# Analyzing Complex Models using Data and Statistics

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# Models and assumptions

#### What is a Model?

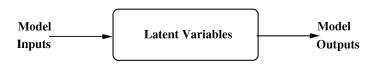
A model is a representation of a postulated relationship among inputs and outputs of a system, usually informed by observation and based on a hypothesis that best explains the relationship.

- models depend on a hypothesis, and,
- models use the data from observation to validate and refine the hypothesis.

# Analysis Process - in predictive mode

We are interested in the general predictive capabilities of the models, related to the outcomes over a whole range.

- ▶ Stage 1: Parameter Ranges  $P_M(p_1,...,p_{N_M}) \sim \bigotimes_{i=1}^{N_M} Unif(a_{i,M},b_{i,M}).$
- ▶ Stage 2: Simulations and Data Gathering



**Model Evaluation (Simulator)** 

Figure: Models and variables

Stage 3: Results Analysis

#### Statistics of latent variables - dominance factors

Dominance factors provide insight into the dominance of a particular latent variable over the others.

### Definition (dominance factors)

Let  $(F_i)_{i \in I}$  be random variables on  $(\Omega, \mathcal{F}, P_M)$ . Then,  $\forall i$ , the dominant variable is defined as:

$$\Phi := \left\{ \begin{array}{ll} \max_i |F_i|, & \text{if not null;} \\ 1, & \text{otherwise.} \end{array} \right.$$

In particular, for each  $j \in I$ , the dominance factors are defined as:

$$p_j := P_M \left\{ \Phi = |F_j| \right\}.$$



# Statistics of latent variables - expected contributions

Random contributions are obtained dividing the latent variables by  $\Phi$ , and hence belong to [0,1].

## Definition (expected contributions)

Let  $(F_i)_{i \in I}$  be random variables on  $(\Omega, \mathcal{F}, P_M)$ . Then,  $\forall i$ , the random contribution is defined as:

$$C_i := \frac{F_i}{\Phi},$$

where  $\Phi$  is the dominant variable. Thus,  $\forall i$ , the expected contributions are defined by  $E[C_i]$ .

# Modeling of geophysical mass flows

The depth-averaged Saint-Venant equations that result are:

$$\frac{\partial h}{\partial t} + \frac{\partial}{\partial x} (h\bar{u}) + \frac{\partial}{\partial y} (h\bar{v}) = 0$$

$$\frac{\partial}{\partial t} (h\bar{u}) + \frac{\partial}{\partial x} \left( h\bar{u}^2 + \frac{1}{2} k g_z h^2 \right) + \frac{\partial}{\partial y} (h\bar{u}\bar{v}) = S_x \qquad (1)$$

$$\frac{\partial}{\partial t} (h\bar{v}) + \frac{\partial}{\partial x} (h\bar{u}\bar{v}) + \frac{\partial}{\partial y} \left( h\bar{v}^2 + \frac{1}{2} k g_z h^2 \right) = S_y$$

Source terms  $S_x$ ,  $S_y$  characterize Mohr-Coulomb (MC), Pouliquen-Forterre (PF) and Voellmy-Salm (VS) models.

# Main assumptions - all the models include curvature effects.

#### Mohr-Coulomb

- Basal Friction based on a constant friction angle.
- ▶ Internal Friction based on a constant friction angle.

#### Pouliquen-Forterre

- Basal Friction is based on an interpolation of two different friction angles, based on the flow regime and depth.
- Normal stress is modified by a hydrostatic pressure force related to the flow height gradient.

#### Voellmy-Salm

- Basal Friction is based on a constant coefficient, similarly to the MC rheology.
- Additional turbulent friction is based on the local velocity by a quadratic expression.

#### Overview of the case studies

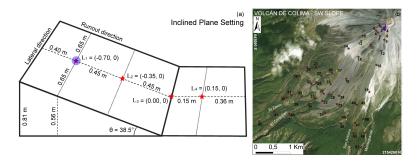


Figure: [Left] Inclined plane description, including local samples sites (red stars). Pile location is marked by a blue dot.[Right](a) Volcán de Colima (México) overview, including 51 numbered local sample sites (stars) and four labeled major ravines channeling the flow. Pile location is marked by a blue dot. Reported coordinates are in UTM zone 13N. Background is a satellite photo.

# Small scale flow - observable outputs

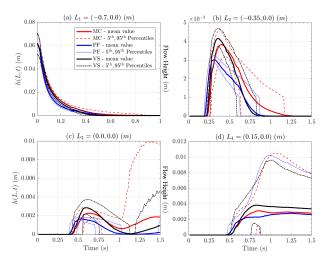


Figure: Flow height in four locations. Bold line is mean value, dashed/dotted lines are  $5^{\rm th}$  and  $95^{\rm th}$  percentile bounds. Different models are displayed with different colors. Plots are at different scale, for simplifying exposition.

# Small scale flow - power integrals

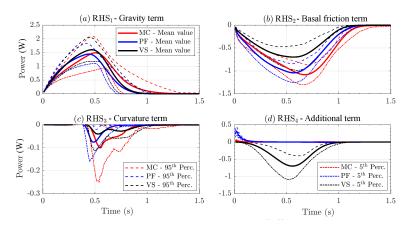


Figure: Spatial integral of the RHS powers. Bold line is mean value, dashed lines are  $5^{\rm th}$  and  $95^{\rm th}$  percentile bounds. Different models are displayed with different colors.

# Large scale flow - proximal to the initial pile

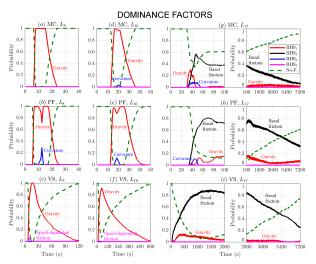


Figure: Dominance factors of **RHS** forces in three locations in the first km of runout. Different models are plotted separately: (a,d,g) assume MC; (b,e,h) assume PF; (c,f,i) assume VS. Different colors correspond to different force terms. No-flow probability is displayed with a green dashed line.

# Large scale flow - proximal to the initial pile

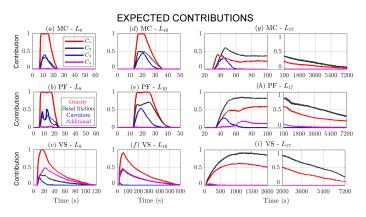


Figure: Expected contributions of **RHS** forces in three locations in the first km of runout. Different models are plotted separately: (a,d,g) assume MC; (b,e,h) assume PF; (c,f,i) assume VS. Different colors correspond to different force terms.

# Large scale flow - distal from the initial pile

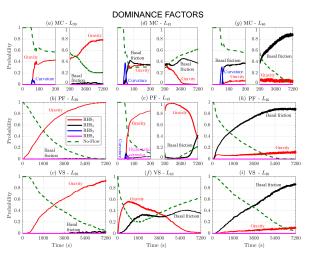


Figure: Dominance factors of RHS forces in three locations after 2 km of runout. Different models are plotted separately: (a,d,g) assume MC; (b,e,h) assume PF; (c,f,i) assume VS. Different colors correspond to different force terms. No-flow probability is displayed with a green dashed line.

## Large scale flow - distal from the initial pile

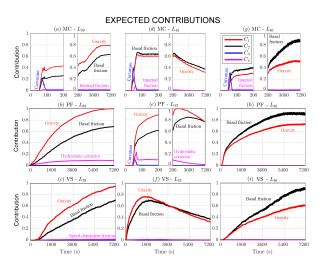


Figure: Expected contributions of **RHS** forces in three locations after 2 km of runout. Different models are plotted separately: (a,d,g) assume MC; (b,e,h) assume PF; (c,f,i) assume VS. Different colors correspond to different force terms.

## Large scale flow - flow extent and spatial integrals

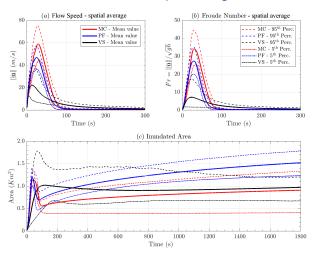


Figure: Comparison between spatial averages of (a) flow speed, and (b) Froude Number, in addition to the (c) inundated area, as a function of time. Bold line is mean value, dashed/dotted lines are  $5^{\rm th}$  and  $95^{\rm th}$  percentile bounds. Different models are displayed with different colors.

#### Conclusions

- ▶ our approach evaluates the statistics of a range of simulations, produced by the couple  $(M, P_M)$ .
- the new statistical framework processes the mean and the uncertainty range of either observable or latent variables in the simulation.
- analysis is performed at selected sites, and spatial integrals were also performed, illustrating the characteristics of the entire output.
- the new concepts of dominance factors and expected contributions, enable a simplified description of the local dynamics.