

An Application of Reification for a Rainfall-Runoff Computer Model

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For the next 30 minutes...

- ▶ Background Computer models/Simulators
- ▶ What the heck is *Reification*?
- ▶ Motivation for this work - justify reification
- ▶ Application: Rainfall-Runoff Model
- ▶ Discussion

Why might we use simulated data?

To assess the behavior of **complex physical systems**,



we often start by collecting data.

Simulator based inferences include...

- ▶ When we use simulated data to assess characteristics of a complex system we have/model the following:

z : Real world observation(s)

y : Characteristic(s) of the complex system

f : Deterministic simulator or computer model

X : The input variables

- ▶ For a Design of Experiments (DOE), we have

$$X = \{x_1, \dots, x_n\}$$
$$\{f\}_{[n]} = \{f(x_1), \dots, f(x_n)\}$$

- ▶ Example: Assess water runoff from Swiss catchment

z_t : Measured run-off at time t

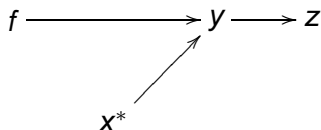
y_t : Actual run-off at time t

$f_t(x_i)$: Simulated run-off at time t given $X = x_i$



Hierarchical model

- ▶ Conditional independence is often desired
- ▶ Given a tuned input x^* ,



- ▶ And, we might have the following hierarchical model

$$z = h(y) + e_z, \quad e_z \sim \text{No}(0, \sigma_z^2)$$

$$y = f(x^*) + e_y, \quad e_y \sim \text{No}(0, \sigma_y^2)$$

$$x^* \sim \text{Un}(a, b)$$

$$f(x) = g(x)\beta + e(x), \quad e(x) \sim \text{GP}(0, \Sigma_f)$$

One Source of Model Uncertainty

The model f itself is uncertain!

- ▶ $f(x)$ is inevitably *wrong*
→ depends on theories, assumptions, comp. resources...
- ▶ In turn, the orthogonal assumption:

$$\text{Cov}[\epsilon_y, \{f, \mathbf{x}^*\}] \neq 0$$

is rarely, if ever, appropriate.

- ▶ Existence of structural error in f ; f does not mimic y perfectly
- ▶ What do we do?
 - a) Ignore the data weakness?
 - b) Re-code f and repeat a simulation experiment?



Reification (Goldstein and Rougier, 2008)

- ▶ Answer: Take advantage of being a Bayesian and elicit good, prior, expert judgments about the **improved model**, f^*
- ▶ The orthogonal assumption may hold for a hypothetical, ideal model $f^*(\tilde{x}^{**})$ at it's best input (\tilde{x} may represent the same variables as x or the union of x and other variables)

$$y = f^*(\tilde{x}^{**}) + \epsilon_y^*, \quad \text{where } \epsilon_y^* \perp \{f, f^*, x^{**}\}.$$

- ▶ So, f^* represents our **reified** model (where, f is our *realized* model)
- ▶ M-W Dictionary: To reify is to regard (something abstract) as a material or concrete thing
- ▶ How do we come up with a good f^* ?

Two Types of Reification: General

- ▶ A model reified generally is simply a realized model with more uncertainty.

$$f^*(x) = g(x)\beta^* + e^*(x) \quad (1)$$

where $E[\beta^*] = E[\beta]$ and $E[e(x)] = E[e^*(x)]$, but

$$\text{Var}[\beta^*] = \Sigma_{\beta}^* > \Sigma_{\beta}, \quad \text{Var}[e^*(x)] = \Sigma_{e(x)}^* > \Sigma_{e(x)}$$

- ▶ Note, emulator (1) can be written as

$$f^*(x) = g(x)(\beta + \beta^+) + e(x) + (e^*(x) - e(x))$$

where $E[\beta^+] = 0$ and $\text{Var}[\beta^+] = \Sigma_{\beta}^+$. Judgments about the change in β and $e(x)$ might be easier for an expert than direct judgments about β^* and $e^*(x)$.

Two Types of Reification: Structural

- ▶ Structural reification adds one or more hypothetical input variables ν to a current, realized simulator/emulator.
- ▶ In some cases, there exists a value ν_0 , such that the reified outcome $f^*(x, \nu)$ equals the realized outcome $f(x)$ when $\nu = \nu_0$.

$$\begin{aligned} f^*(x, \nu) &= g^*(x, \nu)\beta^* + e^*(x, \nu) & (2) \\ &= g(x)\beta + e(x) + g^+(x, \nu)\beta^+ + e^+(x, \nu) \\ &= f(x) + g^+(x, \nu)\beta^+ + e^+(x, \nu), \end{aligned}$$

where, $f(x)$ is learned from the data; $g(x, \nu_0) = 0$; and $e(x, \nu_0) = 0$.

Combine reification methods

► Combine:

1. Reify f structurally and emulate by $f'(x)$
2. Reify $f'(x)$ generally and emulate by $f^*(x)$

► We have the following hierarchical model,

$$f^*(x) | f'(x), x, \nu = f'(x) + g^{+'}(x, \nu) \beta^{+*} + e^{+*}(x, \nu)$$

$$f'(x) | f(x), x, \nu = f(x) + g^{+'}(x, \nu) \beta^{+'} + e^{+'}(x, \nu)$$

$$f(x) | x, \nu = g(x) \beta + e(x) \quad \text{where,}$$

► If $E[\beta^{+*}] = E[e^{+*}(x, \nu)] = E[\beta^{+'}] = E[e^{+'}(x, \nu)] = 0$,

$$E[f^*(x) | f'(x), x, \nu] = f'(x)$$

$$E[f'(x) | f(x), x, \nu] = f(x)$$

$$E[f(x) | x] = g(x) E[\beta] + E[e(x)].$$

Criticisms of Reification/Motivation of Work

- ▶ Quantitative comparisons between a realized and reified model do not exist because reified models are hypothetical
- ▶ Only expert critiques can validate the subjective judgments used to specify a reified model
- ▶ So, I asked,

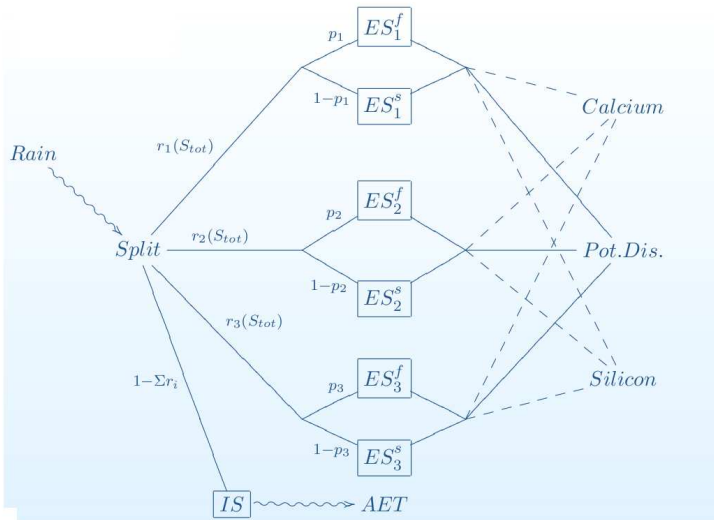
How do we elicit a good f^ from experts that other experts and non-experts might accept?*

- ▶ One idea:
 1. Create a reduced vs. complete scenario
 2. Reify the reduced model *honestly*
 3. Compare the reified and complete models
 4. If the reification is successful, repeat for the complete model
- ▶ Example...



Graph of the Rainfall-Runoff Model

Three compartment mixing model with parallel transfer
(Iorgulescu, I., Beven, K.J., and Musy, A., 2005)

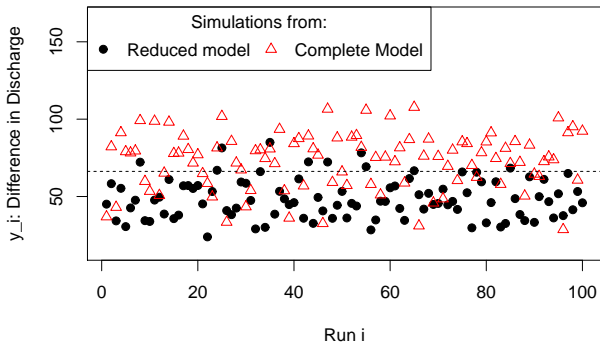


Details of the Complete Rainfall-Runoff Model

- ▶ Research Goal: Assess the true hydrological process for how rainfall becomes runoff, based on estimates of **total discharge** and **tracer concentrations** from **real-world measurements** and **computer simulations**,.
- ▶ Rainfall flows into one of 3 compartments (DP, GW, AS)
- ▶ Inputs: 6 parameters per compartment $p^{(k)}, c_f^{(k)}, c_s^{(k)}, a^{(k)}, b^{(k)}, k^{(k)}$, $k \in [PD, GW, AS]$ and one constraint totaling to 17 inputs.
- ▶ Forcing functions: measured rainfall $rain(t)$ and assumed evaporation rates $AET(t)$, $t \in [1, \dots, 839]$ hours
- ▶ Outputs include 3 time series: water discharged (PD), and the concentration of two tracer chemicals (Ca and Si) at each hour.
- ▶ For now, consider only PD(620)

Reduce the RR Model

- ▶ Exclude compartment AS; 11, not 17, inputs
- ▶ We create a 17 dimensional Latin hypercube (LHC) design, and remove the appropriate columns to create an 11-dimensional LHC



- ▶ Thus, results from the complete design, could represent what we might have had from the reduced model *if* it were complete.

Reify (honestly) the RR Model

- ▶ Judgment:
Conditional on equal rainfalls (water input) and input parameters,
 1. More compartments \rightarrow more discharge
 2. Compartment DP behaves similarly to AS
- ▶ So, a reified model (which includes AS) might predict 50% more discharge than the realized model or double the output from DP.
- ▶ We include these judgments by reifying the reduced RR model

General reification of RR Model

$$f^*(x) = g(x)(\beta + \beta^{+*}) + h(x)(A + A^{+*}) + e$$

where, $E[\beta^{+*}] = E[A^{+*}] = 0$ and the variances $\text{Var}[\beta^{+*}]$ and $\text{Var}[A^{+*}]$ are greater than zero. We estimate the variances of β^{+*} and A^{+*} based on $\text{Var}[\beta]$, $\text{Var}[A]$, and pre-specified constants $c_i \in \mathfrak{R}^+$ for $i \in [1, \dots, 4]$:

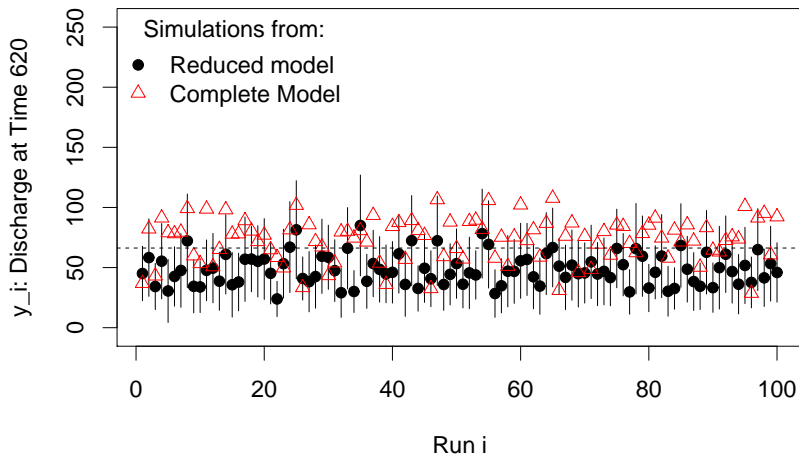
$$\text{Var}[\beta^{+*}] = c_1 \text{Var}[\beta] + c_2$$

$$\text{Var}[A^{+*}] = c_3 \text{Var}[A] + c_4.$$

The size of the constants c_i reflect our judgment that the reified outcomes could equal approximately 1.5 times the realized outcomes.



Graph of general reification of RR Model



56% of the intervals contain the complete model outcomes.

Structure reification of RR Model

► Two options:

1. Add six new parameters to emulator:

$$p^{(\text{new})}, c_f^{(\text{new})}, c_s^{(\text{new})}, k^{(\text{new})}, a^{(\text{new})}, b^{(\text{new})}$$

2. Exaggerate the effect of a compartment (DP) that is both currently in the realized model and expected to behave similarly to the new compartment, and add a k -parameter so that $k^{(\text{DP})} + k^{(\text{GW})} + k^{(\text{new})} = 1$.

► We chose 2. $f'(x, \nu) = f(x) + g^{+'}(x, \nu)\beta^{+'} + e^{+'}(x, \nu)$

$$g^{+'}(x, \nu) = [\nu_1 g(x), \nu_1 \nu_2 (1 - k^{(\text{DP})})]$$

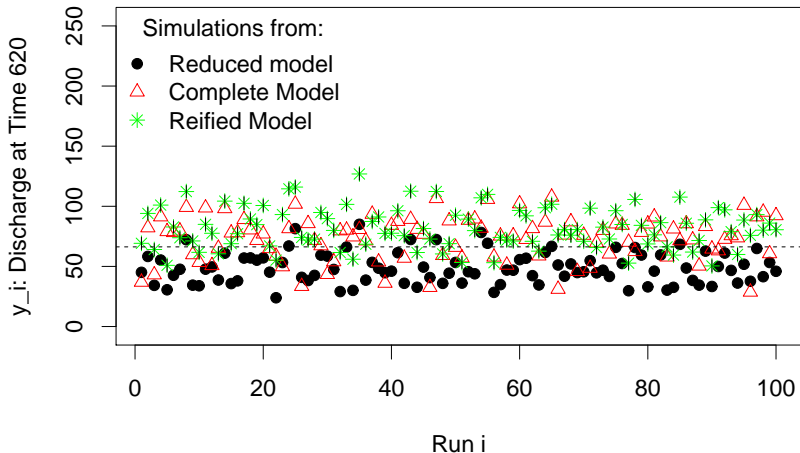
$$e^{+'}(x, \nu) = \nu_1 (\nu_3 e(x) + \delta^+)$$

$$\delta^+ \sim \pi(0, \nu_4)$$

$$\beta_j^{+'} = \begin{cases} 0 & \text{if } x_j \in (p^{(\text{GW})}, c_f^{(\text{GW})}, c_f^{(\text{GW})}, a^{(\text{GW})}, b^{(\text{GW})}) \\ \beta_j + s_j & \text{if } j = 0 \text{ or } x_j \in (p^{(\text{DP})}, c_f^{(\text{DP})}, c_f^{(\text{DP})}, a^{(\text{DP})}, b^{(\text{DP})}, k^{(\text{DP})}) \\ \beta_{k^{(\text{DP})}} + s_j & \text{if } j = 12, \end{cases}$$

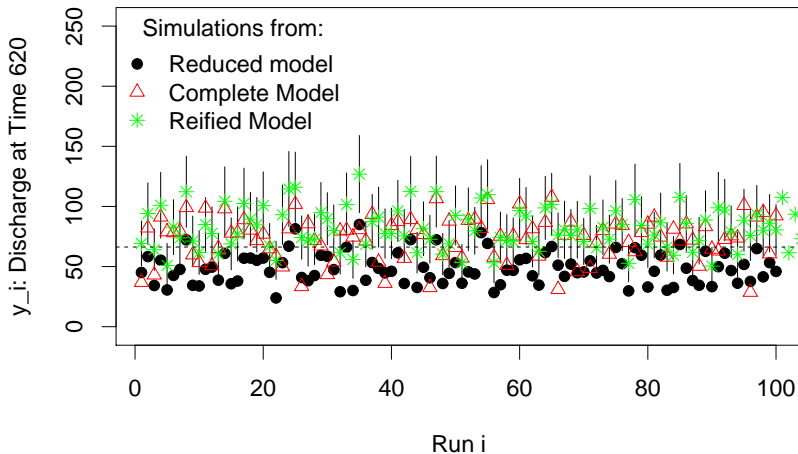
where $E[s_j] = 0$ and $\text{Var}[s_j] = \nu_5$.

Graph of structure reification of RR Model



21% of the intervals contain the three-compartment model outcomes.

Graph of structure and general reification of RR Model



We cover 67% of the complete model outcomes!

Summary/Discussion

- ▶ Computer Models and Reification
- ▶ I propose a way to, possibly, build confidence in reification
- ▶ Possibly justification in it's own right
- ▶ Possibly a check that should be used whenever opting to reify a model
- ▶ If the latter, the model must be accessible
- ▶ Didn't show how to use the reified model for inference
- ▶ Future: Extend to the multivariate outputs

Thank you!

