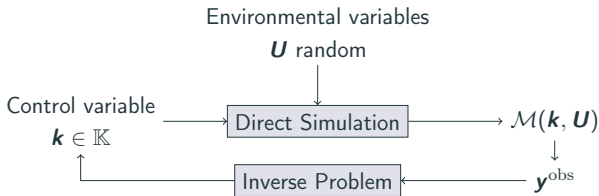


Parameter control in the presence of uncertainties

- \mathbf{k} : calibration parameter \rightarrow epistemic uncertainty
- \mathbf{U} : environmental/uncertain parameter \rightarrow aleatoric uncertainty



The misfit between some observations and the output of the model is $J(\mathbf{k}, \mathbf{U})$, that needs to be minimized with respect to \mathbf{k} .

Robustness under parametric model misspecification

How to get an estimate $\hat{\mathbf{k}}$ that is robust with respect to the inherent variability of \mathbf{U} ?

Relative regret

Main idea: For each $\mathbf{u} \sim \mathbf{U}$, *compare* the value of the cost function to its optimal value $J^*(\mathbf{u})$ and define $\mathbf{k}^*(\mathbf{u}) = \arg \min_{\mathbf{k} \in \mathbb{K}} J(\mathbf{k}, \mathbf{u})$:

- Either set $\alpha \geq 1$, and define the probability of being α -optimal

$$\Gamma_{\alpha}(\mathbf{k}) = \mathbb{P}_{\mathbf{U}} [J(\mathbf{k}, \mathbf{U}) \leq \alpha J^*(\mathbf{U})] \quad (1)$$

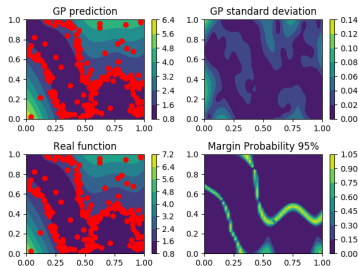
and maximize it

- or for a given $0 < p \leq 1$, find the quantile of order p of the r.v. $J(\mathbf{k}, \mathbf{U})/J^*(\mathbf{U})$ and minimize it

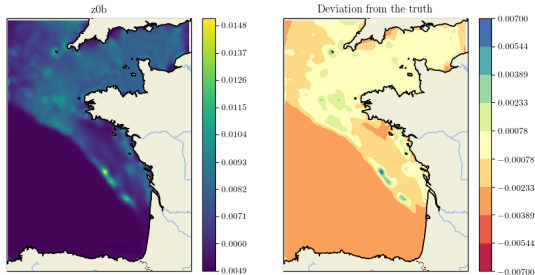
Puts more importance in the estimation in the region where good performances are possible (small $J^*(\mathbf{u})$)

Numerical methods

- Computer code is expensive to run
 - Even for a given \mathbf{u} , getting $J^*(\mathbf{u})$ can be tricky
- Need for specific methods to
- gather information on $J(\mathbf{k}, \mathbf{u})$ prior to the evaluation
 - gather information on $J^*(\mathbf{u})$ prior to the optimization
 - improve precision on the estimation of the probability of coverage Γ_α
 - improve precision on the estimation of the quantiles of J/J^*



Application: estimation of the bottom friction in the Atlantic Ocean



- High-dimensional problem: $\mathbf{k} \in \mathbb{R}^{\sim 20000}$: need to reduce dimension
- Sensitivity to the number of number of tides components added