

I've been asked to talk  
“for about 7 minutes” on

# Risks Analysis of Emerging Technologies with High Uncertainty

and also about

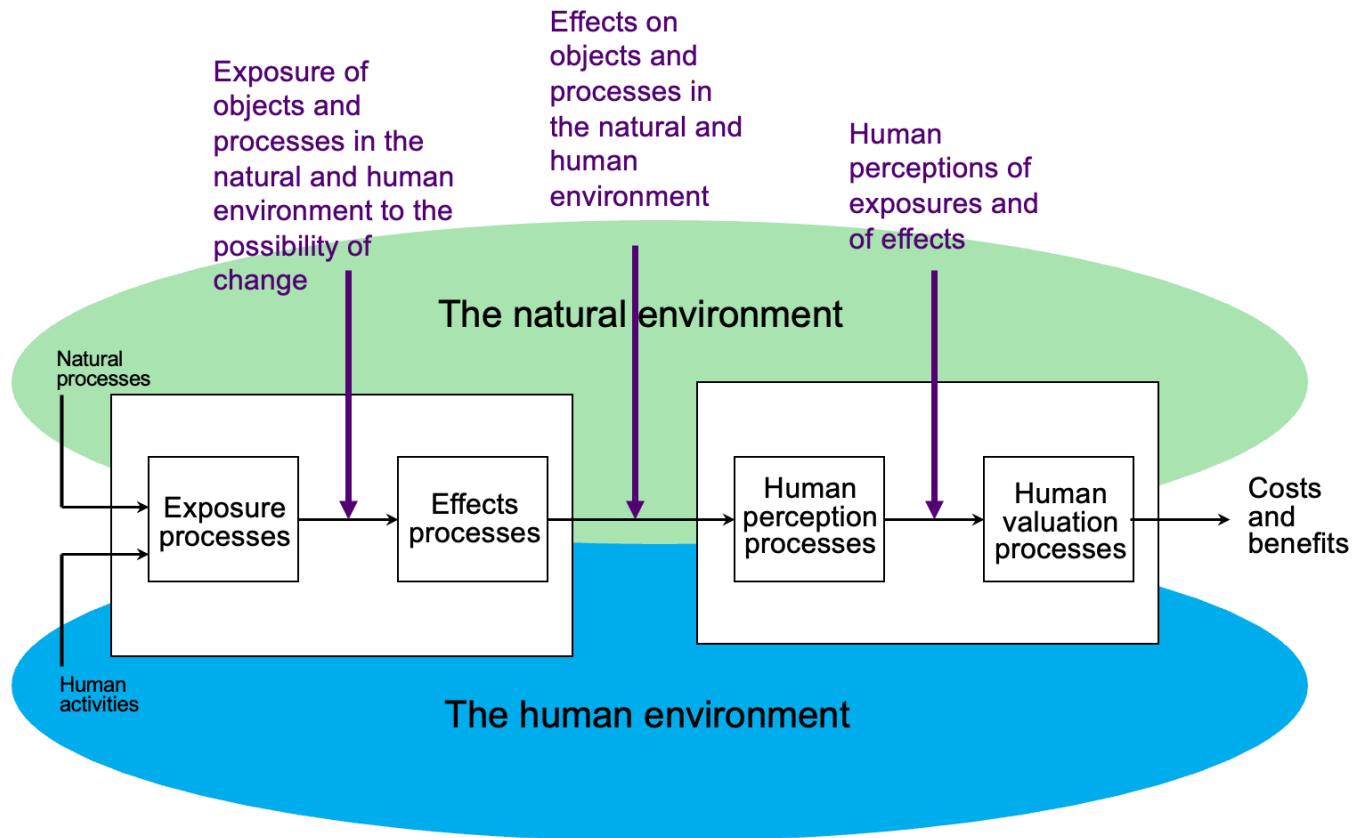
## Overconfidence

and to do it with 5 slides.

I'll stick to the 7-minute time constraint but violate the 5-slide constraint.

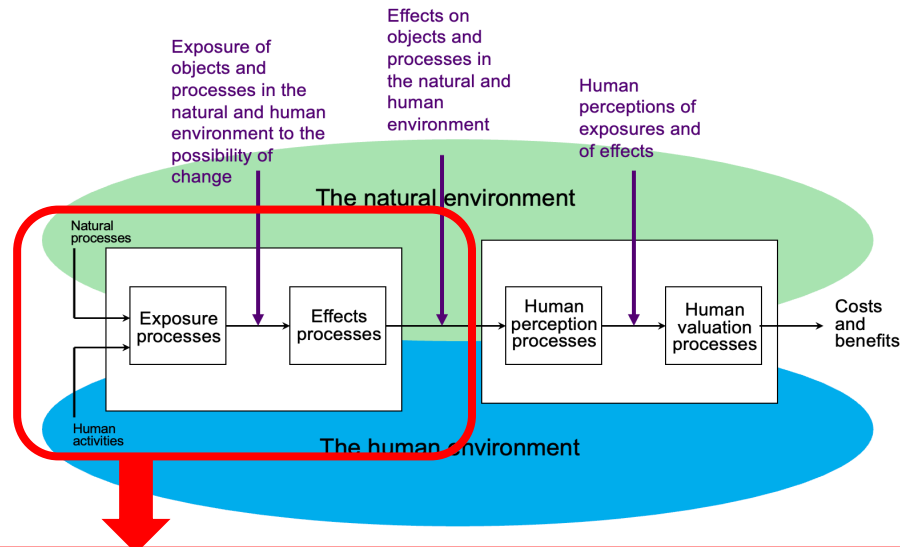
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# Here is a standard framework for risk analysis

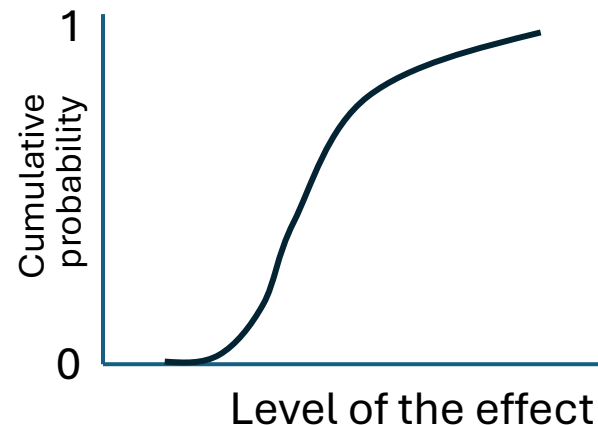


If you would like, later we can come back to talk more about these several parts

But, let's focus on just a few issues,  
starting with these first two boxes



When there is uncertainty, out of the analysis we do in these first two boxes we may be able to obtain probability distributions for the effects of interest that might look something like this



# Where do we get the probabilities?

From available data whenever possible...

...from expert judgment when those data do not exist...

...and, if there aren't any real experts, and/or the specific evidence is so poor that eliciting experts is not a defensible strategy, using order of magnitude and bounding analysis.

## The Neglected Art of Bounding Analysis

*Environmental Science & Technology*

April 1, 2001 / Volume 35, Issue 7 / pp 162 A — 164 A.

[M. GRANGER MORGAN](#)

**Are the answers provided by today's detailed risk analyses reasonable? Is valued insight being overlooked as a result of analysts' focus on the intimate details of environmental problems? If so, what can we do about this?**

Environmental risk analysis has fallen into a standard front-to-back mode of operation: Estimate the magnitude and pattern of releases of the pollutants of concern; model their transport and transformation through the environment; estimate the location and physiological state of people, animals, and plants and the exposures they will receive; apply dose-response functions; and estimate the resulting impacts.

All of this makes perfect sense if the relevant science is pretty well known and good data are available on factors such as the behaviors of the populations at risk. However, in practice, the science is often highly uncertain. The release rates may not be known with precision. There is often great uncertainty about transport and transformation processes.

**PERSPECTIVE**

**Use (and abuse) of expert elicitation in support of decision making for public policy**

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Edited by William C. Clark, Harvard University, Cambridge, MA, and approved March 18, 2014 (received for review October 22, 2013)

The elicitation of scientific and technical judgments from experts, in the form of subjective probability distributions, can be a valuable addition to other forms of evidence in support of public policy decision making. This paper explores when it is sensible to perform such elicitation and how that can best be done. A number of key issues are discussed, including topics on which there are, and are not, experts who have knowledge that provides a basis for making informed predictive judgments; the inadequacy of using qualitative uncertainty language; the role of cognitive heuristics and of overconfidence; the choice of experts; the development, refinement, and iterative testing of elicitation protocols that are designed to help experts to consider systematically all relevant knowledge when they make their judgments; the treatment of uncertainty about model functional form; diversity of expert opinion; and when it does or does not make sense to combine judgments from different experts. Although it may be tempting to view expert elicitation as a low-cost, low-effort alternative to conducting serious research and analysis, it is neither. Rather, expert elicitation should build on and use the best available research and analysis and be undertaken only when, given those, the state of knowledge will remain insufficient to support timely informed assessment and decision making.

Society often calls on experts for advice that requires judgments that go beyond well-established knowledge. In providing such judgments, it is common practice to use simulation models, engineering-economic assessment, and similar tools. Although such analytical strategies can provide valuable insight, they can never hope to include all relevant factors. In such situations, the community of applied decision analysis has long used quantitative expert judgments in the form of subjective probability distributions that have been elicited from relevant experts. Most such applications have been undertaken in support of decisions being made by private parties (1–4). Sometimes the resulting distributions are used directly, and sometimes they are fitted to formal functions and used in various Bayesian decision models (2, 5).

The use of expert elicitation in public sector decision making has been less common. Several studies have explored issues such as the health impacts of fine particle air pollution (6–12) and of lead pollution (13), the likely nature and extent of climate change (14–16), the various impacts that may result from climate change (17, 18), herbicide-tolerant oilseed crops (19), and the likely cost and performance of various energy technologies (20–24). The Environmental Protection Agency (EPA) has begun to make use of elicitation methods to address uncertain issues in environmental science (25), and those who work in both the Department of Energy and the Food and Drug Administration (FDA) have expressed interest in possibly using the method.

Done well, expert elicitation can make a valuable contribution to informed decision making. Done poorly it can lead to useless or even misleading results that lead decision makers astray, alienate experts, and wrongly discredit the entire approach. In what follows, I draw on relevant literature and 35 y of personal experience in designing and conducting substantively detailed expert elicitations, to suggest when it does and does not make sense to perform elicitations, how they should be designed and conducted, and how I believe the results should and should not be used. In contrast to much of the literature in Bayesian decision-making and applied decision analysis, my focus is on developing detailed descriptions of the state of understanding in some field of science or technology.

**First, Are There Any Experts?**

To conduct an expert elicitation, there must be experts whose knowledge can support informed judgment and prediction about the issues of interest. There are many topics about which people have extensive knowledge that provides little or no basis for making informed predictive judgments. For example, the further one moves away from questions whose answers involve matters of fact that are largely dependent on empirical natural or social science and well-validated models into realms in which individual and social behavior determine the outcomes of interest, the more one should ask whether expertise, with predictive capacity, exists. For example, given a specified time series of future radiative forcing and other relevant physical variables, in my view, it is reasonable to ask climate scientists to make probabilistic judgments about average global temperature 150 y in the future. I am far less persuaded that it makes sense to ask “experts” questions that entail an assessment of how the stock market, or the price of natural gas will evolve over the next 25 y, or what the value of gross world product will be 150 y in the future.

**The Interpretation of Probability**

A subjectivist or Bayesian interpretation of probability (5, 26–28) is used when one makes subjective probabilistic assessments of the present or future value of uncertain quantities, the state of the world, or the nature of the processes that govern the world. In such situations, probability is viewed as a statement of an individual's belief, informed by all formal and informal evidence that he or she has available. Although subjective, such judgments cannot be arbitrary. They must conform to the laws of probability. Further, when large quantities of evidence are available on identical repeated events, one's subjective probability should converge to the classical frequentist interpretation of probability.

Partly as a result of their different training and professional cultures, different groups of experts display different views about the appropriateness of making subjective probabilistic judgments, and have different levels of willingness to make such judgments. Although every natural scientist and engineer I have ever interviewed seemed to think naturally in terms of subjective probabilities, others, such as some experts in the health sciences, have been far

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www.pnas.org/cgi/doi/10.1073/pnas.1319841111

# When uncertainty is high, and consequences could be serious, we should exercise caution

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## Risk-based decision analysis in support of precautionary policies

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### Abstract

**A decision-analytic model for avoiding a risky activity is presented. The model considers the benefit and cost of avoiding the activity, the probability that the activity is unsafe, and scientific tests or studies that could be conducted to revise the probability that the activity is unsafe. For a single decision maker, thresholds are identified for his or her current subjective probability that the activity is unsafe. These thresholds indicate**

392

DeKay et al.

whether the preferred course of action is avoiding the activity without further study, engaging in the activity without further study, or conducting a test or research programme to obtain additional information and following the result. When these thresholds are low, precautionary action is more likely to be warranted. When there are multiple stakeholders, differences in their perceptions of the benefit and cost of avoidance and differences in their perceptions of the accuracy of the additional information provided by the test or research programme combine to create differences in their decision thresholds. Thus, the model allows for the rational expression of differences among parties in a way that highlights disagreements and possible paths to conflict resolution. The model is illustrated with an application to phytosanitary standards in international trade and examined in terms of recent empirical research on lay perceptions of risks, benefits, and trust. Further research is suggested to improve the elicitation of model components, as a way of fostering the legitimate application of risk-based decision analysis in precautionary policy making.

**KEY WORDS:** decision analysis, precautionary principle, risk, standard of proof, threshold, trust

### 1. Introduction

Risk-based decision analysis provides a structured framework for making rational decisions when outcomes are uncertain. It does so by considering alternative actions intended to enhance benefits or avoid or mitigate losses, the possible outcomes associated with each of these actions, and the probabilities and relative desirabilities of these outcomes. It can also be used to evaluate options for reducing the uncertainty surrounding the probabilities and magnitudes of possible outcomes through diagnostic tests or additional research.

For many proposed activities, such as the introduction of a new drug, the administration of growth hormones to beef cattle, or the incineration of municipal or industrial waste, a clear set of benefits can be anticipated by those who support the action (though the value, allocation, and sustainability of these benefits may be subject to controversy). However, the risks associated with these programmes, should 'things not go as planned,' are usually much more uncertain. This is often the case when risks arise from low-probability, high-consequence events and processes within complex systems, such as those involving human health, ecology, and sociocultural institutions. The complexity and low probability of these effects can make it very difficult to assess and reduce the resulting uncertainty – so much so that some believe that problems of this type strain the capabilities of risk-based decision analysis.

Such concerns are part of the motivation for precautionary approaches that would substitute highly protective decision rules for the calculus and tradeoffs of risk-based decision making. In a classic paper, Page (1978) offered a characterization of the type of risk for which support for precautionary approaches is most likely to arise. Page described the (still) emerging set of technological, health, and environmental risks for which:

- there is an *ignorance of mechanism* – so that our knowledge of the physical processes that determine the likelihood and magnitude of the risk is poor

# An example of overconfidence

Elicited cumulative distribution functions of the likely future costs of photovoltaic technology in 2030 obtained from seven energy experts in 2009-10.

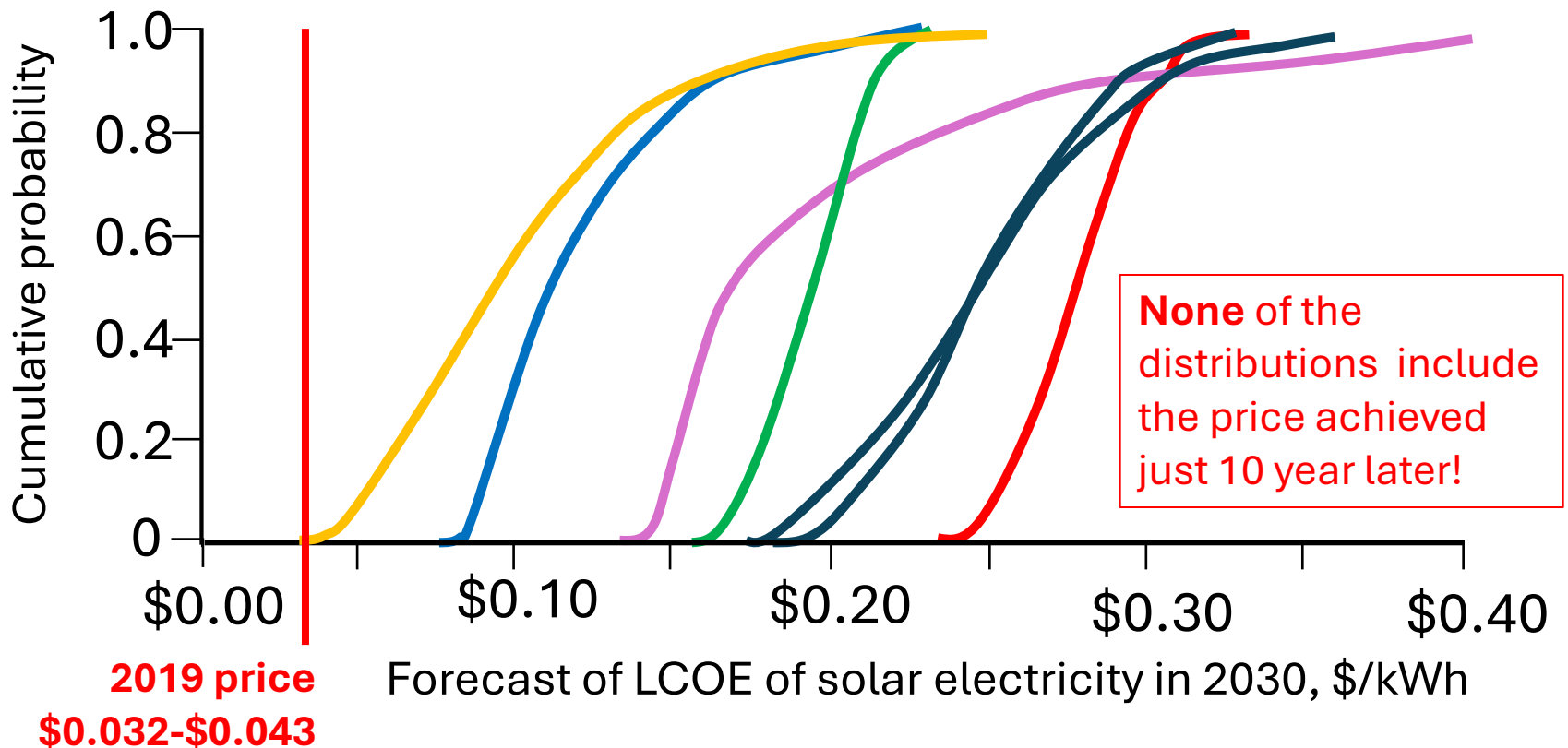
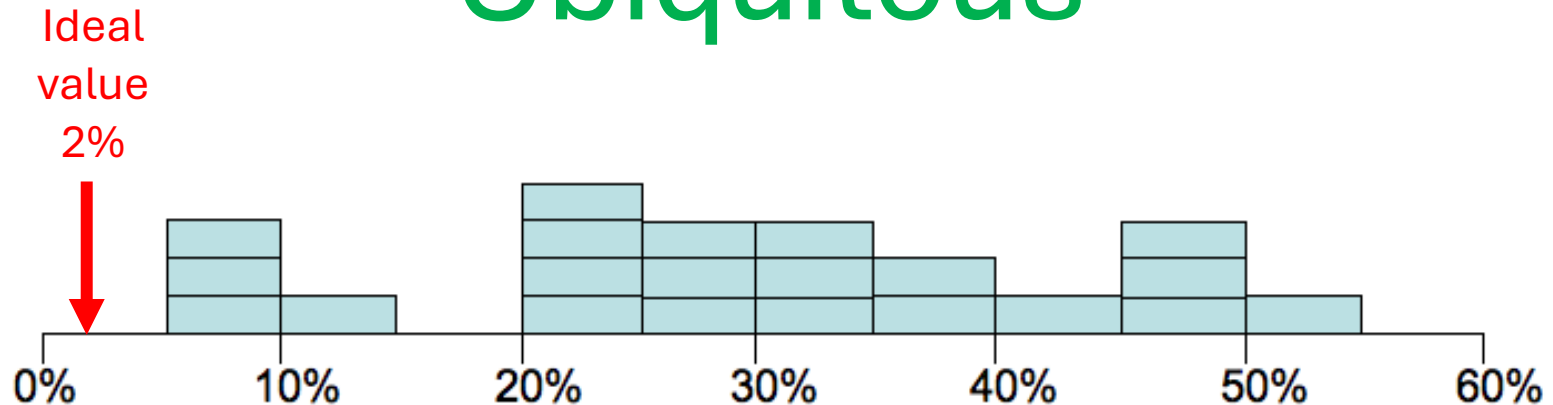


Image from: Savage, T., Davis, A., Fischhoff, B., & Morgan, M. G. (2021). A strategy to improve expert technology forecasts. *Proceedings of the National Academy of Sciences*, 118(21), e2021558118.

# Overconfidence is Ubiquitous



Percentage of estimates in which the true value lay outside of the respondent's assessed 98% confidence interval.

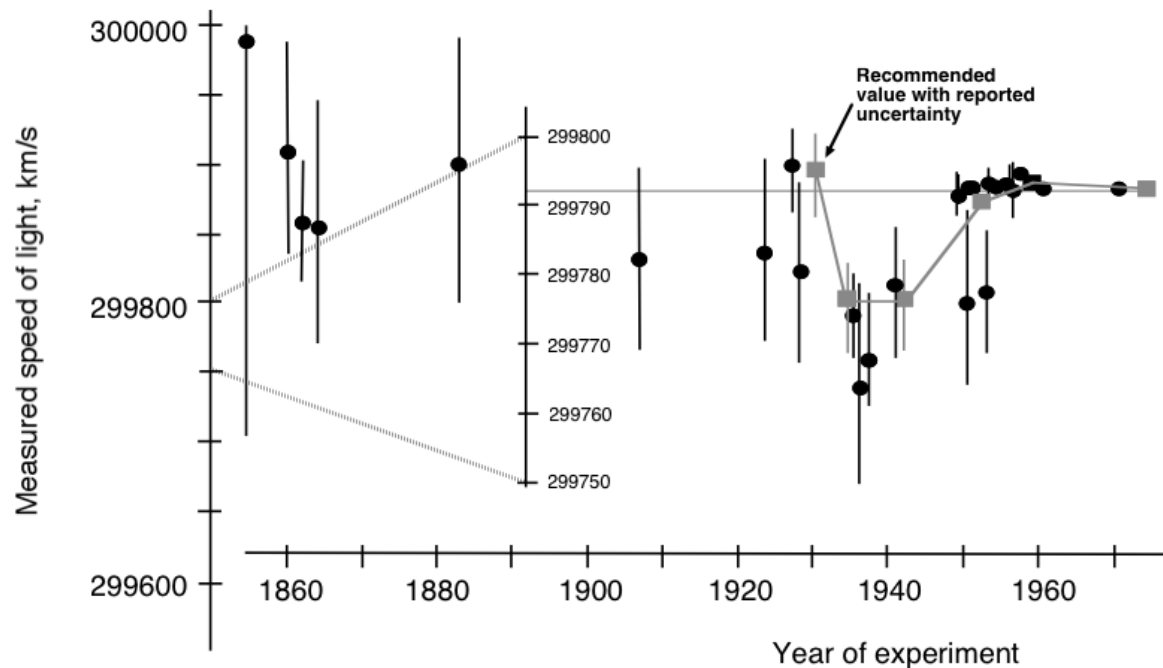
In 21 different studies involving over 10,000 assessment questions people were asked to assess 98% confidence intervals in a variety of judgments.

This histogram reports that value of the "surprise index," fraction of the time answer lies outside the assessed 98% confidence interval (ideal value = 2%)

These results indicate clearly the ubiquitous tendency to overconfidence (i.e., assessed probabilities that are too narrow). A more detailed summary is provided in Morgan and Henrion, *Uncertainty*, Cambridge, 1990.



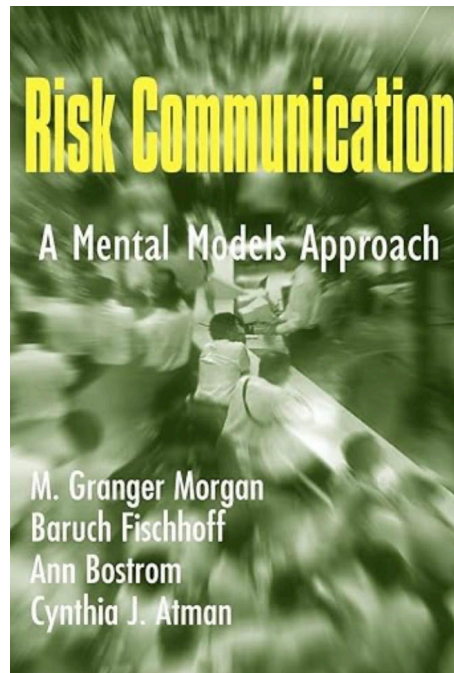
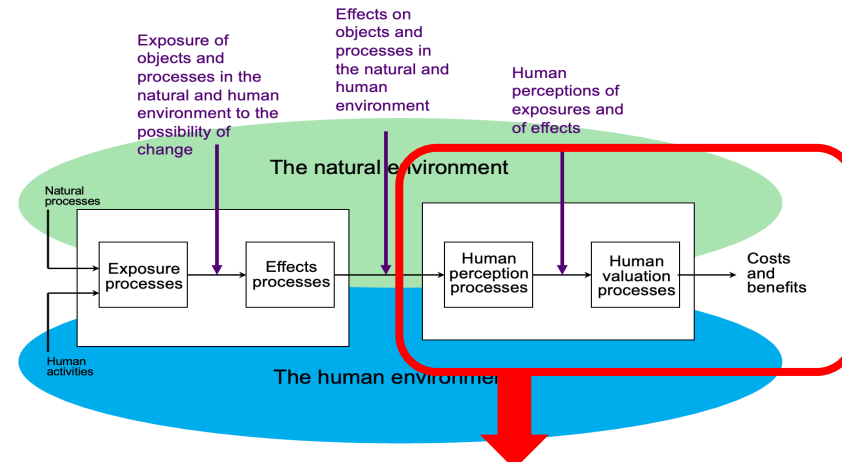
# One more example for experts



Published estimates of the speed of light. The light gray boxes that start in 1930 are the recommended values from the particle physics group that presumably include an effort to consider uncertainty arising from systematic error. Note that for over two decades the reported confidence intervals on these recommended values did not include the present best-measured value. Henrion and Fischhoff (1986), from which this figure is combined and redrawn, report that the same overconfidence is observed in the recommended values of a number of other physical constants.



# A few issues involving the next two boxes

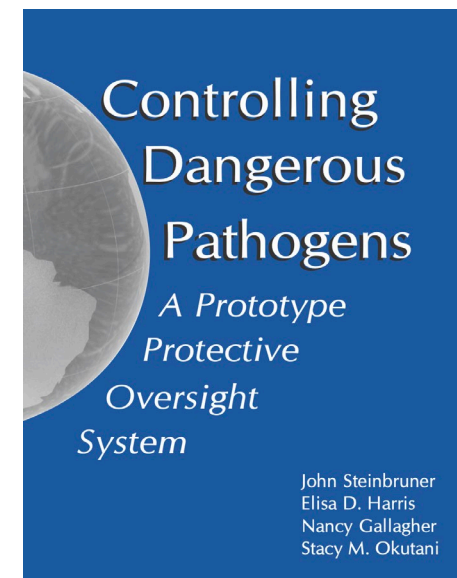
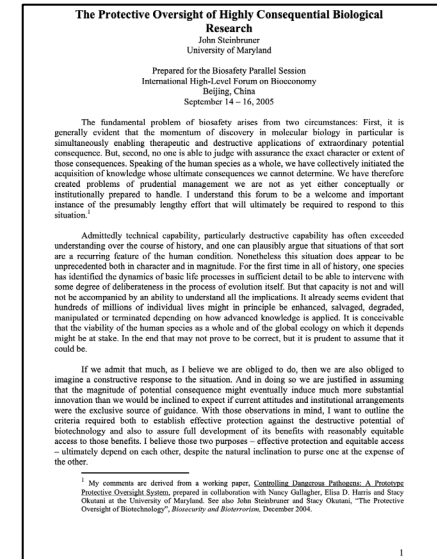


John D. Steinbruner

# Before his untimely death, security expert John Steinbruner and colleagues...

...worked on managing the risks posed by synthetic biology. That work focused on biology that entailed risks that are ***much*** more immediate, and presumably larger, than those we are considering today.

BUT, one key contribution from their work was a set of arguments about the importance of being *open* and *transparent* about the research.



# This suggests the need to consider this balance



With that I'll say thanks.

I'll be happy to talk more  
about any of this during the  
discussion.