



Interfacing physical experiments and computer models



Preliminary remarks
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Linking simulators to reality

- ▶ Before we can think about designing observational studies we have to think about how the simulation model relates to reality
- ▶ It is absolutely vital to recognise that all models are wrong
- ▶ It is also important to distinguish between two objectives
 - ▶ Calibration – learning about parameters in the simulator
 - ▶ Prediction – and in this case there's a distinction between
 - ▶ Interpolation = prediction within the domain of observations
 - ▶ Extrapolation = prediction outside that domain



Traditional formulation

- ▶ The problem
 - ▶ We have a simulator that predicts a real world phenomenon
 - ▶ We have some observations of the real world
- ▶ Formally, the simulator takes two kinds of inputs
 - ▶ Calibration parameters θ
 - ▶ Control inputs x
 - ▶ Simulator is written $y = f(x, \theta)$
 - ▶ Observation z_i is obtained at control input values x_i
 - ▶ Experimental design involves choosing those x_i points in x space
- ▶ It is usual to write
$$z_i = f(x_i, \theta) + \varepsilon_i$$
 - ▶ where ε_i are independent observation errors
- ▶ But this is **wrong** because the model is wrong

Model discrepancy

- ▶ It is necessary to acknowledge model discrepancy
 - ▶ There is a difference between the model with best/true parameter values and reality
$$y = f(\mathbf{x}, \theta) + \delta(\mathbf{x}) + \varepsilon$$
 - ▶ where $\delta(\mathbf{x})$ accounts for this discrepancy
- ▶ Kennedy and O'Hagan (JRSSB, 2001) introduced this model discrepancy
 - ▶ Modelled it as a zero-mean Gaussian process
 - ▶ Conditional on some hyperparameters
 - ▶ They claimed it acknowledges additional uncertainty
 - ▶ And mitigates against over-fitting of θ
 - ▶ Subsequent experience suggests that caution is needed

The bottom line

- ▶ If you don't include some representation of model discrepancy, you can expect to get nonsense
 - ▶ That applies to experimental design, too
- ▶ Even if you do include model discrepancy, it's essential to **think** about it and model it carefully
 - ▶ Especially for calibration
 - ▶ Use knowledge about aspects of reality that are not adequately represented in the model
- ▶ Even if you do think about it and model it carefully,
 - ▶ You will not be able to learn the true physical values of calibration parameters, or to make accurate extrapolations
 - ▶ Not even with an unlimited number of physical observations
 - ▶ But interpolation predictions may be quite accurate
 - ▶ And increasingly so with more observations



And finally

- ▶ I am not aware of any work on design of physical experiments that includes any formal representation of model discrepancy
 - ▶ Whether for purposes of prediction or calibration
- ▶ But I hope to be corrected!

