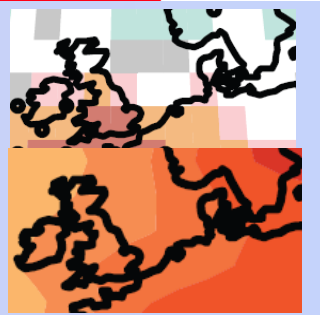


On the Relevance of Climate Model Output for Decision Support

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Scotland in White & Red (SPM6 & 7)

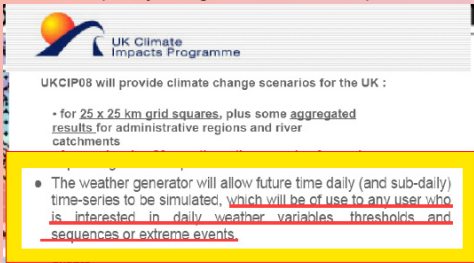
ABSTRACT: How exactly can decision-support and policy making benefit from the use of multiple climate model experiments in terms of coping with the uncertainties on climate change projections? Climate modelling faces challenges beyond those of weather forecasting or even seasonal forecasting, as with climate we are now (and will probably always be) required to extrapolate to regimes in which we have no relevant forecast-verification archive. This suggests a very different approach from traditional methods of mixing models and skill based weighting to gain profitable probabilistic information when a large forecast-verification archive is in hand. In the case of climate, it may prove more rational to search for agreement between our models (in distribution), the aim being to determine the space and timescales on which, given our current understanding, the details of the simulation models are unimportant. This suggestion and others from Smith [4] are interpreted in the light of recent advances. Climate models are large nonlinear dynamical systems which insightfully but imperfectly reflect the evolving weather patterns of the Earth. Their use in policy making and decision support assumes both that they contain sufficient information regarding reality to inform the decision, and that this information can be effectively communicated to the decision makers. There is nothing unique about climate modelling and these constraints, they apply in all cases where scientific modelling is applied to real-world actions (other than, perhaps, the action of improving our models). Starting with the issue of communication, figures from the 2007 IPCC Summary for Policy Makers will be constructively criticized from the perspective of decision makers, specifically those of the energy sector and the insurance/reinsurance sector. More information on basic questions of reliability and robustness would be of significant value when determining how heavily to weight climate model output in the decision process; one obvious example is the question of over what spatial and time averages modellers expect information in current climate distributions to be robust. The IPCC itself suggests continental/seasonal, while distributions at **10 of kilometres/hourly** are on offer. Our aim here is not to resolve this discrepancy, but to develop methods with which it can be addressed. This is illustrated in the context of using another physically based, imperfect model setting: using Newton's laws in an actual case of NASA hazard evaluation. Our aim is to develop transparent standards of good practice managing expectations, which will allow model improvements over the next decades to be seen as progress by the users of climate science.

Should climate scientists tell users what we know, or what they want to hear?

Numerate users of climate information (eg the energy, insurance and financial sectors) are both numerous and influential. While they would love high space-time resolution insights of 2080, as offered by the BBC, UKCIP and others; they question whether this(say 5km-hourly) information is any more quantitatively relevant than economic forecasts at that lead time. The credibility of science and science-based policy hangs on how we respond.



Only ~5% of UK actuaries hold climate change is a myth, and ~60% believe it will impact their work.



Observational uncertainty: noise and sampling issues

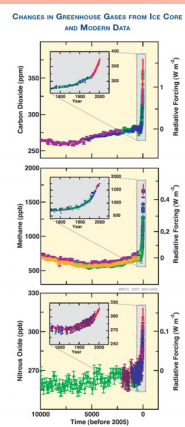


Figure SPM4. Atmospheric concentrations of carbon dioxide, methane and nitrous oxide from the last 1800 years (left panels) and since 1750 (right panels). Measurements are shown from ice cores (left) and direct observations (right). The corresponding radiative forcings are shown on the right hand side of the right panels. (Figure 6.4)

The first and third figures of the IPCC SPM [1], treat a well understood form of uncertainty, and while statistical scrutiny may bring details into question (a la Nychka), the physical picture is clear:

The uncertainty "bars" and "bands" reflect limits imposed by deficiencies in the instruments and sampling, and where unknown independent estimates, if discovered, are expected to fall.

Yet the SPM gives bands and bars not only for these well understood uncertainties, but also for "model range" in the future, where the method to apply this information in decision support is, at best, less clear.

Can we quantify the value-added by running large ensembles of GCM simulations? And improve the presentation to optimise communication with the users of climate science leading to better decisions and policy in practice?

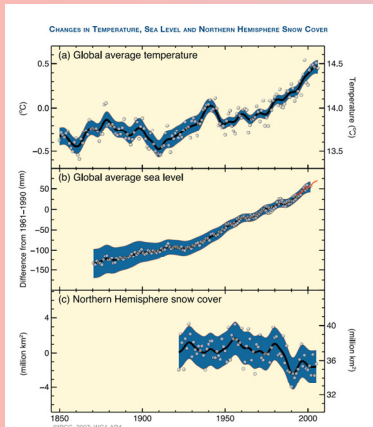


Figure SPM5. Observed changes in (a) global average surface temperature, (b) global average sea level from tide gauge data and satellite altimetry, and (c) Northern Hemisphere snow cover for March-April. All changes are relative to corresponding averages for the period 1961-1990. The shaded areas represent the 95% confidence intervals estimated from a comprehensive analysis of known uncertainties in (a) and (b) and from the time series in (c). (Figure 5.1)

Multi-model variability is not a traditional form of uncertainty

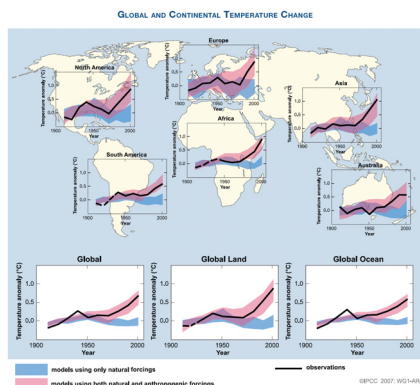


Figure SPM4. Comparison of observed continental- and global-scale changes in surface temperature with results simulated by climate models using natural and anthropogenic forcings. Decadal averages of observations are shown for the period 1980 to 2000 (black line) plotted against the centre of the decade and relative to the corresponding average for 1961-1990. Lines are dashed where spatial coverage is less than 50%. Blue shaded bands show the 5-95% range for 19 simulations from five climate models using only the natural forcings due to solar activity and volcanoes. Red shaded bands show the 5-95% range for 18 simulations from 14 climate models using both natural and anthropogenic forcings. (Figure 6.2, Figure 7)

The graph on the left, SPM4, again shows observation based curves, but this time with model based "bands." It is far from clear how one sure interpret these bands, when the models are known **not** to relate to any simple sampling interpretation. SPM5 (right) segues into the future, with uncertainty bands in the past **and** model ranges in the future. What do these mean?

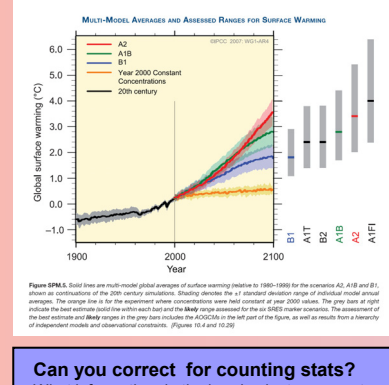


Figure SPM5. Solid line is multi-model global average of surface warming relative to 1880-1980 for the scenarios A2, A1B and B1, shown as contributions of the 20th century simulations. Shaded bands show the 5-95% confidence range of individual model annual averages. The shaded line is for the experiment where concentrations were held constant at year 2000 values. The grey line at right indicates the best estimate of the total warming and the black line at left the best estimate of the total warming. The assessment of the best estimate and 5-95% range in the grey line includes the AGOCS in the left panel of the figure, as well as results from a sensitivity of independent models and observational constraints. (Figure 10.4 and 10.5)

Can you correct for counting stats?
What information do the bands aim to convey to policy makers here in the SPM?
Why is the stronger forcing (red band) more narrow than the green? (17, 21, 21, 10)
[Thanks to Reto Knutti for numerous discussions on this.]

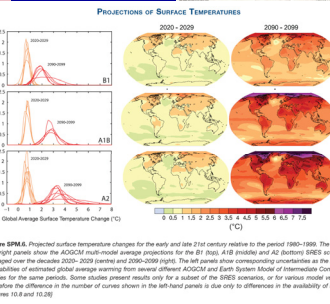
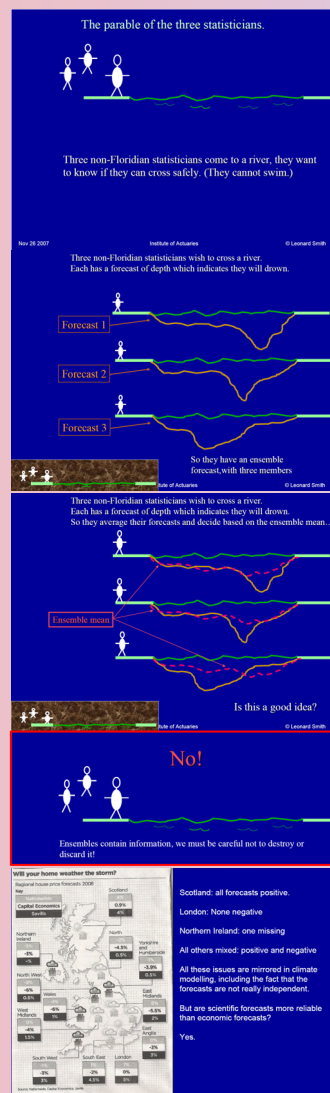


Figure SPM6. Projected surface temperature changes for the early and late 21st century relative to the period 1980-1999. The central and left panels show the AGOCS multi-model average projections for the B1, A1B, A1B (middle) and A2 (bottom) scenarios. The right panel shows the AGOCS multi-model average projections for the period 2000-2099. The shaded areas represent the 95% confidence intervals estimated from a comprehensive analysis of known uncertainties in (a) and (b) and from the time series in (c). (Figure 5.1)

Do we risk our credibility by selling what we know is not robust?

2.6. Regional climate models are useful tools for assessing the impacts of climate change on the energy industry because they provide a **comprehensive, physically consistent, prudent projection** of future climate. Their **main** weaknesses are that their horizontal resolution is currently at best about 10 km, they are not integrated with energy industry infrastructure, e.g. water abstraction and there are **uncertainties in the predictions**.

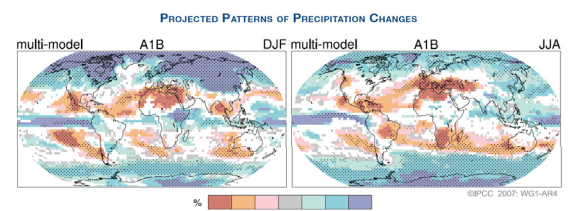
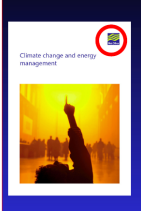


Figure SPM7. Relative changes in precipitation (in percent) for the period 2090-2099, relative to 1980-1999. Values are multi-model averages based on the GISS-A1B scenario for December to February (left) and June to August (right). White areas are where less than 60% of the models agree in the sign of the change and stippled areas are where more than 90% of the models agree in the sign of the change. (Figure 10.9)

If we believe our models are able to give probability forecasts that can be quantitatively used as such, then it is reasonable to expect **Probabilistic Similarity**: future forecasts will refine current forecasts, we do not expect revolutions due to resolving some model inadequacy. Given the currently level of disagreement between models below continental scales, is it reasonable to describe our projections at 10km scales as **"comprehensive, physically consistent and prudent"** for Scotland (top right SPM7)? And why is the model grid so unclear on the lower blow-up? Where is the justification of treating a PDE in this way?

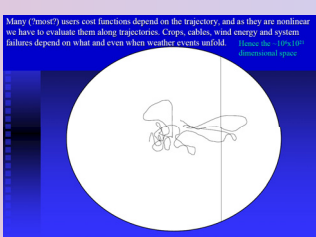
Is the greatest thing climate science has to fear climate science itself?

Negative impacts on science-based policy and decision support can be expected from:
(1)The now diminishing voice of nay-sayer sceptics, who are exhausting their well worn ammunition. (N. Orsekes)
(2)The increasing voice of non-physical scientists (statisticians) making innocent category errors. (S Armstrong)
(3) Climate scientists trying to deliver what is "required" rather than what is known, thereby feeding (1) & (2).

Numerate users are not unfamiliar with ensemble forecasts, and their limitations (see House forecasts in centre), yet they need guidance on whether they should treat climate as they do car accident-risk given gender-age, or as capital risk in the property market. What methodology do we hold they "should" adopt? Is figure SPM2 relevant here?

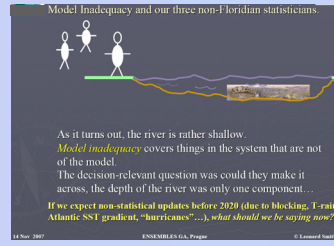
Open Questions for Climate Science in Decision Support:

What space and time scales do we have decision relevant information, probabilities that could be used as such? When models disagree (in distribution) can we expect to do better than *potential pathways*? How do we clarify the difference between economic modeling and physical simulation? Why trust simulation more? Can we argue our simulations are **Structurally Robust**? Are the distributions of interest robust against model details? Is there any mathematical foundation for the claim that we will converge toward something (in distribution) where the details (parameterizations, parameter values and structure) do not dominate the outcome at scales of interest? Can we provide a better method of application than pathways and analogy, if our probabilities are not to be used as such?



New ways Forward

Users often have nonlinear, threshold dependent cost functions; reproducing means and autocorrelations is of little value. Looking for shadowing trajectories [4,5,6] while noting their duration at various resolutions might inform users and model developers. How can we make our models more relevant? How can we exploit their empirical targets to speed improvements in model adequacy?



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References: Many available from the publications pages of www.lsecats.org
Discussion welcome at cats.lse.ac.uk/forum

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