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### Article (Accepted version) (Refereed)

#### Original citation:

<u>Dietz, Simon (2011) High impact, low probability?: an empirical analysis of risk in the economics of climate change. Climatic Change, 108 (3). pp. 519-541. ISSN 0165-0009</u>

DOI: 10.1007/s10584-010-9993-4

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Available in LSE Research Online: January 2015

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## High impact, low probability? An empirical analysis of risk in the economics of climate change

Published in Climatic Change, 103(3), 519-541, 2011

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**Abstract** 

To what extent does economic analysis of climate change depend on low-probability, high-impact

events? This question has received a great deal of attention lately, with the contention increasingly

made that climate damage could be so large that societal willingness to pay to avoid extreme

outcomes should overwhelm other seemingly important assumptions, notably on time preference.

This paper provides an empirical examination of some key theoretical points, using a probabilistic

integrated assessment model. New, fat-tailed distributions are inputted for key parameters

representing climate sensitivity and economic costs. It is found that welfare estimates do strongly

depend on tail risks, but for a set of plausible assumptions time preference can still matter.

**Keywords** 

Catastrophe; climate change; cost-benefit analysis; discount rate; integrated assessment; risk;

uncertainty

JEL codes

Q54

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#### 1. Introduction

To what extent does economic analysis of climate change depend on low-probability, high-impact events? This question has received a great deal of attention lately, stimulated by the integrated assessment modelling undertaken by the Stern Review on the Economics of Climate Change (Stern 2007), and in particular by Martin Weitzman's critique of it (Weitzman 2007), which he has subsequently developed into a fuller discussion of how risk and structural uncertainty affect the empirical analysis of climate change (Weitzman 2009). Weitzman shows that, loosely speaking, uncertainty about the economic impacts of climate change could be so large that societal willingness to pay to avoid extreme outcomes should overwhelm other seemingly important assumptions, notably on time preference. If this analysis is correct, it represents a significant departure in the economics of climate change. Without wishing to oversimplify, our understanding of the problem shifts away from one cast largely in terms of a relatively sure intergenerational investment, where the key consideration is equity, towards one cast largely in terms of insuring against very unlikely but catastrophic risks.

This paper is an intuitive empirical analysis of the conceptual issues Weitzman raises, using an integrated assessment model (the PAGE model built by Chris Hope: Hope 2006a). In terms of methodology, the take-home message from Weitzman's (2009) analysis is that estimates of the economic impacts of climate change are better derived from probabilistic studies, in which, crucially, the key relationships linking (i) greenhouse gas emissions with warming (e.g. 'climate sensitivity') and (ii) warming with economic costs (the damage function) are described by distributions with a heavy or 'fat' tail of low-probability, high-impact outcomes. Thus I approximate one of the distributions for the climate sensitivity reported in the most recent assessment of the Intergovernmental Panel on Climate Change (IPCC: Meehl, Stocker et al. 2007), which has amongst the fattest tails of high-temperature outcomes, based on the largest ever sample of runs from a general circulation model (Stainforth, Aina et al. 2005), and I specify a

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<sup>&</sup>lt;sup>1</sup> There is some variability in the usage of terms such as heavy and fat tails (see also Nordhaus 2009). In some cases, the terms are used in a very broad sense to describe any probability distribution with a relatively high (though still very small absolute) risk of tail outcomes, especially in comparison with the normal distribution. In other cases, the terms are given specific, statistical meaning, though here too there is some variability. In this paper, I use the terms heavy and fat tail interchangeably to mean any distribution that has heavier tails than the exponential distribution.

damage function linking warming with economic costs that has uncertain curvature, with a probability distribution for that curvature which also has a fat tail.

While a few of the by-now numerous integrated assessments of the cost of climate change have worked with fat tails (e.g. Roughgarden and Schneider 1999; Tol 1999; Tol 2003; Mastrandrea and Schneider 2004), this paper takes an alternative angle, looking more closely at the implications of a potential climate catastrophe for the standard welfare model. The results of the modelling are analysed in terms of the probability of a 'catastrophe', which is taken to be the relative frequency with which arbitrarily large costs (e.g. 90% of consumption) occur, both under a scenario of business-as-usual emissions, and assuming emissions are controlled. Moreover, I explore the properties of welfare estimates as climate damage approaches the equivalent of 100% of consumption. This is a key element of Weitzman's (2009) analysis. In the most widely used empirical model of welfare, with constant and strictly positive relative risk aversion, marginal utility tends to infinity as consumption tends to zero, so if we think it possible that climate damage reaches 100% of consumption, then we have to find some means of bounding damages. One way to do so, suggested by Weitzman (2009), is to impose an upper bound derived from an analogy to the value of statistical life, but pertaining to whole societies. I show how estimates of expected utility vary as a function of this upper bound.

Overall, I conclude that welfare estimates do strongly depend on tail risks, but for a set of plausible assumptions time preference still matters. With fat tails on the climate sensitivity and on the curvature of the damage function, the PAGE model estimates some positive probability of catastrophe, even when the atmospheric stock of greenhouse gases is stabilised at 550 parts per million (ppm). To be more precise, PAGE's standard damage function produces a risk that the value of climate damages exceeds the total value of consumption at some point in the future, both in certain regions of the world and globally. Thus the magnitude of the catastrophe has to be bounded at less than 100% of consumption, and the expected welfare cost of climate change is very sensitive to small changes in where the upper bound is placed. However, accepting that there are *a priori* grounds for capping the total economic impact of a climate catastrophe, either when it occurs or as a matter of *ex ante* decision making, then time preference can still be an important consideration, because these tail-risk events are by definition unlikely to occur, and are relatively far off. Thus the rate at which utility is discounted, and the elasticity of substitution between consumption at different times, in different regions and in different states of nature can still matter.

Section two sets out the grounds for including catastrophic climate risks in the analysis, and assesses in more detail the extent to which existing economic studies have taken them into account. Section three discusses the properties of the standard economic model, with constant and strictly positive relative risk aversion, when faced with a climate catastrophe. Section four explains the modelling strategy, section five outlines the results, and section six provides a discussion and some concluding comments.

#### 2. Catastrophic climate risks

Economists' growing interest in catastrophic climate change has been largely stimulated by two concerns, first, that greenhouse gas emissions might lead to very large increases in global temperatures, and, second, that large (perhaps even modest) increases in temperatures, as an index of wider climatic changes, might lead to very large economic costs.

The former concern, about the relationship between greenhouse gases accumulating in the atmosphere and warming, has very often been captured by the notion of climate sensitivity, that is, the equilibrium increase in global mean temperature resulting from a doubling in the atmospheric concentration of carbon dioxide.<sup>2</sup> The IPCC's Fourth Assessment Report compiled a number of recent estimates of the climate sensitivity. Inspection of 18 probability density functions (pdfs) and frequency distributions in box 10.2 (Meehl, Stocker et al. 2007, p798) reveals that, in most if not indeed all cases, they have a positive skew, with a long tail of high estimates. These tails can be attributed to uncertainty about feedbacks, related for example to clouds and water vapour, and about the cooling effect of aerosols. This uncertainty is inherently difficult to reduce through measurements and modelling. In summarising this evidence, IPCC (2007) concludes that the best estimate of the climate sensitivity is 3°C, that there is a greater than 66% chance of it falling in the range 2-4.5°C (the IPCC's "likely" range), and a less than 10% chance of it being lower than 1.5°C ("very unlikely"). This leaves around a 17% chance that the climate sensitivity exceeds 4.5°C, and Weitzman (2009) roughly estimates, on the basis of the 18 pdfs, that there is a 5% chance of it exceeding 7°C, and a 1% chance of it exceeding 10°C.

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<sup>&</sup>lt;sup>2</sup> This is adequate for a first cut of this sort at the relationship between emissions and warming. However, the rate of warming depends on other factors than just the climate sensitivity.

In comparison with this evidence, previous integrated assessments seem in the main to have been too sanguine. In most cases, the climate sensitivity has been set to a single, best estimate, as part of a deterministic analysis. The best estimate has tended to be placed somewhere in the region 2.5-3°C (e.g. Nordhaus and Boyer 2000; Nordhaus 2008). The small number of probabilistic studies have naturally explored a wider range of values, but while some have considered distributions with a large positive skew (e.g. Tol 1999; Mastrandrea and Schneider 2004; Hope 2006b), others have employed ranges that tend to look unreasonably narrow when set against the evidence presented by IPCC (2007). This particularly concerns the Stern Review (Stern 2007), which used the PAGE model, representing the climate sensitivity with a triangular pdf fitted on a minimum value of 1.5°C, a best estimate of 3°C, and a maximum value of 4.5°C. Thus there was no chance of the climate sensitivity exceeding 4.5°C.

In addition to uncertainty about the climate-sensitivity parameter, the relationship between greenhouse gas emissions and warming is rendered still more uncertain by the risk of a significant positive feedback from the carbon cycle, effectively increasing the atmospheric stock of greenhouse gases for a given pulse of emissions. This feedback could come from a weakening of terrestrial biospheric and oceanic carbon sinks, from the release of methane as permafrost melts and wetland areas undergo drying, and/or even from the release of methane from oceanic stores of gas hydrates. Amongst the prominent economic integrated assessment models, Warren et al. (2009) explain that positive feedbacks have received little attention, with the exception of the PAGE model (Hope 2006a), which has a strong positive feedback calibrated on a weakening in terrestrial biospheric and oceanic sinks.

The latter concern above, about the relationship between warming and economic costs, is the preserve of economic models, and has been captured in a variety of ways.<sup>3</sup> For our purposes, an oft-used and tractable approach is to represent the general relationship between warming and climate damage by an aggregated damage function, expressing economic costs as a function of global mean temperature. This approach is taken in several well-known models (e.g. the PAGE and DICE models: Hope 2006a; Nordhaus 2008). Such a damage function is calibrated first through an estimate of overall impacts at 2.5°C or 3°C warming and second through an estimate of the

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<sup>&</sup>lt;sup>3</sup> Weitzman (2009) provides a theoretical treatment, showing that the standard approach is a special case of a more general model.

functional form. A very simple specification of such a function, used for example in PAGE, is as follows:

$$d_{j,t,s} = \alpha_{j,s} \left(\frac{T_{j,t,s}}{2.5}\right)^{\beta_s} \tag{1}$$

Where T is the increase in global mean temperature above pre-industrial in region j and at time t, d is the per-capita economic cost of climate change,  $\alpha$  sets the region-specific cost of 2.5°C warming and  $\beta$  is the damage-function exponent, determining curvature. s indexes Monte Carlo draws of the random parameters of a stochastic integrated assessment model like PAGE, including  $\alpha$  and  $\beta$ , and various parameters that jointly determine T. What is very important to bear in mind about d is that the economic cost of climate change includes not only losses in the production of goods and services that command market prices such as agricultural crops, but also the consumption equivalent of losses in non-market sources of welfare, such as health and natural ecosystems. I shall have cause to return to this point in section 3.

As there is some evidence to constrain estimates of the costs of 2.5°C warming (i.e.  $\alpha$ ), the more uncertain issue is what to assume about the functional form,  $\beta$ . Across the range of previous studies, the damage function has tended to be quadratic (which has the convenient analytical property of giving a linear marginal benefit function, but not much more justification than that) or less steep. Peck and Teisberg (1992) explored a cubic function in sensitivity analysis, and following this Hope (2006a) has specified a triangular pdf in the PAGE model, with a minimum value for  $\beta$  of unity (i.e. linear), a best estimate of 1.3, and a maximum value of 3. However, if we assume the global cost of 2.5°C warming is the equivalent of about 1.3% of GDP (which is the best estimate from PAGE)<sup>4</sup>, even a cubic function will lead to costs equivalent to only c. 10% of global GDP for 5°C warming, despite the fact that such warming could well trigger huge environmental changes (being for instance a greater difference in temperature than exists between the present day and the peak of the last ice age, the Last Glacial Maximum).<sup>5</sup> The general problem in specifying  $\beta$  is that we have almost no evidence on which to base it. While the belief is widely shared that impacts will

<sup>&</sup>lt;sup>4</sup> Ignoring for the moment the regional weighting on  $\alpha$ .

<sup>&</sup>lt;sup>5</sup> PAGE includes a separate function representing an economic catastrophe as a result of climate change (Hope, 2006a), but it has been shown to add relatively little (Warren, Hope et al 2006). To avoid double-counting, it is always switched off in the analysis of this paper.

become disproportionately more severe with rising temperatures, and while there are numerous partial examples of this<sup>6</sup>, we just do not know what a contemporary world that is on average 5°C warmer than pre-industrial times would look like; there is no data point. Nevertheless, it is increasingly recognised that values of  $\beta$  up to and including 3 may not give enough credence to the economic consequences of very large climatic changes, at least in the tail of the distribution.

#### 3. Welfare analysis of a climate catastrophe

In the applied welfare analysis of climate change it is usual to specify a social welfare function according to which utility is additive across regions, weighted by their populations, time periods, weighted by the utility discount rate (i.e. the rate of pure time preference or impatience), and states of nature, weighted by their probabilities:

$$E(V) = \sum_{i=1}^{J} \sum_{t=0}^{\tau} \sum_{s=1}^{S} p_s(u_{j,t,s}) N_{j,t} (1+\delta)^{-t}$$
 (2)

Where V is aggregate social welfare,  $\delta$  is the utility discount rate, and  $N_j$  is the size of the population of region j. There are J regions,  $\tau$  time periods and S Monte Carlo draws.

Clearly, a wide range of utility functions could be used in tandem with equation (2), but virtually all numerical studies specify the utility of a representative individual as an isoelastic or constant-relative-risk-aversion (CRRA) function:

$$u(c_{j,t,s}) = \begin{cases} c_{j,t,s}^{1-\eta} / (1-\eta) & \text{if } \eta \neq 1\\ \ln(c_{j,t,s}) & \text{if } \eta = 1 \end{cases}$$
 (3)

Where u is utility per capita, c is consumption per capita, and  $\eta$  is the elasticity of the marginal utility of consumption, which simultaneously captures aversion to inequality in consumption over regions and time, and relative risk aversion.

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<sup>&</sup>lt;sup>6</sup> For example, the relationship between warming and crop productivity at high temperatures (see chapter 3 of Stern, 2007, for a summary). There are of course counter-examples.

The CRRA utility function is the mainstay of applied welfare analysis, principally because it is very tractable, but this comes at the price of its limit properties, which are in fact of the utmost relevance when we think of a climate catastrophe, whose value might be close to, or even exceed, the value of 'baseline' consumption (i.e. the counterfactual value of consumption that would have been attained without climate change). In the limit as consumption tends to zero, the CRRA function has the following properties:

$$\lim c \to \begin{cases} 0 \\ 0 \\ 0^+ \end{cases} \quad u(c_{j,t,s}) = \begin{cases} -\infty \\ 0 \\ -\infty \end{cases} \quad \text{if } u(c_{j,t,s}) = \begin{cases} c_{j,t,s}^{1-\eta} / (1-\eta) & \text{and } \eta > 1 \\ c_{j,t,s}^{1-\eta} / (1-\eta) & \text{and } 0 < \eta < 1 \\ \ln(c_{j,t,s}) \end{cases}$$
 (4)

Where  $\eta \ge 1$ , utility tends to minus infinity as consumption tends to zero, while where  $0 < \eta < 1$ , utility tends to zero as consumption tends to zero. In all cases of strictly positive relative risk aversion  $(\eta > 0)$ , marginal utility, u', tends to infinity as consumption tends to zero.

The limit property of marginal utility is at the core of Weitzman's (2009) result, because his analysis is based on the expected stochastic discount factor, or in other words the amount of present consumption the agent would willingly give up to obtain one extra sure unit of consumption in the next period. If there exists one future state of nature in which the value of climate damage is at least as great as baseline consumption, even if it is very, very unlikely, society's *marginal* willingness to pay to avoid zero consumption is then unbounded, for strictly positive relative risk aversion.<sup>7</sup>

To formalise this point, note that in an integrated assessment model, all that is required for expected marginal utility to be unbounded is that  $\eta$ >0 and:

$$\exists s, \quad s \in [1, S] : d_{j,t} \ge \hat{c}_{j,t} \quad j \in [1, J], t \in [0, \tau]$$
 (5)

Where d is the per-capita cost of climate change (e.g. from equation 1), and  $\hat{c}$  is baseline consumption per capita. That is, there exists a state of nature s (i.e. a Monte Carlo draw), belonging to the set of states of nature S, such that damages in some region j at some time t are at least as

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<sup>&</sup>lt;sup>7</sup> Of course any utility function with the same limit property for u' will have the same effect.

large as baseline consumption per capita (or, in other words, such that *net* consumption per capita is no greater than zero).

In the catastrophe region j at the time of the catastrophe t, expected marginal utility will be unbounded when  $\eta>0$ , as it will be at the global level (infinity plus a finite number gives infinity). Now, if we look not at marginal analysis but at the computation of the underlying social welfare function in equation (2), expected total utility will also be unbounded at the regional and global level, as long as  $\eta\ge1$ . But, interestingly, if  $0<\eta<1$  then expected total utility will be zero in the catastrophe region j at the time of the catastrophe t, which in turn allows expected total utility to exist globally, and it will presumably be positive given positive (net) consumption and therefore utility in other regions. This wrinkle is worth reiterating. For  $0<\eta<1$ , expected total utility exists even though expected marginal utility does not. Therefore, strictly speaking, we could proceed with the computation of social welfare even if the condition in equation (5) were satisfied, as long as  $\eta<1$ . However, since there are several arguments with regard to the economics of climate change suggesting  $\eta>1$  (e.g. Dasgupta 2007; Weitzman 2007), this particular way of circumnavigating the problem is unsatisfactory.

It is also important to note that the condition in equation (5) is in fact different to the condition required in Weitzman (2009) for expected marginal utility to be unbounded. Since Weitzman takes the expectation of a stochastic discount factor over an infinite pdf, he does not require the probability that  $d_{j,t} \ge \hat{c}_{j,t}$ ,  $j \in [1,J]$ ,  $t \in [0,\tau]$  be positive. Rather, expected marginal utility can be unbounded even if the probability of catastrophe is zero, as long as the probability of extremely low levels of net consumption per capita falls more slowly in the limit as  $c \to 0$  than marginal utility rises, so that the latter dominates asymptotically. The point is that this asymptotic effect cannot be reproduced by any empirical study such as this one, which works with a finite set of states (i.e. a finite Monte Carlo sample of draws of the infinite pdf), even if the Monte Carlo 'grid' is infinite in the sense of not having a finite support on the pdf (cf. Weitzman 2009 pages 16-17). In a finite sample, net consumption is either positive or it is not, and utility either exists or it does not. Therefore we work with the stronger condition (5).

Considering evidence on the climate sensitivity and damage function, condition (5) is far from implausible. In standard applications of PAGE, the most vulnerable region of the world to climate

change is India and Southeast Asia, in which the best estimate of  $\alpha$  from equation (1) (the damage cost of climate change for 2.5°C warming) is 3.1% of regional GDP. Assuming  $\beta$  takes its maximum value in standard PAGE of 3, condition (5) will be met for India and Southeast Asia if regional temperature exceeds around 7.5°C above pre-industrial at some point over the modelling horizon, which is two hundred years. A global catastrophe is not implausible either. Let  $\alpha$  and  $\beta$  equal 2.5% of global GDP and 3 respectively. 8 Then condition (5) will be met with respect to global aggregate consumption if the global mean temperature exceeds about 8.5°C above pre-industrial. Assuming high business-as-usual emissions of greenhouse gases throughout this century (the A1FI SRES scenario: Nakicenovic and Swart 2000), IPCC estimates up to a 17% chance of global mean temperature exceeding 6.4°C by 2100 alone (IPCC 2007). With the more general assumption of stochastic emissions, a recent study by scientists from MIT goes further, estimating a median increase in global mean temperature of 5.2°C by 2100, with a 10% chance of more than 7.4°C warming (Sokolov, Stone et al. 2009). Moreover, the argument was made above that even higher values of  $\beta$  should be considered. Therefore, certainly if our task is to evaluate the cost of businessas-usual emissions over long periods, and even perhaps in a scenario in which emissions are controlled but climate and 'welfare' sensitivity are very high (e.g. in India and Southeast Asia, or in Africa), we need to bound the problem in some way.

Let net consumption per capita be bounded from below, so that damages never exceed some share  $\phi$  of baseline consumption per capita:

$$c_{j,t,s} = \min[\hat{c}_{j,t} - d_{j,t,s}, \hat{c}_{j,t}(1 - \phi)]$$
(6)

In effect, the probability distribution on the random variable  $\tilde{c}_{j,t}$  is truncated. In principle  $0 \le \phi < 1$ , but since the upper bound only exists to create a finite solution where it would not otherwise exist (i.e. where  $d_{j,t} \ge \hat{c}_{j,t}$ ), in practice  $\phi$  is close to, but strictly less than, unity.

There is another way in which the problem might be bounded, which is to deliberately specify a functional form for climate damage, where  $d_{j,t} < \hat{c}_{j,t}$  for all finite values of T. Nordhaus' (2008) DICE model does just this, specifying  $d_t = 1 - 1/(1 + \alpha T_t^{\beta})$ , according to which  $\lim_{T \to \infty} d = \hat{c}$ . Thus, for

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 $<sup>^{8}</sup>$   $\alpha$ =2.5% is the maximum assumed value in standard PAGE, ignoring regional weighting.

any finite value of *T*, the value of damages cannot exceed 100% of baseline consumption per capita. Should this restriction be placed on the damage function? Opinions will differ, but it is possible to argue that it should not be, for two related reasons. First, recall that our measure of climate damage includes the consumption equivalent of losses in non-market sources of welfare, so when we allow for the possibility of zero *net* consumption per capita, this is not the same as saying consumption of marketed goods and services is zero. Second, notice that the question is whether the value of climate damage can exceed the flow of net consumption in a given year, not the total stock of net wealth. It seems possible that the flow of what I am labelling *net* consumption is negative for a period of time, even though the stock of net wealth remains positive. The two reasons relate to each other, because it would clearly be implausible to say that the sum of our willingness to pay to avoid climate change (based on avoided reductions in the production of marketed goods and services, as well as to avoided damages to non-market sources of welfare) exceeds our total wealth. Similarly, no rational forward-looking agent with CRRA utility would let consumption of marketed goods and services go to zero or below if the stock of wealth is positive. But neither of these are in general implied by condition (5).

If it is considered possible that condition (5) could be met in reality, the PAGE damage function in equation (1) allows for this.  $\phi$  is then required to create a finite solution and logically has the interpretation that a decision maker has a finite *ex ante* willingness to pay to reduce the risk of zero net consumption (which is positive). Even if condition (5) were thought impossible to meet,  $\phi$  can still be used to represent the maximum proportion of net consumption that could feasibly be lost.

Weitzman (2009) takes the approach of limiting ex ante willingness to pay, specifying  $\phi$  as a function of the value of statistical life. The benefit of this is that the bound on net consumption losses can at least be related to a familiar concept in economics, one that has been estimated many times, with the key point being that the evidence (and indeed intuition) suggests it is finite (Viscusi and Aldy 2003). In other words, we do not personally appear to be willing to liquidate all of our wealth to reduce the probability of our own death. The wrinkle in this case is that, since we are dealing with whole societies, if not indeed the whole of human civilisation, the directly relevant construct is "the value of statistical civilization as we know it or maybe even the value of statistical life on Earth as we know it" (Weitzman, 2009, p6).

The purpose of this analysis is not to consider in detail the estimation of  $\phi$ , but rather to reason from the opposite side by looking at its role in estimating the welfare cost of climate change, for a plausible range of values close to unity. To proceed then, substituting (6) into (2) gives:

$$u(c_{j,t,s}) = \begin{cases} \{\min[\hat{c}_{j,t} - d_{j,t,s}, \hat{c}_{j,t}(1-\phi)]\}^{1-\eta} / (1-\eta) & \text{if } \eta \neq 1\\ \ln\{\min[\hat{c}_{j,t} - d_{j,t,s}, \hat{c}_{j,t}(1-\phi)]\} & \text{if } \eta = 1 \end{cases}$$

$$(7)$$

From (7) and (4) it follows straightforwardly that, in any state of nature where  $d_{j,t} \ge \hat{c}_{j,t}$ , for some region j and time t, as  $\phi$  tends in the limit to unity, net consumption per capita will tend to zero, utility per capita will tend either to minus infinity or zero depending on  $\eta$ , and marginal utility will tend to positive infinity when  $\eta > 0$ .

At this point, we can make three postulates about the results of the empirical analysis. First and foremost, given that marginal utility is very high when consumption per capita is very low, we might anticipate that, if condition (5) is met, then the expected loss in social welfare  $E(\Delta V)$  due to climate change will be sensitive to the upper bound on damages,  $\phi$ . Even superficially small differences in  $\phi$  (say, raising it from 0.99 to 0.995) might lead to large differences in  $E(\Delta V)$ .

Second, if there exists some state of nature s in which  $d_{j,t} \ge \hat{c}_{j,t}$ , and in the plausible case where  $\eta \ge 1$ , then in the limit as  $\phi \to 1$ , the marginal change in social welfare with respect to the rate of impatience,  $\delta$ , will tend to zero *in that state*. In fact, it is also true that the marginal change in social welfare with respect to  $\eta$  will tend to zero *in that state*. To see this, take that catastrophic state of nature to be s=1. From (7), given condition (5) is satisfied:

$$\inf_{d_{j,t,1}} \geq \hat{c}_{j,t} 
\Rightarrow \lim_{d \to 1} u(c_{j,t,1}) = -\infty \quad \text{iff } \eta \geq 1$$
(8)

$$\Rightarrow V_1 = \sum_{j=1}^{J} \sum_{t=0}^{\tau} (u_{j,t,1}) N_{j,t} (1+\delta)^{-t} = -\infty$$
(9)

$$\Rightarrow \frac{\partial V_1}{\partial \delta} = 0 \quad and \quad \frac{\partial V_1}{\partial \eta} = 0 \tag{10}$$

This is the postulate that, in the limit, only the risk of catastrophe determines the welfare cost of climate change, and the social discount rate is unimportant.

But, third, as long as  $\phi$ <1 and the solution is therefore finite,  $\delta$  and  $\eta$  will still be important determinants of the expectation of social welfare when integrating over *all* states of nature. To illustrate, consider just two states of nature, s=1,2. For s=1, as before  $d_{j,t} \geq \hat{c}_{j,t}$ , so the upper bound is applied as per equation (6). As  $\phi$  becomes vanishingly close to unity, the loss in social welfare in that state,  $\Delta V$ , becomes vanishingly close to 100%. But for s=2, where  $d_{j,t} < \phi \hat{c}_{j,t}$ , the marginal change in social welfare with respect to  $\phi$  is zero; i.e. the upper bound never kicks in. If  $d_{j,t}$  in state of nature 2 is equal to, say, zero, then the expected welfare loss  $E(\Delta V)$  is bounded from above at just under 50% as  $\phi \rightarrow 1$  in the limit.

The key conceptual point here relates to non-uniform convergence. For any given  $\phi$  <1,  $E(\Delta V)$  is an arbitrarily small function of  $\delta$  and  $\eta$ . Conversely, for any given  $\delta$  and  $\eta$ ,  $E(\Delta V)$  is infinite when  $\phi$  =1 and  $d_{j,t} \geq \hat{c}_{j,t}$  for some s, and if  $\eta \geq 1$ . This is a specific formulation of the more general theorem proposed in Stern (2008, p20). In the very simple example above,  $d_{j,t} \geq \hat{c}_{j,t}$  for s=1 and  $d_{j,t}$ =0 for s=2, so as  $\lim \phi \to 1$ ,  $E(\Delta V)$  is bounded from above by  $p[d_{j,t,s} \geq \hat{c}_{j,t}] = 0.5$ . This result generalises of course to many states of nature, with different economic losses. The only condition is that the probability of a catastrophe should be less than unity. Then as  $\lim \phi \to 1$ ,  $E(\Delta V)$  will asymptote to some arbitrarily small function of  $\delta$  and  $\eta$ .

#### 4. Modelling strategy

To reiterate, the purpose of this paper is to test empirically the extent to which economic analysis of climate change depends on low-probability, high-impact events in the climate-economy system. I do so using the PAGE integrated assessment model (Hope 2006a). PAGE is particularly suitable for the present study because it is probabilistic, set up to conveniently run Monte Carlo-type simulations. Furthermore, at least as far as previous, thin-tailed assessments are concerned, PAGE has been demonstrated to provide a good approximation to the results of a range of other models, both of parts of the climate-economy system and of the whole system. As an example of the former, Warren et al. (2009) show that PAGE on the whole provides a good approximation to the

projections of atmospheric carbon dioxide concentrations, radiative forcing and warming made by much more complex general circulation models of the climate system, for a consistent set of assumptions about carbon dioxide emissions. As an example of the latter, the mean estimate of the social cost of carbon dioxide made by Hope (2006a) is close to the central estimate from a range of peer-reviewed studies (Tol 2005).

In common with most applications of integrated assessment models, I avoid an equation-by-equation description of PAGE for the sake of brevity. However, the model is set out in full in Hope (2006a) and Hope (2008) and unless otherwise stated no changes have been made. Here I focus on the parameterisation of two relationships described that are likely to have a particular bearing on my main question; first, the climate sensitivity, and, second, the damage function.

The pdf of the climate sensitivity in most previous applications of PAGE has been triangular. In the Stern Review (Stern 2007), this triangular distribution was fitted on a minimum value of 1.5°C, a best estimate of 3°C, and a maximum value of 4.5°C. Yet in view of the scientific evidence collected by IPCC (Meehl, Stocker et al. 2007) and discussed above, this range looks to be too narrow. Indeed, the triangular distribution is *par excellence* an example of a thin-tailed pdf. Therefore I impute an alternative pdf for the climate sensitivity, fitted on the analysis of Stainforth et al. (2005), using the commercial software package @Risk to make an approximation. @Risk works by fitting a wide range of different kinds of probability distribution to the data and selecting the distribution with the best fit in terms of the lowest root-mean-square error. The resulting pdf in this case is log-logistic<sup>9</sup>, and is reproduced in figure one alongside the Stern Review pdf for comparison. The log-logistic distribution is polynomial in its tails, and in this case the cumulative density function is fitted such that

$$F(x) = \frac{1}{1 + (\frac{x - 2.32}{1.73})^{1.44}} \tag{11}$$

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<sup>&</sup>lt;sup>9</sup> i.e. the probability distribution of a random variable (the climate sensitivity) whose logarithm has a logistic distribution.

 $<sup>^{10}</sup>$  Other fat-tailed pdfs provide a reasonably good fit of the data too. For example, a lognormal distribution was fitted to the data with a root-mean-square (RMS) error of  $4.96*10^{-4}$ , which is 20% higher than the RMS error of  $4.15*10^{-4}$  for the log-logistic distribution. However, the results in section 5 change little when the former distribution is substituted for the latter (data available from the author on request). By contrast, thintailed pdfs such as the triangular distribution (RMS error =  $1.27*10^{-3}$ ) and normal distribution (RMS error =  $2.17*10^{-3}$ ) fit the data much less well, as one would expect.

where x is the climate sensitivity, from which it can easily be verified that the minimum is 1.4°C, the mode is 2.6°C, and the 99<sup>th</sup> percentile is 14.0°C (the log-logistic distribution does not of course have a finite upper support).

Stainforth et al.'s (2005) estimates of the climate sensitivity are one result from the huge *climateprediction.net* experiment, in which distributed computing resources were used to run the first 'grand ensemble' of simulations through a general circulation model of the climate system. A grand ensemble is essentially a Monte Carlo simulation, in which model parameters are varied (a so-called 'perturbed physics' ensemble), alongside initial climate conditions (e.g. initial temperatures and humidity levels) and carbon dioxide emissions or concentrations. This gives one of the most comprehensive available estimates of climate-model uncertainty about equilibrium temperature change, based on the largest ever set of model runs. Other pdfs reported by IPCC (Meehl, Stocker et al. 2007, box 10.2) could also be used, but the Stainforth et al. distribution is a good proxy of their typical features.

#### FIGURE ONE HERE

Another key element in the climatic response to greenhouse gas emissions is the carbon cycle, and in particular the risk of a large-scale positive feedback. This is captured in the basic structure of PAGE by a parameter representing natural stimulation of carbon dioxide from weakening terrestrial biospheric and oceanic carbon sinks, which is as usual triangular distributed. Future work could attempt to substitute this by an alternative pdf, perhaps including the likelihood of natural stimulation of methane from melting permafrost, from drying of wetlands, and even from the oceans. However, there is currently very little evidence upon which such a pdf could be based, compared with the climate sensitivity. Therefore the triangular distribution of natural carbon dioxide stimulation is retained.<sup>11</sup>

In estimating the relationship between warming and economic losses, section two argued that the key uncertainty can be captured in the curvature of a generalised damage function (the parameter

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<sup>&</sup>lt;sup>11</sup> In simulations where the atmospheric concentration of greenhouse gases is stabilised at 550 parts per million, the natural stimulation of carbon dioxide per degree Centigrade of temperature rise is halved relative to its baseline value (Hope, 2006a), following a suggestion by Chris Hope and based on research by van Vuuren et al. (2009).

 $\beta$  in equation 1). Once again, PAGE has in past studies been run with a triangular distribution for  $\beta$ , where the minimum value is unity, the best estimate is 1.3, and the maximum value is 3. Yet section two argued an intuitive case for exploring values higher than 3, as Weitzman (2009) has too. Thus @Risk is once again used to fit a revised pdf, but, given the lack of evidence on  $\beta$ , this time I make a minimal set of assumptions about damages. First I assume that if there is no warming there is no economic cost, all else equal. Second, I assume that the best estimate of  $\beta$  is 2, giving a quadratic damage function. This has the merit of replicating a very common assumption in the literature, although, as mentioned above, the assumption itself has little grounding except in analytical convenience. Third and finally, I make the arbitrary assumption that the 90th percentile is 3, so that there is a 10% likelihood of  $\beta$  exceeding 3. The best fitting distribution (figure two) is lognormal, with a minimum value of 0.6, a best estimate of 1.7 and a 99th percentile of 3.8 (the lognormal distribution does not have a finite upper support either).

#### FIGURE TWO HERE

Economic growth, population growth and greenhouse gas emissions are exogenous to the PAGE model. In this paper they are taken from the IPCC SRES A2 scenario (Nakicenovic and Swart 2000), converted to purchasing power parity and extrapolated from 2100 to 2200 by Hope (2006a). The Stern Review (Stern 2007) also used the A2 scenario, so it provides a point of contrast. The overall modelling horizon is 2000 to 2200. GDP from the A2 scenario is converted to consumption for estimation of welfare using a constant exogenous saving rate of 20%. Business-as-usual emissions are considered, alongside a scenario that sets out to stabilise the atmospheric concentration of greenhouse gases at 550 parts per million (ppm), although the carbon-cycle feedback results in an eventual concentration somewhat higher than 550 ppm. This provides estimates of the reduction in the risk of catastrophe brought about by emissions reductions.

In performing Monte Carlo simulations, a Latin Hypercube Sample is taken, with 1000 simulation draws. Latin Hypercube Sampling improves coverage of the full range of input values relative to a simple, non-stratified Monte Carlo procedure, an especially advantageous feature when investigating probability distributions with a large skew.

#### 5. Results

The first step in the empirical analysis is to quantify the probability of a high-impact event in the climate-economy system, a catastrophe. Since there is no canonical meaning of 'catastrophe' in economics (or indeed elsewhere), I simply investigate the relative frequency with which arbitrarily large economic costs occur. In particular, table one begins by estimating the relative frequency with which the global cost of climate change, at any point in time between 2000 and 2200, exceeds three thresholds, 75% of global baseline consumption, 90%, and 100%. In terms of the discussion above, the last threshold of 100% of baseline consumption is especially important, as its breaching will directly lead to infinite or zero social welfare, and infinite marginal utility for strictly positive relative risk aversion. Hence we will gain an initial insight into the relevance of the upper bound on damages,  $\phi$ .

Table one presents estimates of the probability of a catastrophe conditional on two sets of assumptions about the climate sensitivity and the damage-function exponent ( $\beta$ ). The first set comprises the triangular pdf for the climate sensitivity and for  $\beta$  from the Stern Review (Stern, 2007: labelled 'Stern' for convenience). The second set comprises the log-logistic pdf for the climate sensitivity fitted on data from Stainforth et al. (2005), and the lognormal pdf for  $\beta$  outlined above. Estimates are presented both for business-as-usual emissions and for the stabilisation scenario.

#### TABLE ONE HERE

Looking first at business-as-usual emissions, the probability that the global cost of climate change exceeds 75% of baseline global consumption is 0.1% (i.e. 1 in 1000 simulation draws), assuming the Stern Review's triangular pdfs. The probability of costs greater than 90% of consumption, and in turn greater than 100% of consumption, is zero, in terms of zero draws from the Latin Hypercube Sample. However, for the more plausible assumption that the climate sensitivity is log-logistically distributed, while  $\beta$  is lognormally distributed, the probability of a global climate catastrophe is positive across all three thresholds. The probability of costs greater than 75% of global consumption is 7.2%, while the probability of costs greater than 90% of consumption is 6.2% and the probability of costs reaching 100% of consumption is 5.9%. This is worrying, of course. It also gives immediate empirical backing to the relevance of the issues raised by Weitzman (2007; 2009), including  $\phi$ .

Turning to the stabilisation scenario, we can see that emissions reductions buy down the probability of catastrophe. With the 'Stern' climate sensitivity and damage-function exponent, damages in the sample never exceed the three threshold levels of consumption. On the other hand, with the fat-tailed distributions, the probability of catastrophe is not eliminated in the sample. Global damages exceed the 75% threshold in 2% of draws, the 90% threshold in 1.7% of draws and reach the 100% threshold in 1.3% of draws.

In fact, the estimates in table one may still underplay the importance of high-impact events, because they may not pick up the existence of a regional catastrophe, where climate damage exceeds 75%, 90%, or reaches 100% of regional baseline consumption, but net consumption in other regions is sufficient to compensate. Therefore table two takes the same set of simulation draws and this time counts every draw in which there is a catastrophe in any one or more of PAGE's eight major world-regions. <sup>12</sup>

#### TABLE TWO HERE

As expected, the probability of a regional catastrophe is higher than that of a global catastrophe. Under business-as-usual and assuming the 'Stern' climate sensitivity and damage-function exponent, the probability that costs exceed 75% of regional consumption in any one or more regions is 3.8%, while there is a 2.8% probability of costs exceeding 90% of consumption, and a 2.3% chance of costs reaching 100% of consumption. Substituting the more plausible, fat-tailed distributions for warming and damages, the probabilities rise to an alarming 17.8%, 15.6% and 14.4% respectively. The regions most likely to suffer a catastrophe are India and Southeast Asia, Africa and the Middle East, and Latin America, although the only region to escape a catastrophe altogether is the former Soviet Union and Eastern Europe. The risk of catastrophe is highest in the period from 2150 to 2200, although, in the scenario with fat-tailed distributions, there is a 1.8% probability of costs reaching 100% of consumption before 2100. Once again, stabilising

<sup>&</sup>lt;sup>12</sup> (i) Western Europe (i.e. the old EU-15); (ii) the former Soviet Union and Eastern Europe; (iii) the USA; (iv) China and centrally planned Asia; (v) India and Southeast Asia; (vi) Africa and the Middle East; (vii) Latin America; (viii) other OECD.

<sup>&</sup>lt;sup>13</sup> Clearly this depends on the assumptions of the model. Indeed, in Tol (2003), the FUND model generated its only regional catastrophe in the former Soviet Union.

<sup>&</sup>lt;sup>14</sup> Further data on the incidence of a 'catastrophe' by region and/or by time period can be obtained from the author on request. PAGE has ten uneven time steps: 2000; 2001; 2002; 2010; 2020; 2040; 2060; 2080; 2100; 2150; 2200.

greenhouse gas concentrations reduces the risk of a catastrophe, but assuming the fat-tailed distributions, it remains considerable.

It is interesting to briefly digress by examining what values the climate sensitivity and the damage-function exponent ( $\beta$ ) have to realise for a catastrophe to occur. There are various ways to interrogate the underlying data, but one can readily confirm the intuition that higher values of both parameters make catastrophes more likely. Figure 3 illustrates this, plotting  $p[d_t \geq \hat{c}_t], t \in [0,\tau]$  as a function of the climate sensitivity (top panel) and  $\beta$  (bottom panel) respectively, looking at the aggregate, global level (and under business-as-usual emissions). If the climate sensitivity is less than 4.5°C, then the probability of catastrophe is just 1.3%, rising over threefold to 4.3% for climate sensitivities up to 10°C. Similarly, if  $\beta$ <2, the probability of catastrophe is just 1.2%, whereas it rises nearly fourfold to 4.2% for  $\beta$ <3 and to 5.6% for  $\beta$ <4.15

#### FIGURE THREE HERE

Given the risk of a catastrophe, tables one and two clearly show that the introduction of an upper bound on climate damages,  $\phi$ , could be important. Section three also speculated that the precise value of  $\phi$  may be important, even to within a very narrow range in the region of unity. Figure four thus presents estimates of the variation in the welfare cost of climate change, measured as the expected change in social welfare  $E(\Delta V)$  normalised to present GDP per capita<sup>16</sup>, for a range of values of  $\phi$  between 0.75 and very close to unity. As we are now considering the sensitivity of social welfare (equation 3), three curves are plotted, for different combinations of the utility discount rate,  $\delta$ , and the elasticity of the marginal utility of consumption,  $\eta$ . The climate-sensitivity pdf is from Stainforth et al. (2005), while  $\beta$  is lognormally distributed.

#### FIGURE FOUR HERE

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<sup>&</sup>lt;sup>15</sup> The minimum climate sensitivity required for a 100% loss of global consumption is 2.8°C, but in this single outlying case  $\beta$ =4.4 and other uncertain climate parameters such as natural stimulation of CO<sub>2</sub> dictate that warming reaches nearly 9°C by 2200. Symmetrically, the lowest value of  $\beta$  required to generate a catastrophe is just 1.3, but in this similarly outlying case the climate sensitivity is 29.7°C. The complete dataset is available from the author on request.

<sup>&</sup>lt;sup>16</sup> Formally, the change in the balanced growth equivalent (BGE), as used by the Stern Review (Stern, 2007) and fully elaborated in Anthoff and Tol (2009).

The bottom-most curve traces the relationship between the expected loss in social welfare and  $\phi$ , for the Stern Review's (Stern 2007) disputed assumptions that  $\eta$ =1 and  $\delta$ =0.1% per annum. The shape of the curve is striking, showing not only that the welfare cost of climate change is sensitive to  $\phi$ , increasing from 11.7% of present GDP per capita when  $\phi$  is 0.75 to 19.5% when  $\phi$  is very close to unity, but also that  $E(\Delta V)$  is increasing more than proportionately in  $\phi$ . This is a direct consequence of CRRA utility allied with positive risk aversion: as  $\lim \phi \to 1$ , the marginal change in the welfare cost with respect to  $\phi$  tends to infinity. Indeed, the curve looks like the mirror-image of a CRRA utility function plotted against rising consumption per capita. Thus, as we might have anticipated, the analysis is particularly sensitive to very small increases in the value of  $\phi$  in the region of (strictly less than) unity.

However, just as important is to observe the cluster of observations as  $\phi$  tends in the limit to unity. What this shows is that, while the incremental expected loss in social welfare increases, the incremental increase in  $\phi$  itself becomes vanishingly small. Indeed, the latter effect exceeds the former, so that the welfare cost of climate change asymptotes to a particular value, in this case roughly 20% of present GDP per capita. This reflects the third postulate in section three: for any given  $\phi$ <1, the welfare cost of climate change is an arbitrarily small function of  $\delta$  and  $\eta$ . Here, the Latin Hypercube Sample gives a 5.9% probability that climate damages reach 100% of global consumption (table one). As  $\phi$  tends in the limit to unity, the expected loss in social welfare will tend to 100% for these 59 in 1000 simulation draws. But for the remaining 94.1% of draws, climate damages are below  $\phi$ , and hence the increase in  $\phi$  is irrelevant. It is in this sense that the welfare cost of climate change is bounded from above by the probability of a catastrophe.

Turning now to the other two sets of assumptions about  $\delta$  and  $\eta$ , the top-most curve plots the relationship for  $\delta$ =1.5%.  $\eta$  remains at unity. What can be seen is that increasing the utility discount rate lowers the welfare cost of climate change for all values of  $\phi$  (postulate three again). Indeed, the difference is substantial. Setting  $\phi$  to, for example, 0.99, the welfare cost of climate change roughly halves from 15.5% of present GDP per capita to just 8.4% as  $\delta$  is increased from 0.1% to 1.5%. This is unsurprising; the importance of the utility discount rate to economic analysis of climate change is well known. What can also be seen is the same shape of relationship between the expected loss in social welfare and  $\phi$ , namely a more than proportionate increase in cost as  $\phi$  is increased.

The middle curve plots the relationship for  $\eta$ =3, while holding  $\delta$  at 0.1%. Again, the same relationship is traced between the expected loss in social welfare and  $\phi$ . And, again, the expected loss in social welfare is reduced for all values of  $\phi$ , when  $\eta$  is increased. Setting  $\phi$  to 0.99, the welfare cost of climate change falls from 15.5% of present GDP per capita when  $\eta$ =1 to 15.0% when  $\eta$ =3. This could not have been confirmed *a priori*, because in the standard welfare model  $\eta$  simultaneously captures aversion to inequality in consumption over regions and over time, and relative risk aversion. With increasing  $\eta$ , more weight is placed on consumption losses in poor regions, and more weight is placed on the tail risks. Since poor regions are expected to suffer the worst relative impacts of climate change, and PAGE represents this in its parameterisation of  $\alpha_i$  (Hope 2006a), both of these would tend to increase the expected loss in social welfare. However, assuming (net) consumption growth, an increase in  $\eta$  also effectively increases the social discount rate. Thus what figure three indicates is that the temporal effect outweighs the spatial and risk effects in this particular instance.

What figure four also shows is the relative difference in the welfare cost of climate change as  $\phi$  is varied. Consider first the utility discount rate,  $\delta$ . We can see from inspection of figure four that the difference in the expected loss in social welfare between  $\delta$ =0.1% and  $\delta$ =1.5% actually *increases* in  $\phi$ . That is to say, the effect of the utility discount rate on the welfare cost of climate change rises as the upper bound on climate damages rises. This is actually quite intuitive: increasing the impact of these events in the extreme tail, which are generally far off, increases the relevance of  $\delta$ , and higher  $\delta$  reduces the weight placed on these impacts. The expected loss in social welfare has a different asymptote when  $\delta$ =0.1% (about 20% of present GDP per capita) compared with  $\delta$ =1.5% (about 13% of present GDP per capita), since the probability that  $d_{j,i} \geq \hat{c}_{j,i}$  is quite low, and so the loss of social welfare in states of nature where  $d_{j,i} < \phi \hat{c}_{j,i}$ , which is a function of  $\delta$ , determines the asymptote. To reiterate, for any given  $\phi$ <1, the welfare cost of climate change is an arbitrarily small function of  $\delta$ .

Turning to the elasticity of the marginal utility of consumption,  $\eta$ , figure four shows that as  $\phi$  increases, the difference in the welfare cost of climate change initially *decreases*. As discussed above, by the time  $\phi$ =0.99 the difference is only 0.5 percentage points. But as  $\phi$  becomes very close to unity, the difference begins to increase again. What is happening is that, with higher  $\eta$ , the expected loss in social welfare reaches its asymptote more quickly, as more weight is placed on

catastrophic outcomes with very low net consumption. This initially serves to reduce the difference in  $E(\Delta V)$  due to  $\eta$ , but then serves to increase it again (when  $\eta$ =3, the asymptote is reached by the time  $\phi$ =0.999, whereas when  $\eta$ =1 it takes longer).

Finally, table four gives estimates of the social cost of carbon (specifically the marginal damage cost of carbon dioxide) accompanying these experiments. The estimates are conditional on different values of the disputed welfare parameters  $\delta$  and  $\eta$ , comparing the estimate for business-as-usual emissions with the stabilisation scenario.  $\phi$  is set to 0.99. The striking feature is of course that the estimates are very high, conditional on assumptions about  $\delta$  and  $\eta$ . For the Stern Review's (Stern, 2007) assumptions that  $\eta$ =1 and  $\delta$ =0.1%, the mean social cost of carbon along business as usual is a dizzying US\$444.65/tCO2, within a 90% confidence interval of over US\$1800/tCO2. For the stabilisation scenario, the equivalent mean price is US\$377.61/tCO2. However, when  $\eta$ =3 and  $\delta$ =1.5%, the estimates are lower. Thus the table makes two points. First, replacing thin-tailed pdfs for the climate sensitivity and for the curvature of the damage function with more plausible fattailed pdfs, the social cost of carbon rises significantly. Second, the table demonstrates the ongoing importance of assumptions about social discounting, even when catastrophic climate change is possible.

#### TABLE FOUR HERE

#### 6. Discussion and conclusions

To what extent does economic analysis of climate change depend on low-probability, high-impact events? The short answer is a great deal (see also Ackerman, Stanton et al. 2009), but not to the exclusion of other factors that we already know to be very important, in particular the discount rate. Inputting fat-tailed distributions for two key parameters in the analysis, namely the climate sensitivity and the curvature of the damage function, I find that a catastrophe is possible in the integrated assessment model PAGE, both under business-as-usual emissions and even stabilising the atmospheric stock of greenhouse gases at nearly 550 ppm. This brings to the fore the issues recently raised by Martin Weitzman (2007; 2009).

In order to then undertake welfare analysis, it is necessary to bound the catastrophe in some way. Failure to do so in the standard welfare model with risk aversion implies we are willing to pay an infinite amount at the margin to reduce greenhouse gas emissions, while total (as opposed to marginal) willingness to pay is also unbounded for a plausible assumption about the degree of risk aversion (i.e.  $\eta \ge 1$ ). This makes it impossible to come to useful conclusions about, for example, optimal rates of carbon taxation. This paper has explored one way to bound the catastrophe, introducing an upper bound on climate damages that could in principle be derived from the concept of the (finite) value of statistical life, albeit essentially scaled up to the level of whole societies, or even to the ultimate level of human life itself (Weitzman 2009). While such an approach imposes an *ex ante* limit on a decision-maker's willingness to pay to avoid zero consumption, an alternative approach could take the upper bound to represent some real limit to the value of climate damage in proportion to baseline consumption. Unsurprisingly, the valuation of climate impacts turns out to be sensitive to this upper bound, in particular to very small differences in the upper bound towards its limiting value. This is concerning, since there is very little evidence on which to base it.

Nevertheless, if you subscribe to three assumptions made in this paper, there is cause to believe that the importance of catastrophes is bounded, and long-standing controversies over the discount rate continue to matter. The first assumption is that there are *a priori* grounds to 'cap' the total impact of the catastrophe (e.g. in reasoning by analogy to the value of statistical life). This bounds the importance of catastrophes, but is not enough to make controversies over the discount rate material. For them to be material, it is additionally necessary to assume (call this assumption two) that the catastrophe will not occur for decades to come, and that (assumption three) the probability of a catastrophe is, roughly speaking, small. If all three assumptions hold, the welfare cost of climate change will asymptote to some value much less than the total value of the global economy, and moreover the two parameters of the social discount rate, the utility discount rate and the elasticity of the marginal utility of consumption, will be strong determinants of where that asymptote lies.

Besides interpretation of the results, this paper raises a number of open questions and issues for further research. Immediately, there will be suggestions to tweak the empirical approach, trying for instance an alternative probability distribution for the climate sensitivity, and in particular introducing a fat-tailed distribution for the carbon-cycle feedback, or for the curvature of the damage function. These would of course be interesting extensions. They are likely to be important in taking very much that is quantitative from the paper, but on the other hand would be unlikely

in my view to qualitatively alter its main conclusions, unless for some reason a catastrophe can be ruled out through these changes (or through a change in the functional form of the damage function).

Instead, I would like to draw attention to two issues. The first is the interpretation of the upper bound on damages, and indeed alternative ways in which the problem could be bounded. More rigorous thinking on these issues is surely warranted, given the sensitivity of the analysis to where (and quite probably how) it is bounded. As mentioned, Weitzman (2009) draws an analogy with the value of statistical life, since individuals make trade-offs on a daily basis between decreasing present consumption and reducing the probability of their own death, at least tacitly. This, he shows with some ballpark estimates, might lead to an upper bound of 99% of welfare-equivalent consumption. Unfortunately, the analysis in this paper has showed that it can matter whether it is 99%, or 99.9%, and so on.

Nordhaus (2009), on the other hand, reasons by analogy with the actions of public institutions towards mitigating tiny risks of other catastrophes, such as the collision of a very large asteroid with the Earth, or unwanted, catastrophic consequences of research into biotechnology and nanotechnology. We spend just a few million dollars per year researching asteroids, while research into bio- and nanotechnology continues, despite the risks. This, it seems to me, is a promising area for further research, because the decision context – i.e. public institutions governing societal risks – is more directly comparable to the problem of mitigating catastrophic risks from climate change. We do not, however, have to choose. We need more research on both. And there are still other ways forward. For one, the problems discussed in this paper are in large part a consequence of the reliance on a particular form of utility function. Other utility functions do not possess the same, troublesome limiting properties as the CRRA utility function.

A quite different approach would be to rule out zero net consumption *a priori*, for biophysical and/or socio-economic reasons. One recent example of such an approach is Costello et al. (2010), who argue for placing an upper bound on future temperature change, which when fed into a simple economic model produces finite total willingness to pay ( $\eta \ge 1$ ) to avoid climate damage. A

<sup>&</sup>lt;sup>17</sup> The value of statistical life can be seen as a specific application of the more general concept of 'fear of ruin', introduced by Aumann and Kurz (1977). The connection between the two concepts is elaborated in Foncel and Treich (2005).

drawback of their analysis, however, is that they do not give consideration to very aggressive damage functions, so that even extremely large amounts of warming (up to 50°C above preindustrial!) still give rise to fairly modest damages. Put another way, even if we could place a constraint on warming, we could not rule out zero net consumption without also ruling out a catastrophic social and economic reaction to warming below what is judged to be the maximum temperature. And it seems likely that a severe negative social and economic reaction would kick in for amounts of warming significantly less than 50°C above pre-industrial. A different example is Yohe (2003), who has argued that, at the regional level, the upper bound can be interpreted as the receipt of aid in the affected region, from other regions, to maintain living standards above subsistence levels.

The second issue to highlight is the analysis of probabilities. This is a paper about risk, in the sense that it is assumed we know the mean and the variance of the joint probability distribution of climate impacts. Weitzman (2009) also considers this distribution, but his paper concerns how structural uncertainty about the mean and variance propagates through to give a 'fat-tailed' distribution of impacts. The question Heal (2008) raises is whether, if we cannot describe the impacts of climate change probabilistically, we should be working within standard expected-utility analysis at all? There are reasonably close alternatives within economics, in the shape of, for example, ambiguity theory (Henry and Henry 2002). And of course analysis of climate change from outside the economics profession has long applied notions of policies that are 'robust' in some sense to various, non-probabilistic scenarios. While such approaches have their own fair share of weaknesses, this seems to be a potentially rich vein of enquiry as well.

#### Acknowledgements

I am very grateful to Chris Hope for enabling me to use the PAGE model and for invaluable advice. I would also like to thank Frank Ackerman, Alex Bowen, Christian Gollier, Robert Hahn, Cameron Hepburn, Antony Millner, Nicola Ranger, Leonard Smith, Nicholas Stern, Martin Weitzman and Dimitri Zenghelis for comments on earlier drafts, to Stephen Schneider, the late editor of *Climatic Change*, and to two anonymous referees, and finally to participants in the session "The Role Of Integrated Assessment Models In Handling Climate Change Part I" at the IARU International Scientific Congress in Copenhagen on 11th March 2009. This research has been supported by the Grantham Foundation for the Protection of the Environment, as well as the

Centre for Climate Change Economics and Policy, which is funded by the UK's Economic and Social Research Council (ESRC) and by Munich Re. All errors are my own.

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Figure I. Climate sensitivities compared; Stern Review on the left, Stainforth et al. on the right.

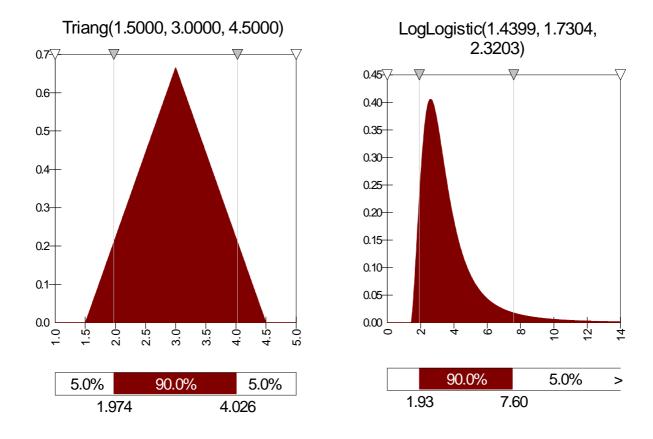


Figure II. The damage-function exponent ( $\beta$ ) compared; standard PAGE on the left, fat-tailed distribution on the right.

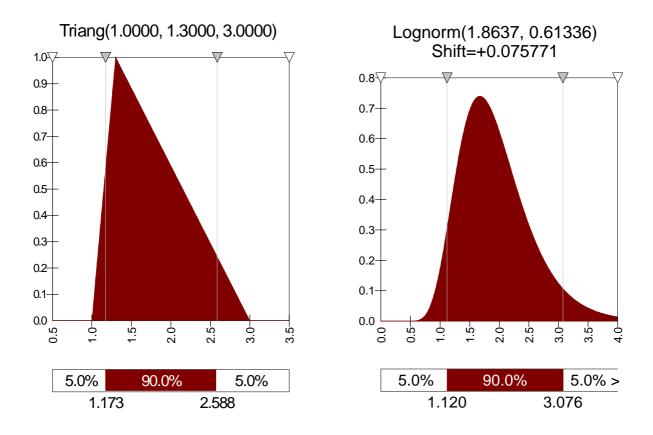


Figure III. The probability that damages are at least as great as baseline global consumption, given particular values of the climate sensitivity (top) and the damage-function exponent (bottom).

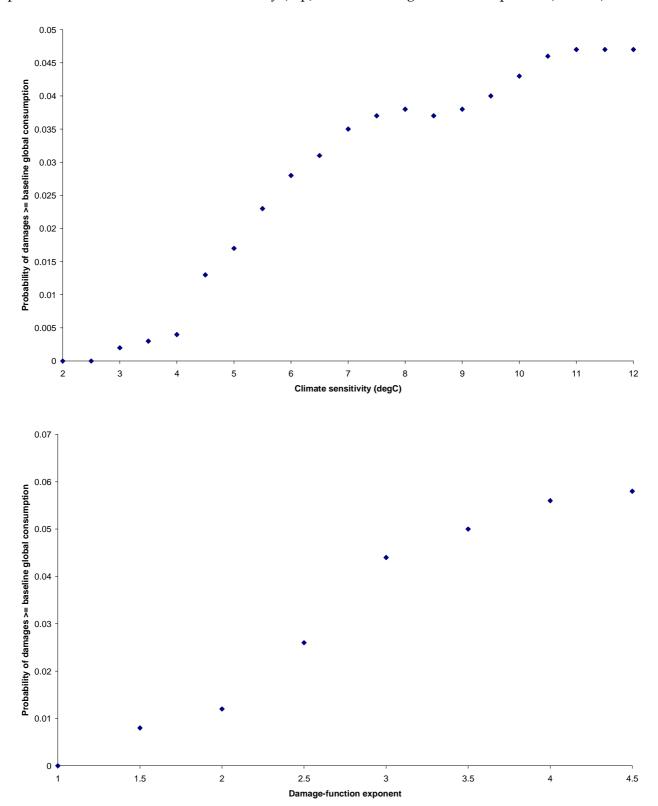


Figure IV. Expected loss in social welfare as a function of the upper bound on damages ( $\phi$ ).

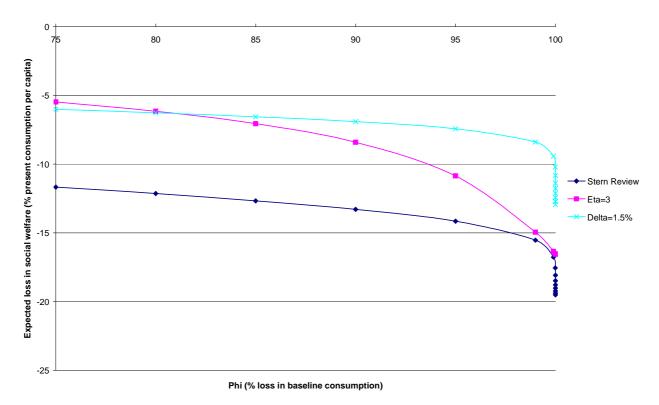


Table I. Probability of a global 'catastrophe'.

| Climate sensitivity                   | Damages   | $p\left[d_{j,t} \ge 0.75\hat{c}_{j,t}\right]$ | $p\left[d_{j,t} \ge 0.9\hat{c}_{j,t}\right]$ | $p\left[d_{j,t} \geq \hat{c}_{j,t}\right]$ |  |  |
|---------------------------------------|-----------|---|--|--|--|--|
| Business-as-usual emissions (IPCC A2) |           |   |  |  |  |  |
| Stern                                 | Stern     | 0.1% (i.e. 1 in 1000)                         | 0%   | 0%   |  |  |
| Stainforth et al.                     | Lognormal | 7.2%  | 6.2%   | 5.9%                                       |  |  |
| 550 ppm stabilisation (IPCC A2)       |           |   |  |  |  |  |
| Stern                                 | Stern     | 0% (i.e. 0 in 1000)                           | 0%   | 0%   |  |  |
| Stainforth et al.                     | Lognormal | 2.0%  | 1.7%   | 1.3%                                       |  |  |

Table II. Probability of a regional 'catastrophe'.

| Climate sensitivity                   | Damages   | $p\left[d_{j,t} \ge 0.75\hat{c}_{j,t}\right]$ | $p\left[d_{j,t} \ge 0.9\hat{c}_{j,t}\right]$ | $p\left[d_{j,t} \geq \hat{c}_{j,t}\right]$ |  |  |  |
|---------------------------------------|-----------|---|--|--|--|--|--|
| Business-as-usual emissions (IPCC A2) |           |   |  |  |  |  |  |
| Stern                                 | Stern     | 3.8% (i.e. 38 in<br>1000)                     | 2.8%   | 2.3%                                       |  |  |  |
| Stainforth et al.                     | Lognormal | 17.8%   | 15.6%  | 14.4%                                      |  |  |  |
| 550 ppm stabilisation (IPCC A2)       |           |   |  |  |  |  |  |
| Stern                                 | Stern     | 0% (i.e. 0 in 1000)                           | 0%   | 0%   |  |  |  |
| Stainforth et al.                     | Lognormal | 5.9%  | 4.7%   | 4.1%                                       |  |  |  |

Table III. The marginal damage cost of carbon dioxide ( $\phi$  =0.99).

| Utility discount                      | Elasticity of marginal         | Marginal damage cost (year 2008 \$US/tCO <sub>2</sub> ) |        |         |  |  |  |
|---------------------------------------|--------------------------------|---|--------|---------|--|--|--|
| rate, $\delta$                        | utility of consumption, $\eta$ | 5%  | mean   | 95%     |  |  |  |
| Business-as-usual emissions (IPCC A2) |                                |   |        |         |  |  |  |
| 0.1                                   | 1                              | 21.13   | 444.65 | 1861.64 |  |  |  |
| 1.5                                   | 3                              | 4.67  | 346.21 | 1358.54 |  |  |  |
| 550 ppm stabilisation (IPCC A2)       |                                |   |        |         |  |  |  |
| 0.1                                   | 1                              | 26.70   | 377.61 | 1740.83 |  |  |  |
| 1.5                                   | 3                              | 4.49  | 224.41 | 335.10  |  |  |  |