

Evaluating Model Analysis of Climate Change Mitigation

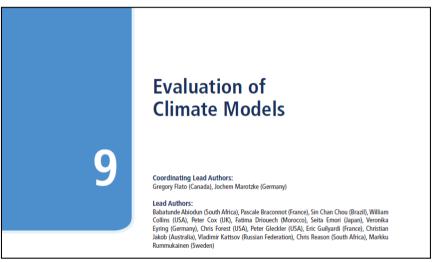
Charlie Wilson
November 2015, IAMC Conference

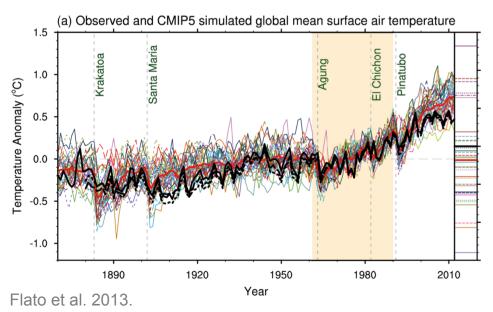
(Elmar Kriegler, Jana Schwanitz Celine Guivarch, Volker Krey, Keywan Riahi, Detlef van Vuuren)



How are IAMs evaluated? **To what end?** Why is IAM evaluation less **visible** than climate model evaluation?

GCMs IAMs



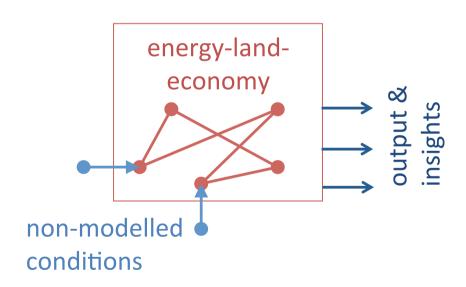




Evaluation is about whether models generate the "right behaviour for the right reasons"

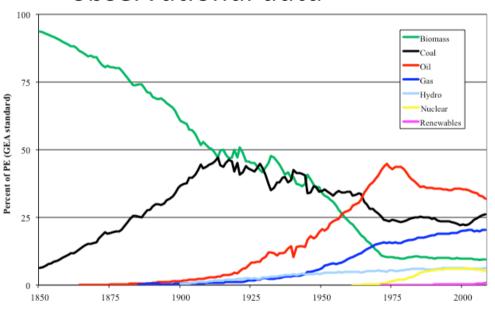
structural validity

model is an accurate representation of the system response being modelled



behavioural validity

model predictions are consistent with observational data

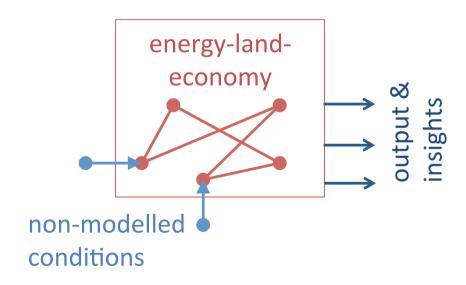




Behavioural validity can not be demonstrated for simulation models of **dynamic**, complex systems

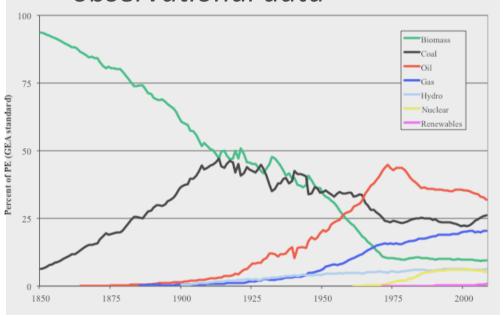
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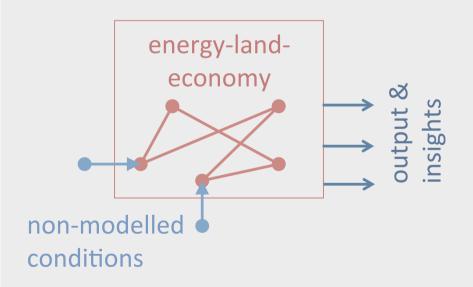
- over-tuning, non-uniqueness
- **★** limited to historical conditions



Structural validity can not be demonstrated for simulation models of dynamic, **complex** systems

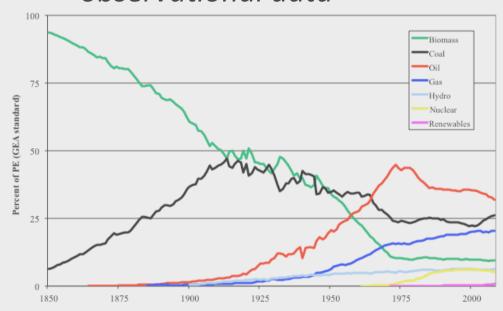
structural validity

model is an accurate representation of the system response being modelled



behavioural validity

model predictions are consistent with observational data



- irreducible uncertainties (data, parametric, structural)
- necessary simplifications



IAM evaluation is an **open-ended process** of testing, learning & improving a model and its performance

Evaluation criteria for IAMs

appropriateness is model purpose and design consistent with the research question?

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interpretability are model results clearly interpretable in light of model structure and parameterisation?

verifiability

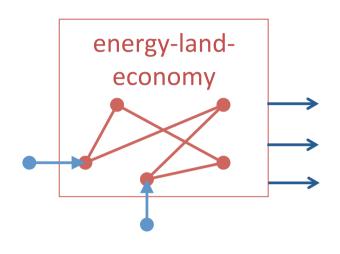
are model results repeatable or
is model structure accessible to 3rd parties?

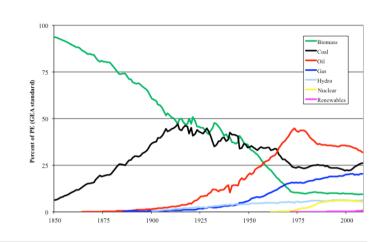
is model seen as good enough for its intended purpose by both users and modellers?

usefulness do model insights help understand uncertainties, trade-offs, alternatives?

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Different evaluation methods are used with IAMs, particularly to test structural validity





behavioural validity

structural validity

model checks

model inter-comparisons

historical trends

historical simulations

transparent documentation

diagnostic indicators

generalisable historical patterns

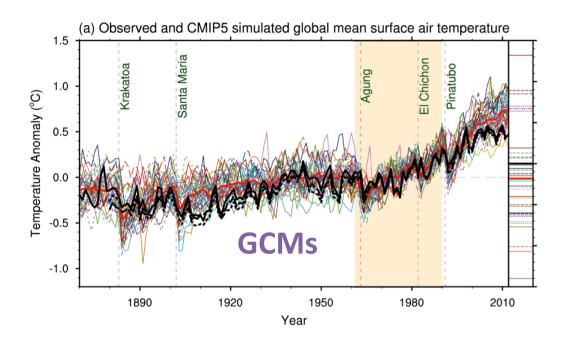
simple models

expert review

sensitivity analysis

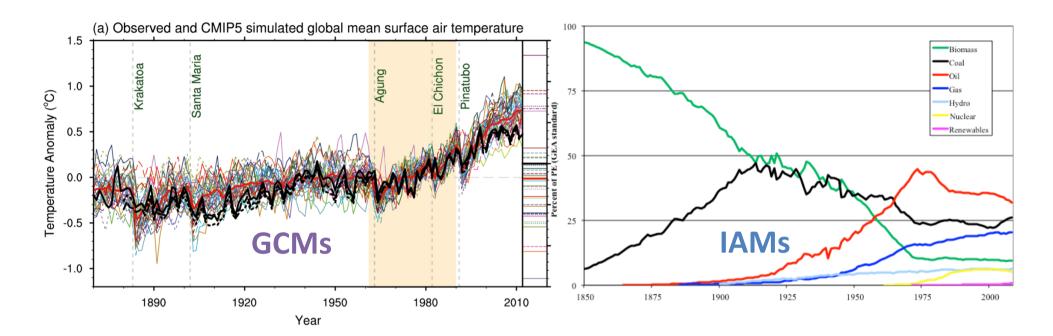


"Climate models ... reproduce many important aspects of **observed climate** ..." [IPCC AR1 - AR5]



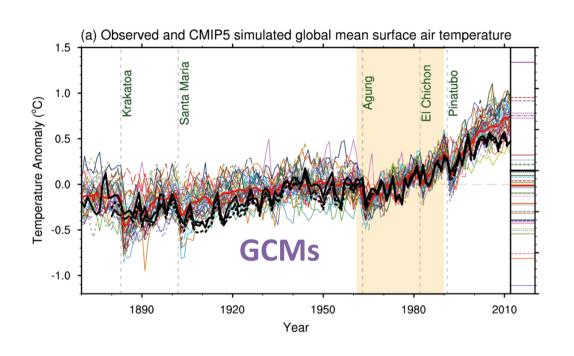
- many simulated (un-tuned) quantities for different processes & scales
- statistical measures of performance

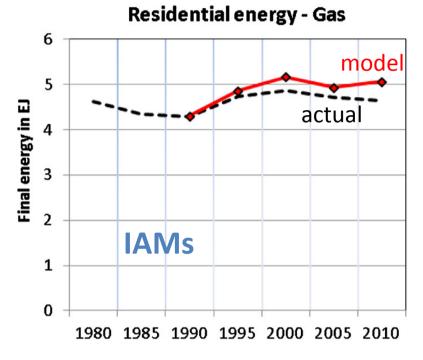




- many simulated (un-tuned) quantities
 for different processes & scales
- statistical measures of performance
- no long-run simulations of aggregate system variables

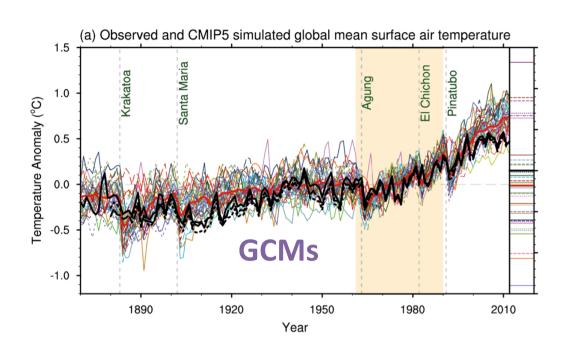


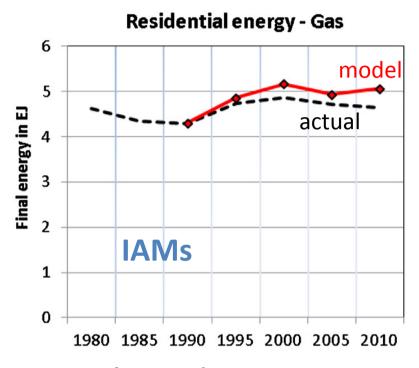




- many simulated (un-tuned) quantities for different processes & scales
- statistical measures of performance
- very limited in scope (process, time horizon)
- divergence -> Δ parameter





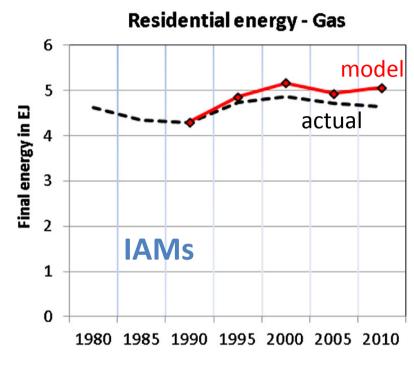


- many simulated (un-tuned) quantities for different processes & scales
- statistical measures of performance
- very limited in scope (process, time horizon)
- divergence -> ∆ parameter

similar issues with behavioural validity testing

over-tuning, non-uniqueness limited historical conditions





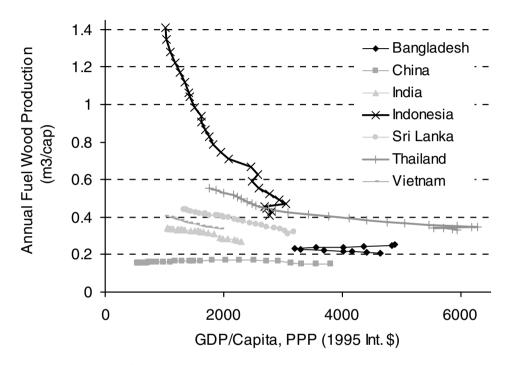
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issues unique to IAMs?

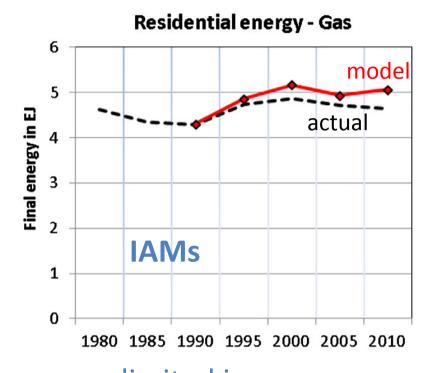
lack of good observational data heterogeneous causal processes (normative design)



The **historical record** can be used for IAM evaluation in other ways - **generalisable historical patterns**



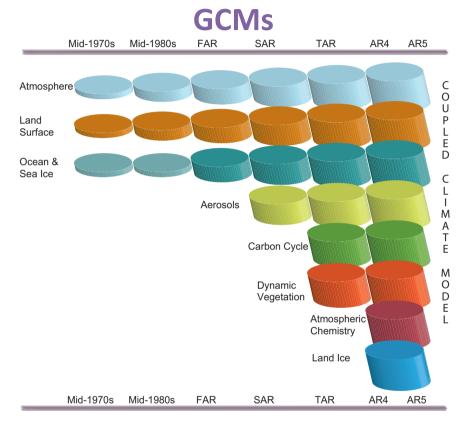
- useful learning exercise ...
 but what then?
- no clear methodology or metrics



- very limited in scope (process, time horizon)
- divergence -> △ parameter



Simple models help understand representations of key processes embedded in more complex models



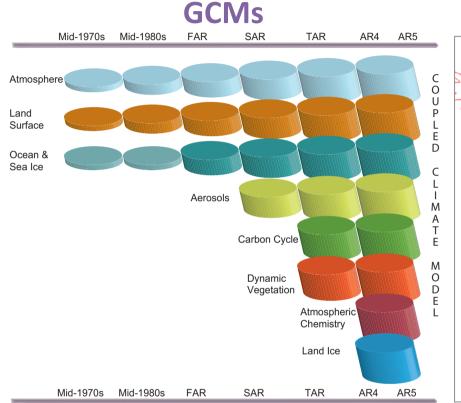
"A complex model may be more realistic, yet ... as we add more factors to a model, the certainty of its predictions may decrease even as our intuitive faith in the model increases."

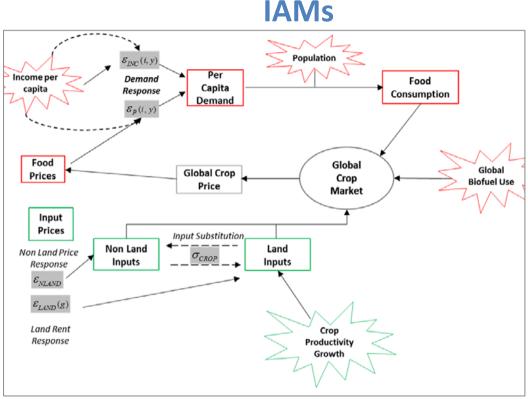


- elegance vs. elaboration
- simpler models preserved in a 'hierarchy of models'



Simple models help understand representations of key processes ... but are not common in IAMs



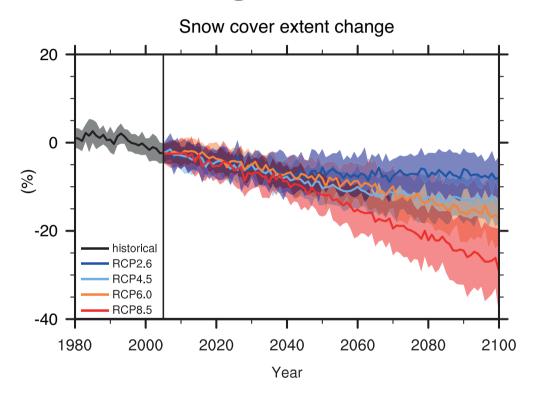


- elegance vs. elaboration
- simpler models preserved in a 'hierarchy of models'
- 'SIMPLE' global agriculture model
 - biophysical, economic
- historical simulations (1961-2000)



Model inter-comparison projects explore structural uncertainty (across different model representations)

GCMs e.g., CMIP5

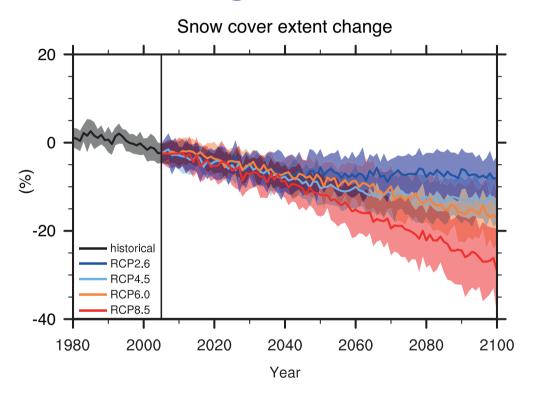


- harmonised experiments & results
- model performance metrics



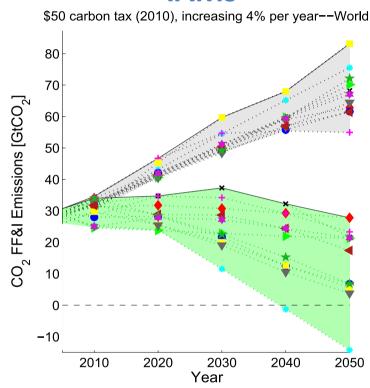
Model inter-comparison is a long tradition for IAMs (9 MIPs contributed 95% of AR5 mitigation scenarios)

GCMs e.g., CMIP5



- harmonised experiments & results
- model performance metrics

IAMs



- emphasis on robust results
- diagnostic indicators (recent)
- link structure <-> behaviour



Evaluation research for GCMs is generally more developed and prominently reported than for IAMs

evaluation method

historical simulations







generalisable historical patterns



simple models



model intercomparisons





- + sensitivity analysis
- + expert review
- + documentation ...

GCM vs. IAM differences

- (1) modelled system
- underlying principles
- observational data
- (2) domain of application
- uniqueness of insights
- expertise of policy users



Each evaluation method has characteristic strengths and weaknesses

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historical simulations

strengths

e.g., use of observations

weaknesses

e.g., limited applicability (time horizon, processes)

generalisable historical patterns

e.g., use of observed dynamics

e.g., unclear implications for structural validity

simple models e.g., insights robust to structural uncertainty

e.g., attribution of divergence to model differences

model intercomparisons e.g., understanding of key system processes

e.g., lack of realism

+ sensitivity analysis

- + expert review
- + documentation



Each evaluation method has strengths & weaknesses ... and contributes more to certain evaluation criteria

	appropriate-	interpret-	evaluation criteria		
evaluation method	ness	ability	verifiability	credibility	usefulness
historical simulations	~	~			
generalisable historical patterns		~		~	
simple models			~	~	
model inter- comparisons					

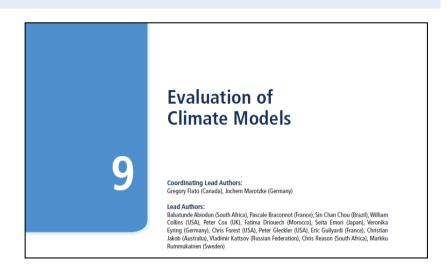
- + sensitivity analysis
- + expert review
- + documentation ...

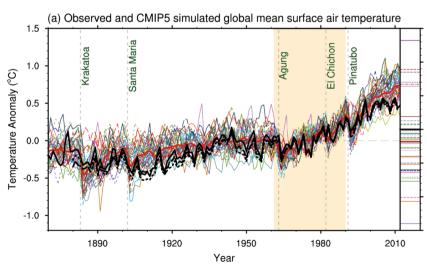


Conclusion: Systematic & more prominent evaluation effort to strengthen and maintain confidence in IAMs

- **systematic**: multiple methods concurrently
- prominent:
 concerted, synthesis products
- learning: insights from GCMs

ongoing articulation of the grounds on which IAMs can be declared good enough for their intended uses









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