

# New approaches to decision-making under uncertainty in the economics of climate change

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# Examples of economic decisions that depend on climate predictions

- Investment in flood defence and other means of adapting to climate change
- Investment in weather-sensitive methods of energy supply, e.g. wind and nuclear
- Growing new markets for insurance against weather
- National and global greenhouse gas emissions reductions targets

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# In the beginning...

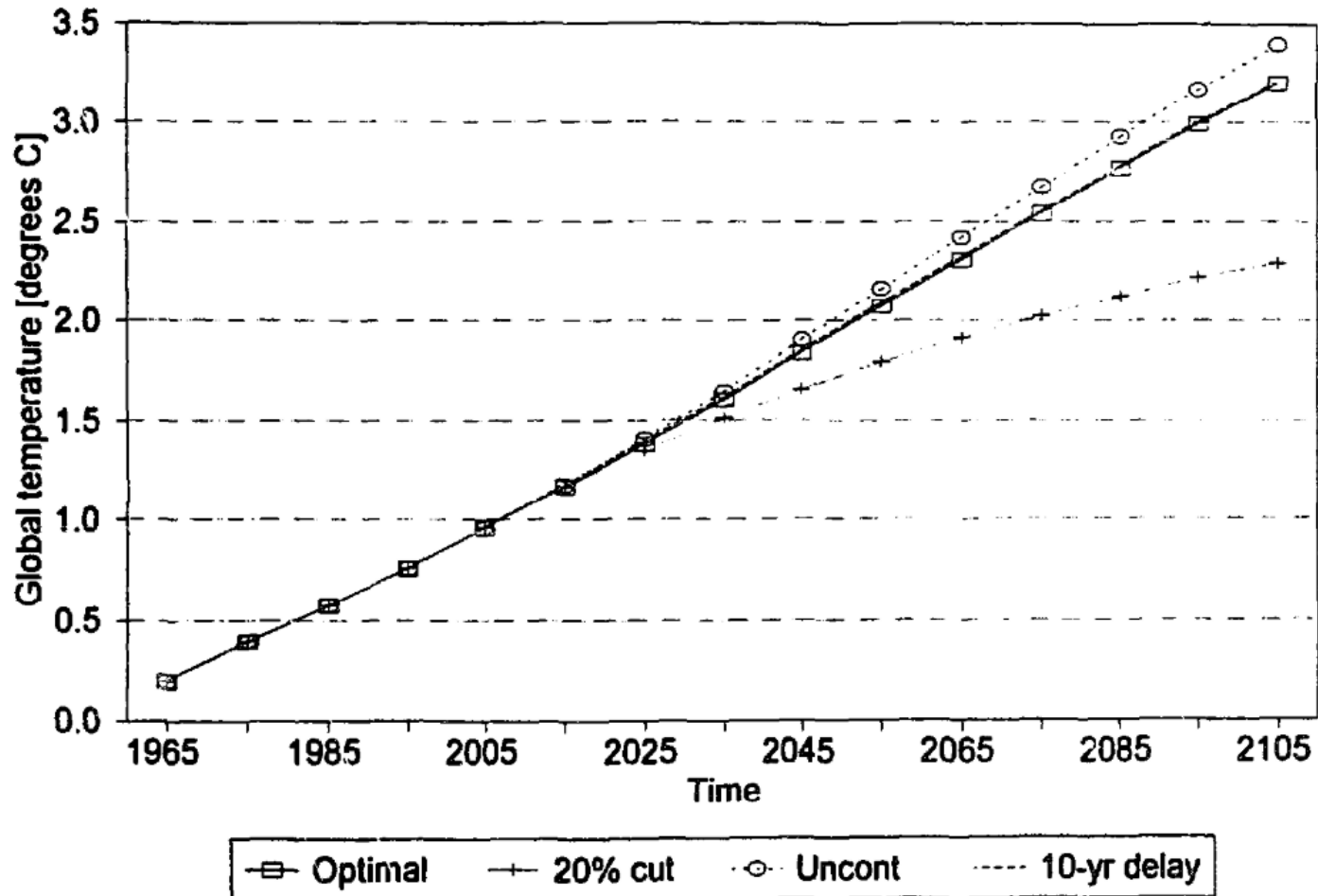
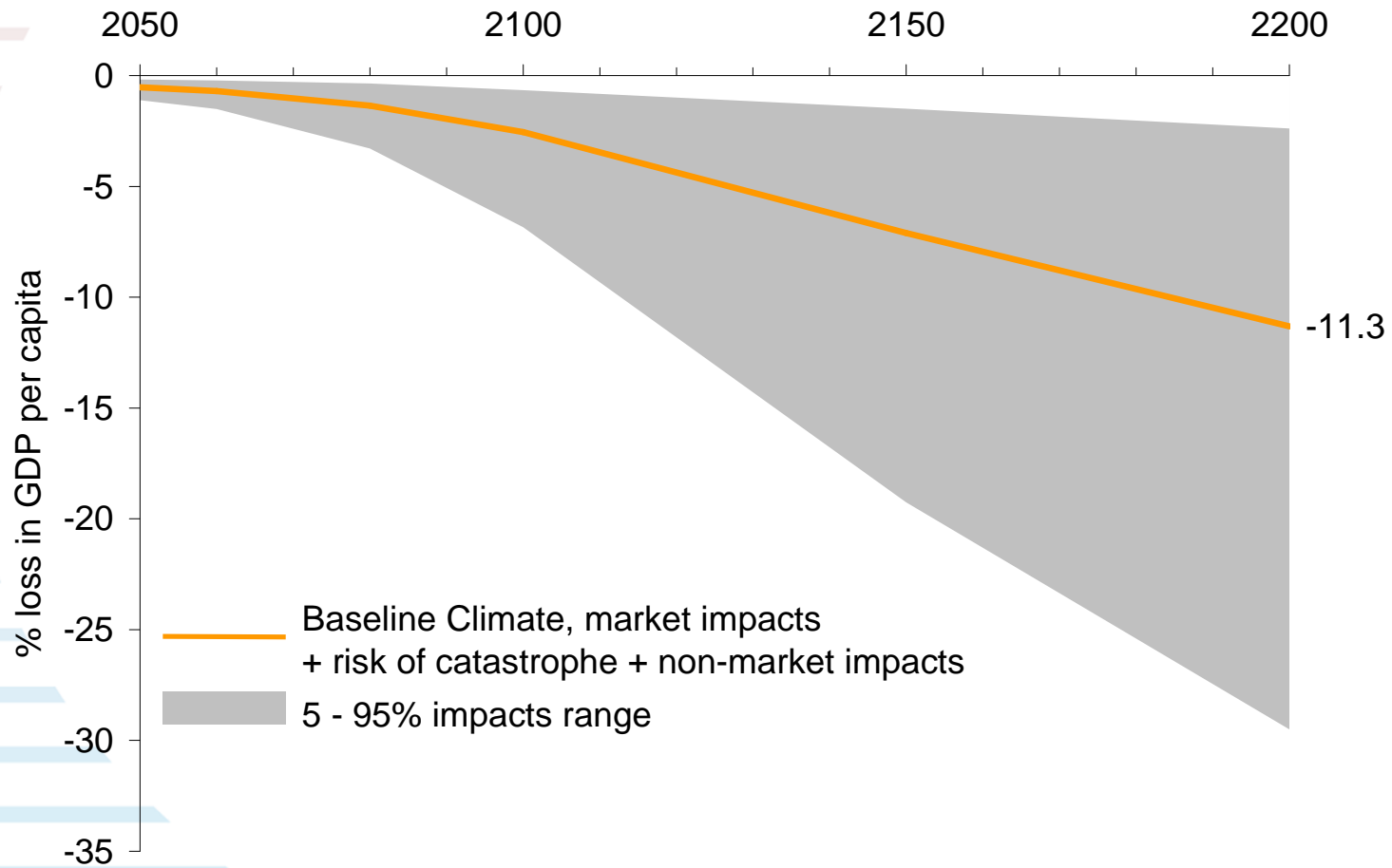


Fig. 4. Global mean temperature ( $^{\circ}\text{C}$ , difference from 1860).

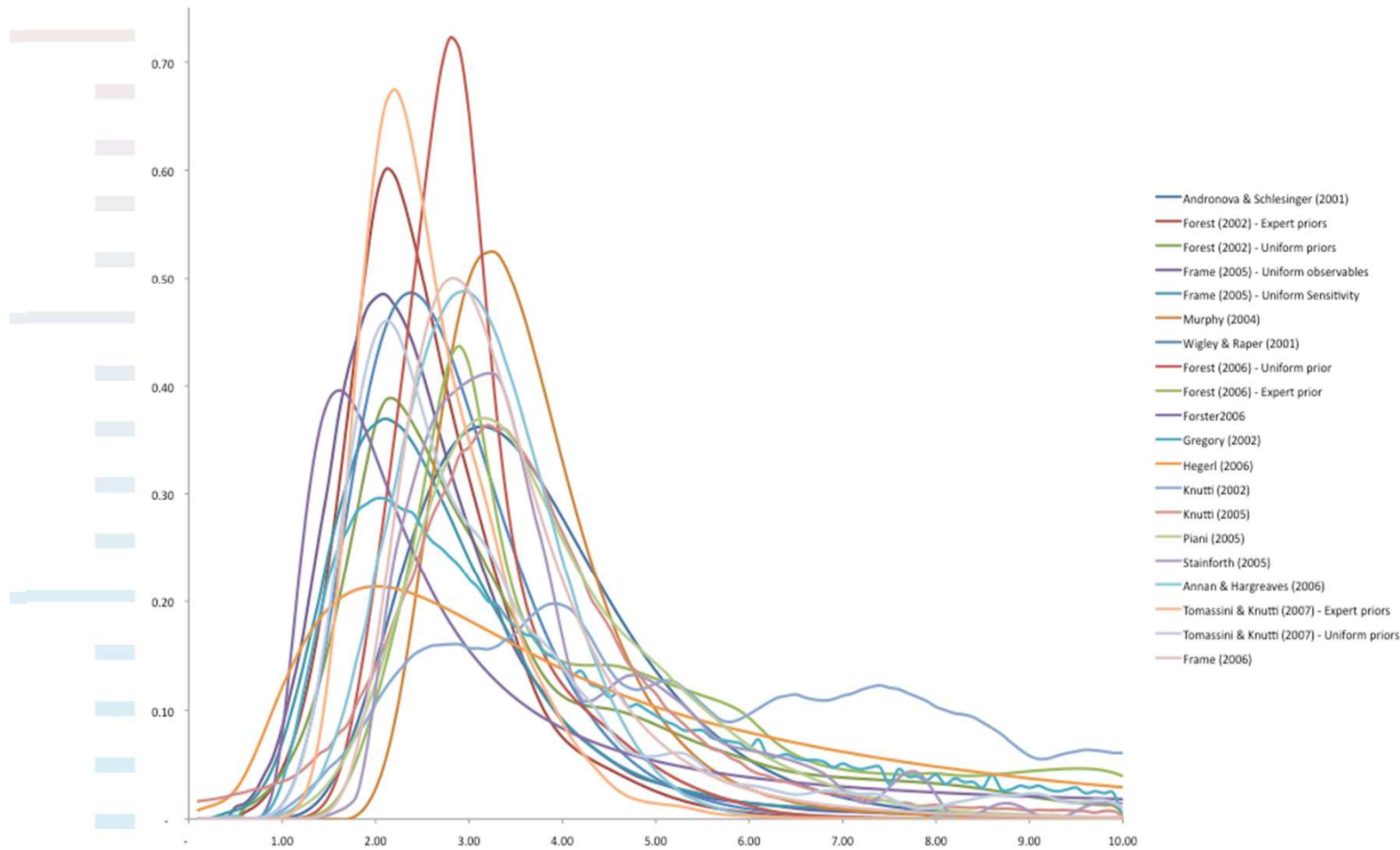
Source: Nordhaus (1993)

# Probabilities and expected utility



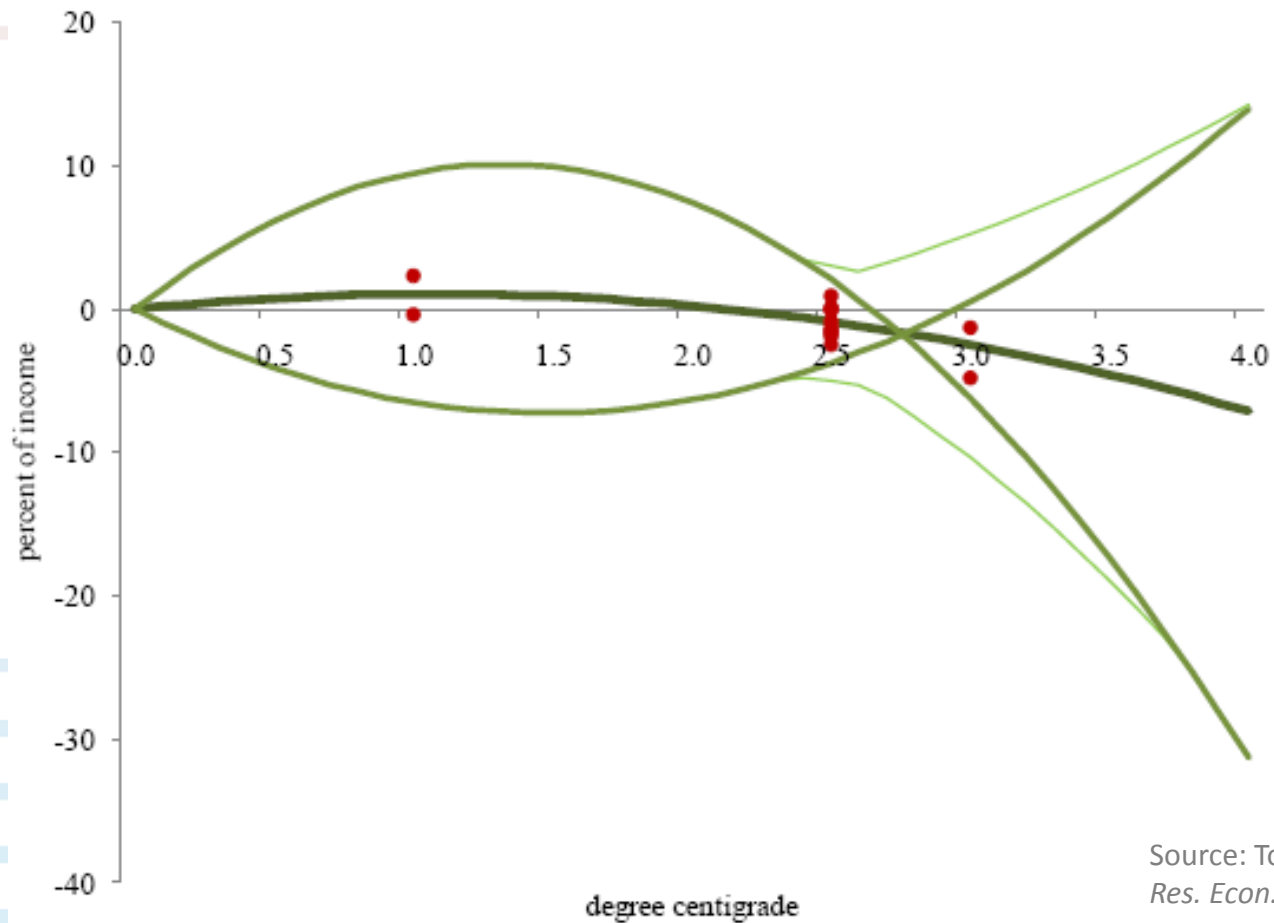
Source: Stern (2007)

# But are our climate predictions of sufficient quality to sustain expected utility approaches?



Source: Malte Meinshausen

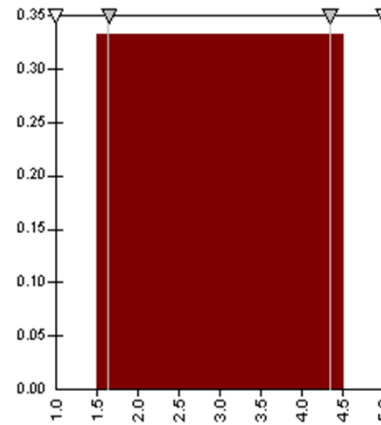
# Never mind about climate change, what about our predictions of its economic consequences?



Source: Tol (2012) in *Env. & Res. Econ.*

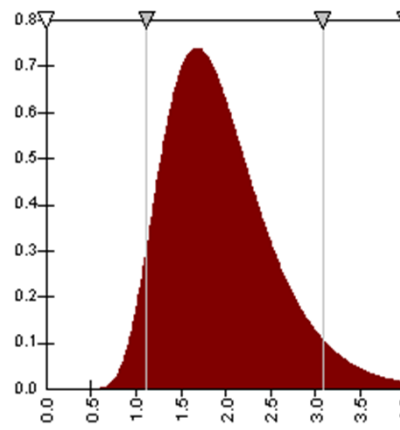
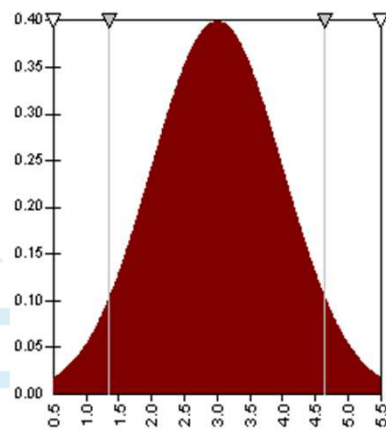
# An alternative: the smooth model of decision making under ambiguity

2<sup>nd</sup> order probabilities

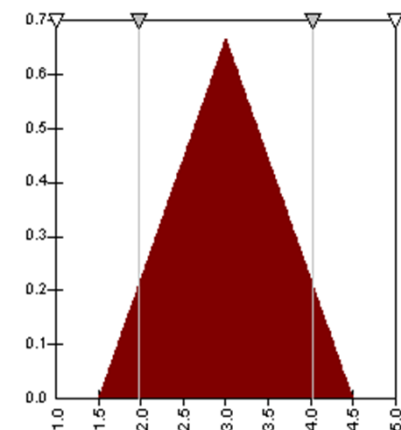


$$V(f) = \sum_m p_m \varphi(EU_m(f))$$

1<sup>st</sup> order probabilities

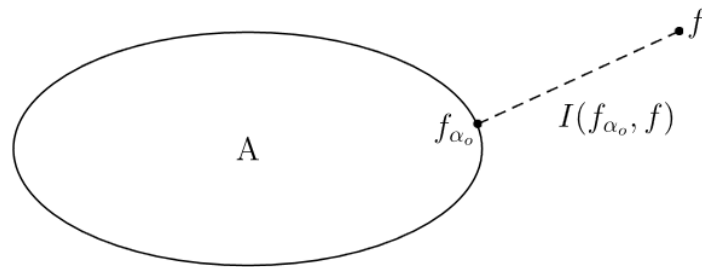


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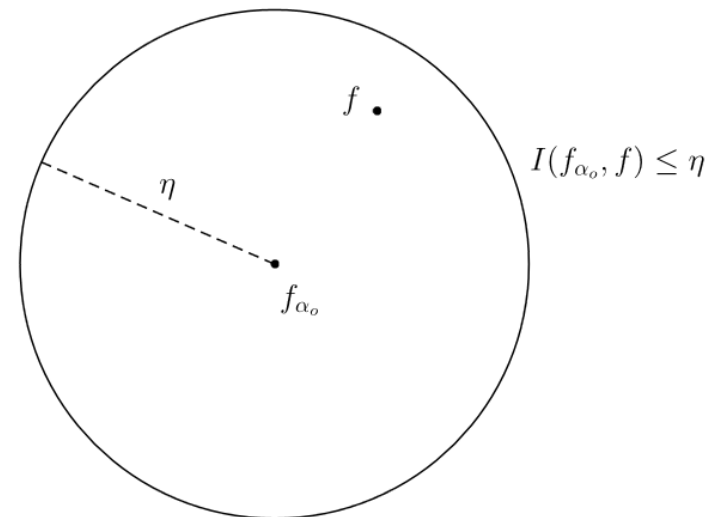




# Another alternative: robust control



**Figure 1.6.1:** Econometric specification analysis. Suppose that the data generating mechanism is  $f$  and that the econometrician fits a parametric class of models  $f_\alpha \in A$  to the data and that  $f \notin A$ . Maximum likelihood estimates of  $\alpha$  eventually select the misspecified model  $f_{\alpha_o}$  that is closest to  $f$  as measured by entropy  $I(f_\alpha, f)$ .



**Figure 1.7.1:** Robust decision making: A decision maker with model  $f_{\alpha_o}$  suspects that the data are actually generated by a nearby model  $f$ , where  $I(f_{\alpha_o}, f) \leq \eta$ .

$$\max V(a) = \min_{f_\alpha} E_{f_\alpha} U(a)$$

# Some concluding thoughts

- Recent developments have enabled research into the economics of climate change to be more realistic and sophisticated about uncertainty
  - But realistic and sophisticated *enough*?
- Policy relevance: general or specific?
  - Confirming the precautionary motive behind emissions targets
  - Some specific lessons for adaptation, e.g. real options story tends to warn against hasty investment in climate-proofing infrastructure