting in the southwest part of the country.

Most earthquake sources in the Turkey Earthquake Model are assumed to follow a Poisson recurrence relationship. However, for the North Anatolian Source Zone and the Marmara Sea Source Zone, time dependency must be accounted for. Furthermore, for the Northwest Strand of the Marmara Sea Source Zone, effect of stress migration is taken into consideration in calculating the probability of stochastic events. Characteristic events are assumed to occur on the fault segment. Hence, for the Marmara Sea region, the probability of an event λ_j depends on three factors: 1) the source model, 2) the timing of the last event, and 3) the amount of stress transferred into one segment by an event taking place in adjacent segments.

To calculate the probability of occurrence of stochastic events in the Marmara Sea region, a logic tree approach is used where weights are associated with each branch of the tree. The weights were determined in consultation with Turkish seismic experts as well as from sensitivity analyses. The background source is modeled as having Poisson recurrence with the mean rate of occurrence calculated from a spatially smoothed historical seismicity. The smooth seismicity is obtained using an adaptive Gaussian Kernel technique.

For the second part of the *hazard* module, the model analyses the attenuation of seismic energy for a given event in order to determine ground shaking levels at a particular site. Attenuation relationships were calibrated against twenty-five historical events in the last 190 years, and amplification of ground shaking is based on high resolution soil and liquefaction data.

The *vulnerability* module builds on an estimate of 17 distinct building classes and four height ranges. The building exposure database is separated according to the year in which the building is constructed, which reflects revisions in the building codes. According to the World Housing Encyclopedia report on Turkey ⁽³⁵⁾ approximately 80 percent of Turkey's urban households live in mid-rise apartment blocks constructed of reinforced concrete with masonry infill. On average, a typical building has ten apartment units (two per floor) and five residents per unit. The representative structure selected in this study is thus a five-story reinforced concrete building with unreinforced masonry infill (similar to the structure analyzed by Smyth ⁽³³⁾ with a replacement value assumed to be \$250,000.

In the aftermath of the 1999 Kocaeli earthquake, most buildings of this type collapsed because the columns lacked adequate transverse steel reinforcement to resist lateral loads. Many buildings are designed with an open ground floor to accommodate other uses, such as parking, and some are characterized by a “soft story,” which means that the ground-floor space – a window or garage door – is situated where a wall might otherwise be. Soft story structures with large ground floor openings are at high risk of collapse from earthquakes with strong ground shaking ⁽³⁶⁾. Another construction detail that contributed to the breaking of the columns and possible collapse of the buildings is the gap between the columns and the infill wall, which reduces the effective height of the column (known as *short column*) ⁽³⁷⁾.

We selected two case study sites in Istanbul, Camlibahce and Atakoy, which are high- and low-hazard locations, respectively (Figure 5 shows the different levels of hazard for a 100-year return period).

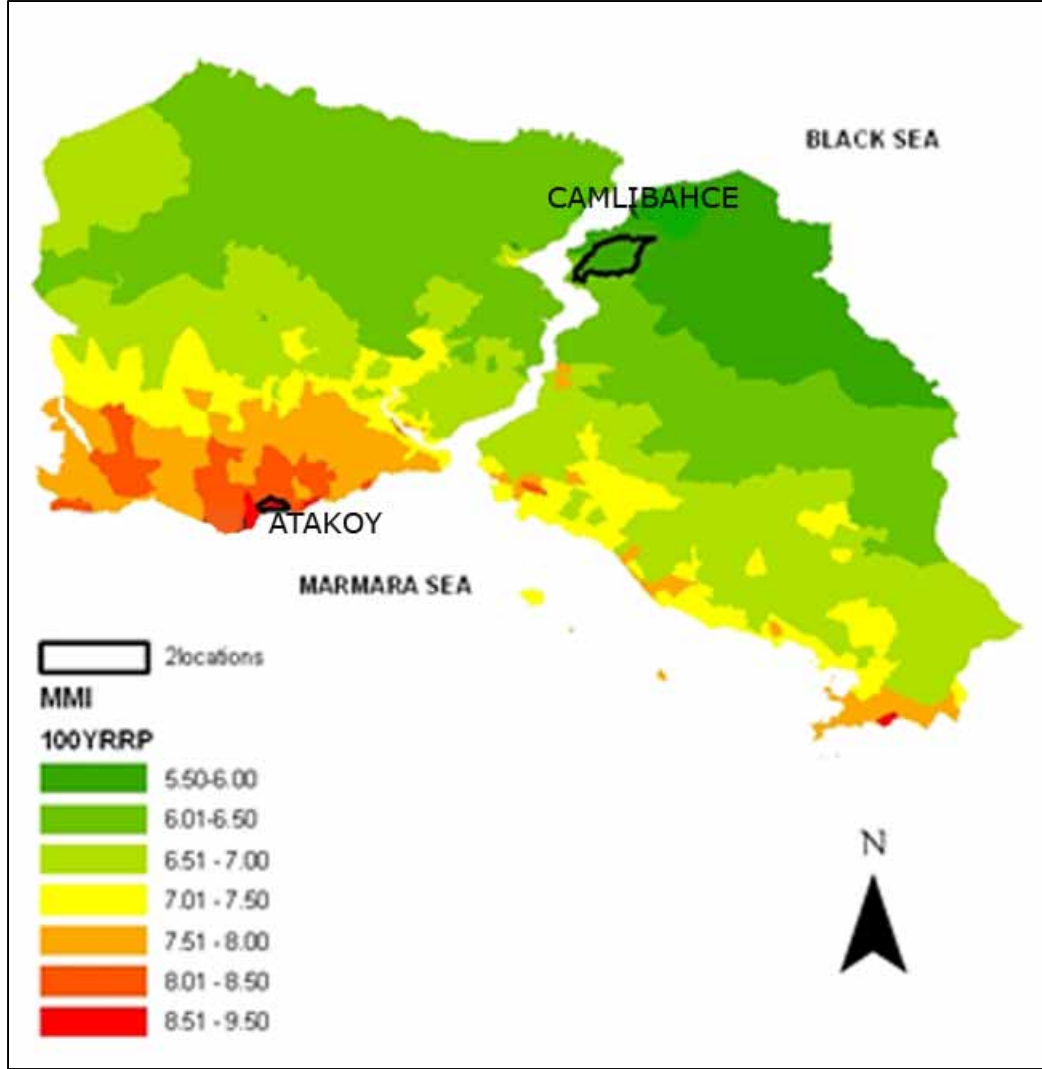


Figure 5: Istanbul 100-year hazard map (Istanbul at Mahalle level resolution) with two study locations representing high and low hazard locations

As shown in Table III, we analyze three different types of buildings, which are differentiated according to two characteristics of their construction, soft story, short column, and whether they need a structural upgrade.

Table III: Type of structures analyzed– unmitigated attributes

Type	Soft Story (SS)	Short Column (SC)	Need Structural Upgrade
Type 1	Yes	No	Yes
Type 2	No	Yes	Yes
Type 3	Yes	Yes	Yes

The flexibility of the frame can be increased by adding partial or full shear walls (structural upgrade) as described in Smyth ⁽³³⁾. In the absence of shear walls, risks to soft story buildings can be reduced through the use of a steel moment frame in the open floor. We assume that the addition of shear walls will correct for the effect of soft story, and therefore there is no additional cost for soft story DRR. The short column effect can be mitigated either through adding masonry inserts at both sides of the column ⁽³⁷⁾ or separation of the infill wall from the surrounding frame. According to RMS vulnerability analysis, Type 1 and Type 3 buildings are respectively about 4 percent and 14 percent more vulnerable than Type 2 structures.

Similar to the St. Lucia case study, for each event in the stochastic set of the hazard module, the possible damage (L_j) to the representative building is calculated. This damage information from individual events is then combined with the rate of occurrence of that event λ_j to generate the aggregate exceedance probability curves for the location.

Figure 6 illustrates the AEP curves for the different building types located in Camlibahce (minimum hazard; the three plain lower curves which overlap each other) and Atakoy (maximum hazard; three dashed upper curves which also somewhat overlap each other).

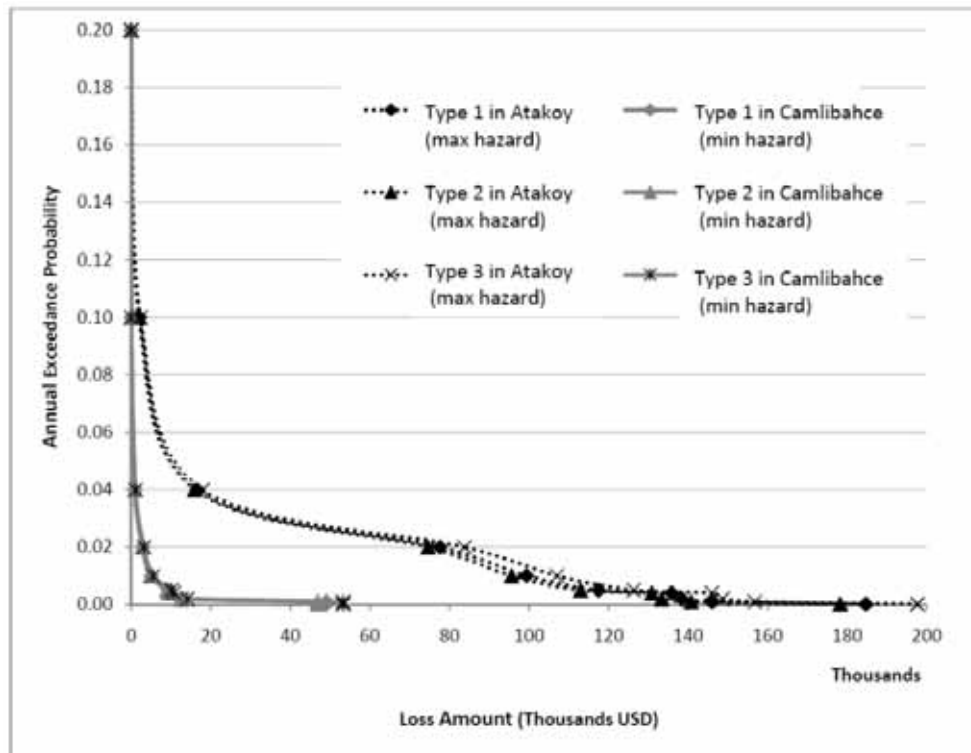


Figure 6. Earthquake risk in Istanbul - AEP curves with no DRR

Three DRR measures for reducing seismic risk to a representative five-story reinforced concrete building are analyzed. Measure 1 consists of retrofit short column (SC), and/or soft story (SS) but no shear walls are added; Measure 2 consists of adding partial shear walls (PSW) (short columns are mitigated if applicable); Measure 3 consists of adding full shear walls (FSW) (short columns (SC) are mitigated if applicable).

Based on Smyth ⁽³³⁾, Erdik ⁽³⁸⁾ and Burnett ⁽³⁹⁾, Table IV shows the cost of these DRR measures for each building type.

Table IV. Costs of alternative DRR measures for each baseline type

DRR Option	Costs (USD) Type 1	Costs (USD) Type 2	Costs (USD) Type 3
1. Mitigating SC/Mitigating SS	25,000	40,000	65,000
2. Mitigating SC/Adding PSW	80,000	120,000	120,000
3. Mitigating SC/Adding FSW	135,000	175,000	175,000

SS=Soft Story; SC=Short Column; PSW=Partial Shear Wall; FSW=Full Shear Wall. A

Cost-benefit calculations

By undertaking a similar analysis to that carried out for St. Lucia, Table V summarizes the B/C ratios for the DRR measures with two selected discount rates (5 and 12 percent) and a building lifetime of 25 and 50 years.

Table V. Earthquake risk in Istanbul: summary of selected B/C ratios

DRR Measure	Time Horizon (years)	Type 1				Type 2				Type 3			
		Atakoy Max Hazard		Camlibahce Min Hazard		Atakoy Max Hazard		Camlibahce Min Hazard		Atakoy Max Hazard		Camlibahce Min Hazard	
		Discount Rate		Discount Rate		Discount Rate		Discount Rate		Discount Rate		Discount Rate	
		5%	12%	5%	12%	5%	12%	5%	12%	5%	12%	5%	12%
Mitigating SC/ Mitigating SS	25	0.22	0.12	0.02	0.01	0.09	0.05	0.01	0.00	0.14	0.08	0.01	0.01
	50	0.28	0.13	0.02	0.01	0.12	0.05	0.01	0.00	0.18	0.08	0.01	0.01
Mitigating SC/ Adding PSW	25	0.22	0.12	0.01	0.01	0.09	0.05	0.01	0.00	0.12	0.07	0.01	0.00
	50	0.25	0.13	0.02	0.01	0.12	0.05	0.01	0.00	0.16	0.07	0.01	0.00
Mitigating SC/ Adding FSW	25	0.11	0.06	0.01	0.00	0.06	0.03	0.00	0.00	0.11	0.06	0.01	0.00
	50	0.15	0.07	0.01	0.00	0.08	0.04	0.01	0.00	0.14	0.06	0.01	0.00

SS=Soft Story; SC=Short Column; PSW=Partial Shear Wall; FSW=Full Shear Wal

The B/C ratios range from 0 to 0.28, meaning that – regardless of the hazard level, the time horizon and discount rate considered here – the costs of the three DRR measures outweigh the benefits. This means that from a purely financial standpoint, these measures are not recommended. However, the picture changes when one takes into account the value of reducing risk to human life as discussed below.

Integrating the value of reducing mortality risk into the CBA

Cost-benefit analyses of projects/investments that reduce mortality risk generally make use of a value of statistical life (VSL) to quantify the benefits of undertaking these measures (see Viscusi ⁽⁴⁰⁾ and Viscusi and Aldy ⁽⁴¹⁾ for a review). If a disaster risk mitigation project reduces the probability that an individual dies, conditional on the disaster event occurring, the project will save a number of *statistical lives* equal to the sum of reductions in the risk of death over the exposed population. The underlying rationale for estimating the VSL is to price life saving measures according to what people are willing to pay for a reduction in their mortality risk.

U.S. federal agencies, and some public agencies in the European Union, routinely make use of a VSL, but in the U.S. it differs greatly across agencies. According to the 2003 Office of Information and Regulatory Affairs guidance a statistical life saved is valued from USD 1 million to USD 10 million. The Environmental Protection Agency uses the point estimate of USD 6 million, while the Department of Transportation and the Department of Health and Human Services tend to use values that range from USD 3 to 5 million ⁽⁴²⁾. These estimates are based on empirical information regarding what people actually spend to reduce their risk of dying (revealed preferences), or how much they need be compensated to take on mortality risk (e.g. hedonic wage studies that estimate compensating wage differentials). Alternatively, the estimates are based on contingent valuation surveys asking people what they are willing to pay to reduce their risk or how much compensation they require to accept risk (stated preferences).

Despite this experience, assigning a specific statistical monetary value to a life, particularly if this value is lower for citizens of developing countries, can be controversial. For this reason, we undertake a sensitivity analysis using a *range* of statistical value of life estimates. As an upper bound of the VSL, we take the highest

practical estimate in the United States, USD 6 million ⁽⁴³⁾. As a lower range, we make use of a method suggested by Cropper and Sahin ⁽⁴³⁾, which scales the VSL (in this case, for Turkey) according to the country's per capita income relative to the U.S. This method yields a Turkish VSL approximately equal to USD 750,000. We use these figures as the lower and upper range of the VSL for the Istanbul case.

In Table VI we show how the B/C estimates change if we include the value of reducing mortality risk in the Istanbul analysis based on the assumption that 50 residents live in the building (10 families of 5 people, since the buildings we study here have 10 units). We take as an example the case of seismic retrofit using steel metal frames for a Type 1 constructed house in a high-risk area. As can be seen, the B/C ratios when VSL is not incorporated in the analysis (ranging from 0.12 to 0.28 depending on the discount rate and time horizon of the building) increase significantly if the value of reducing mortality risk is included. Even for the lowest VSL (USD 750,000) the DRR measure is now attractive (B/C ration higher than 1) assuming a discount rate of 5 percent. With the maximum VSL (USD 6 million) the B/C ratios range from 4.6 to 10.8 as a function of the discount rates and time horizons that we consider in our analyses, making investment in that DRR highly attractive.

Table VI: Earthquake risk in Istanbul: B/C ratios taking into account the value of life for baseline type 1 and measure 1 (ratios above 1 in Bold)

Analysis	Time Horizon (Years)	Atakoy Max Hazard	
		Discount Rate	
		5%	12%
Value of statistical life not included	25	0.22	0.12
	50	0.13	0.28
VSL= USD 750,000	25	1.3	0.7
	50	1.6	0.73
VSL= USD 6 million	25	8.1	4.6
	50	10.8	4.9

These findings confirm the result by Smyth ⁽³³⁾, who applied a VSL of USD 400,000, that only by including the value of statistical lives saved do earthquake strengthening measures for apartment buildings and schools in Istanbul pass the benefit-cost test.

4. DISCUSSION

The St. Lucia and Istanbul case studies underline the importance of examining the hazard, exposure and vulnerability, as well as the costs of the risk reduction measures and the appropriate discount rates and time horizons, for providing policy relevant information on DRR. A probabilistic catastrophe model-based CBA can provide valuable insights to public policy makers considering, for example, legislation on building codes or requiring building retrofits to protect against earthquake and hurricane hazards. Such analysis helps determine what DRR measure is effective for a given location and type of construction given the hazard profile. For instance, in the Istanbul case it is not the most vulnerable building (Type 3) that exhibits the highest B/C ratio, as one might think a priori, but rather the least vulnerable (Type 1) because of the lower costs of retrofit. By better targeting limited resources, probabilistic catastrophe model-based CBA can make DRR much more effective.

Note, however, that our results are based on several simplifying assumptions. As a start, as is the case for many developing countries, there is little information available on physical vulnerability of structures, and as a proxy we made use of information from developed countries (in the case of St. Lucia). On the benefit side, we did not take account of the value of household assets and the loss of livelihoods; nor did we include broader indirect avoided losses from disasters (e.g., health care cost or business interruption), which could increase the B/C ratio. In the St. Lucia case, we did not consider mortality risk, which is especially important for DRR in the developing world, where, as discussed in the introduction, reportedly over 95 percent of deaths from natural disasters occur.

If mortality and morbidity risk can be estimated it is important to ask how fatalities and injuries can be valued so as to be commensurate with other benefits and costs of the project. In the Istanbul case, we described and made use of a methodology for including this mortality risk based on adjusted values of statistical life (VSL) from available empirical studies. We noted that applying one specific VSL can be controversial, and for

this reason we recommend not making use of a point value, but applying a sensitivity analysis to the results over a range of VSL estimates.

As another simplification, we assumed that the risk will remain constant over the period of time for which we evaluated the expected benefits of the DRR measures. In reality the different components of the risk can change over time. This is particularly relevant today since climate change is expected to worsen extreme climate events in terms of duration, frequency and intensity⁽²⁾ in several parts of the world that are exposed to natural disasters, which is relevant for our St. Lucia analysis. One challenge of including climate change into CBAs though is the large uncertainty in the effect of climate change on hazard levels, given that to date it has not been possible to assign probabilities to different climate scenarios^{(44), (45)}.

Climate change is not the only driver of adjustments in risk over time. Any increase in hazard intensity might in the future be outweighed by autonomous decreases in asset *vulnerability* or *exposure* (e.g., individuals moving away from high hazard regions). Other drivers of increasing risk are economic development and urbanisation⁽⁴⁶⁾.

These caveats need to be considered by authorities charged with implementing risk management measures. In many instances, thus, a model-based CBA of disaster risk management may prove to be the most useful for outlining a process for rigorously weighing the benefits and costs (see Kull⁽¹⁷⁾).

Furthermore, while our analysis reveals what DRR measures are cost effective under a series of assumptions on the risk components, costs, discount rates and time horizons, this does not mean that such measures will necessarily be implemented. For instance, St. Lucia has not yet implemented a mandatory national building code, and increasing demand for houses could result in construction being undertaken without proper attention being given to its resiliency to future storms. In Turkey, a 2010 report prepared by a parliamentary research commission indicates that authorities have failed to improve city planning, reinforce substandard buildings, control urban development, and punish people who violate building codes. Due to the rapidly increasing population of Istanbul and resulting housing boom, tens of thousands of buildings are suspected of violating the existing building code. Overcrowding, poor land use planning, lacking infrastructure and services, and environmental degradation have all caused the earthquake risk in the city to

increase. Experts have recommended demolishing some 40,000 buildings that would probably collapse in a major earthquake, with hundreds of other buildings in dire need of reinforcement. Without implementation of DRR measures, Istanbul will likely suffer severe damage and large numbers of fatalities when the next major earthquake strikes ⁽⁴⁷⁾.

There are numerous economic and psychological barriers that can help explain why many countries have not implemented building codes, including the high capital costs of upfront DRR measures, individuals' misperception of risk, overconfidence in the ability to survive disasters, and difficulties in calculating the benefits of DRR. Moreover, recent studies in psychology show that individuals tend to utilize hyperbolic discounting, which gives a much higher importance weight to what can happen today (paying for a DRR measure) than to what can happen 5 or 10 years from now. Note also that our two case studies did not take risk aversion into account. Losses to housing structures were expressed in risk-neutral terms as mean damage ratios or expected losses. To take account of risk aversion, expected utility rather than expected value becomes the basis of the calculations, where expected utility might be expressed in terms of an equivalent monetary gain. How risk aversion plays a role for low-probability high consequences events is important since policymakers, who set policy based only on expected losses, could undervalue the importance of preparing for low-probability extreme events ^{(48), (49)}.

One way to potentially increase cost-effective investment in DRR would be for international organizations and donors to adopt a long time horizon and develop a larger portfolio of loan and grant mitigation programs. Probabilistic catastrophe model-based CBA can help them target the more efficient DRR investments and also measure financially the expected effectiveness of a given DRR program before it is implemented

⁽¹⁰⁾.

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ⁱⁱ Another important advantage is that probabilistic catastrophe models provide a more explicit and transparent description of the uncertainties in the analysis.

ⁱⁱⁱ There is significant variability in loss calculations due to uncertainties associated with the hazard itself, vulnerability calculations and incomplete information on exposure. Catastrophe models can include these secondary uncertainties but for simplification, this paper presents the results without secondary uncertainties. Dong ⁽²²⁾ explains the concept of secondary uncertainty in catastrophe models in great detail.

^{iv} Not only are the codes out of date, there is also a lack of enforcement of the current codes in place. The Development Control Authority of St. Lucia is responsible for enforcing building codes, but it encounters extreme difficulties in monitoring new settlements. Given the housing shortage in St. Lucia, houses can be shoddily constructed virtually overnight.