



# Robust Adaptation of the Train Speed for Energy Saving under Punctuality and Security Constraints

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## Objective of the study

Find deterministic driver's command  $u^*$  for a given entry  $\mathcal{E}$ , minimizing the energy consumed  $F^E$  in a specific domain  $\mathcal{D}$  (respecting security, punctuality, and comfort constraints), robust to uncertainties  $\mathbf{X}$ . It aims to help drivers and prepare the development of autonomous trains.

$$u^*(\mathcal{E}) = \arg \min_{u \in \mathcal{D}} \mathbb{E}_{\mathbf{X}} [F^E(u, \mathbf{X}, \mathcal{E})] \quad (1)$$

## Models for High-Speed Trains

- Entry  $\mathcal{E}$ : Wind and track descriptions.
- Driver's command  $u$ : Time-dependent in  $[-1, 1]$ .
- Longitudinal dynamics for the whole train:

$$M_T k^r \ddot{Y}(t) = \sum_{\alpha} F^{\alpha}(u(t), \mathbf{X}, \mathcal{E}) \quad (2)$$

with  $M_T$  the mass of the train,  $k^r$  a factor including the wheels rotation,  $\ddot{Y}$  the train acceleration.  $F^{\alpha}$  regroups traction and braking forces, the Davis or resistant force, a corrective force in curve, and the weight.

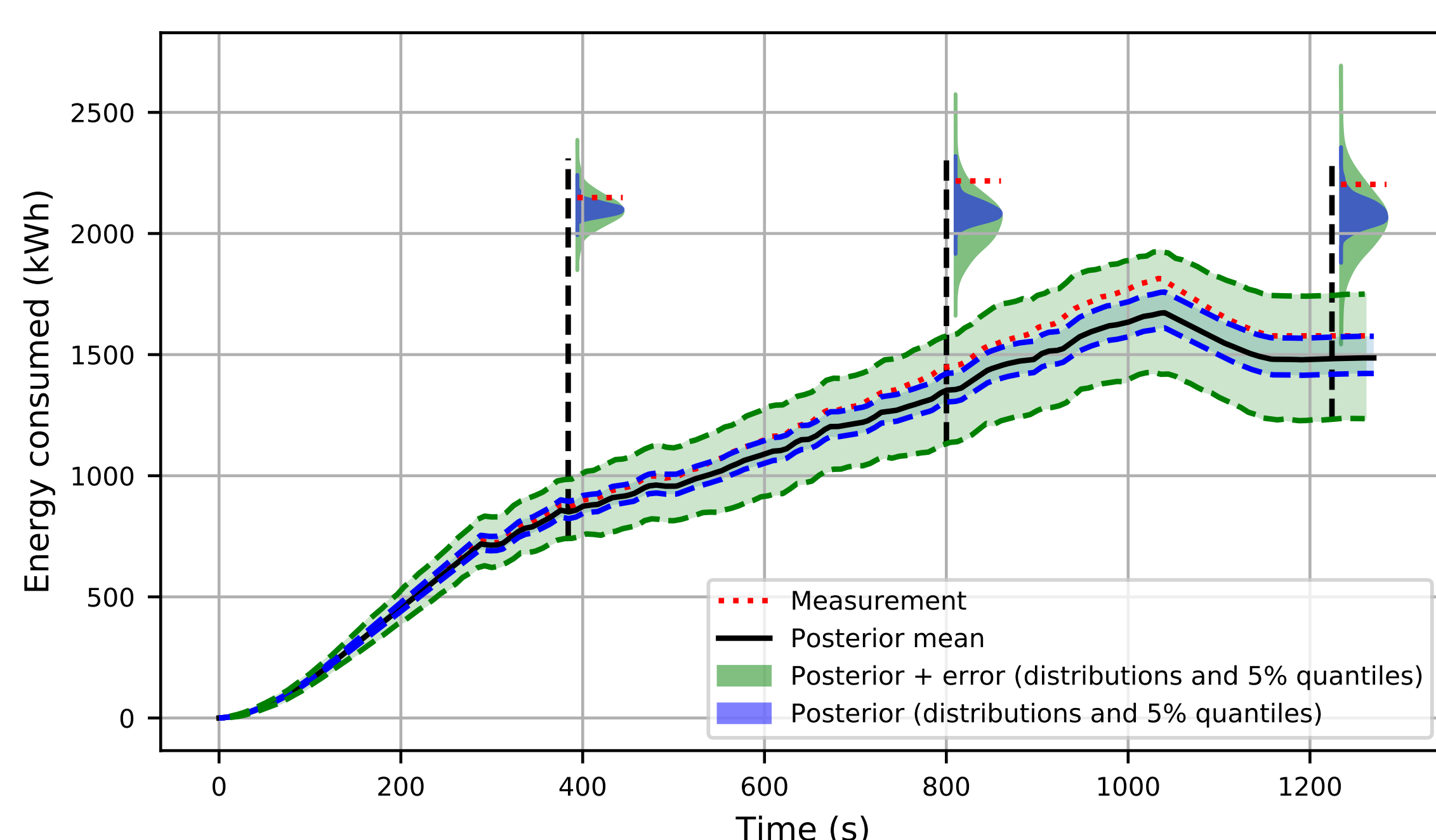
- Energy consumed by the train:

$$F^E(u, \mathbf{X}, \mathcal{E}) = \int_{t_s}^{t_f} P^E(u(t), \mathbf{X}, \mathcal{E}) dt, \quad (3)$$

where  $P^E$  is the electric power and  $t_s$ ,  $t_f$  are the initial and arrival time.

## Modeling the Uncertainties

- Sensitivity analysis on the uncertain parameters  $\mathbf{X}$ .
- 4 dynamic and 5 energy consumption parameters.
- Prior distribution: Describe the available information.
- Likelihood function: Quantify the plausibility of a simulation with measurements on commercial trains.
- Markov Chain Monte Carlo (MCMC): Metropolis-within-Gibbs to reach the posterior distribution.



## First optimization method

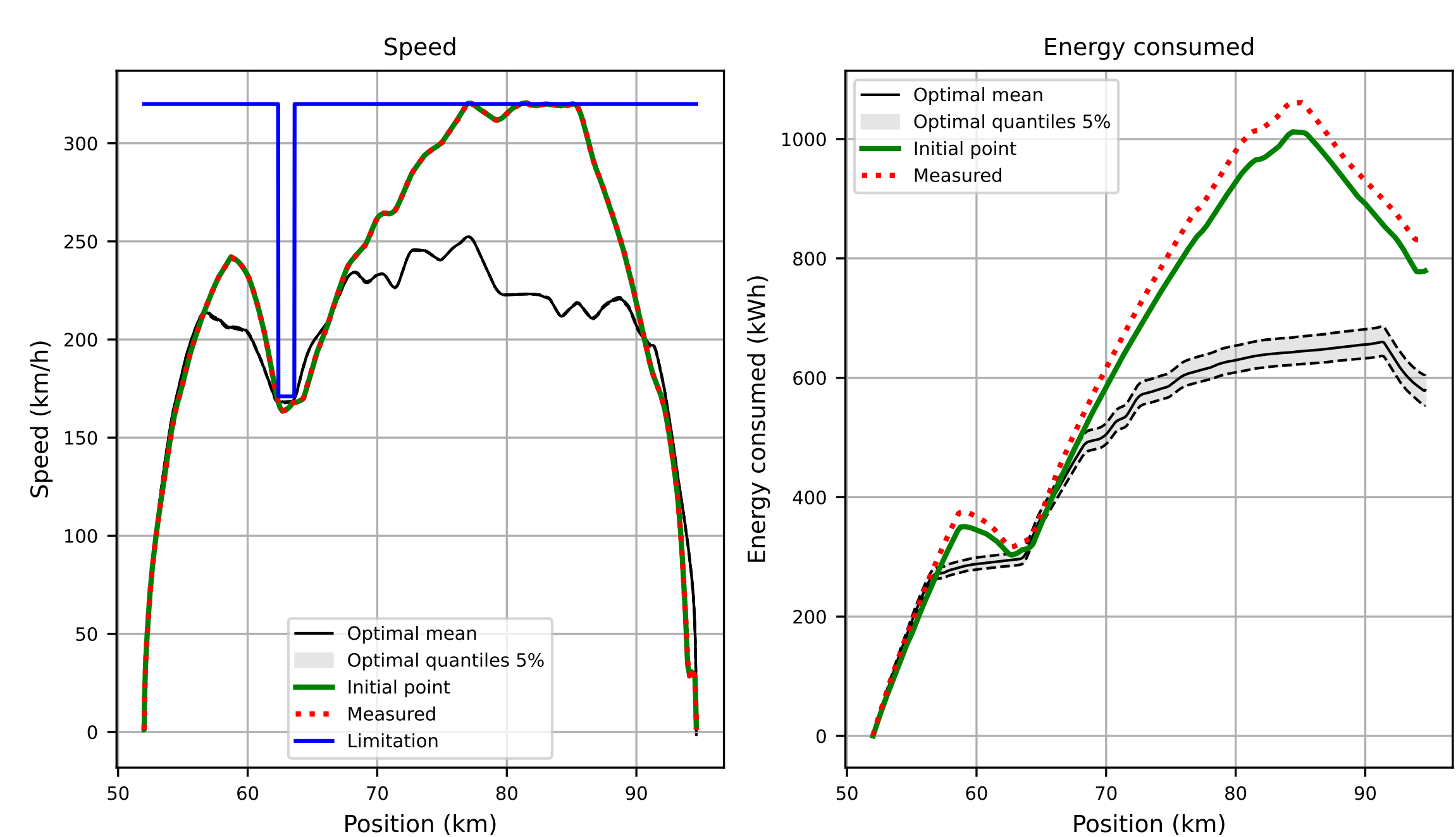
- Idea: Find a family  $(u_k^*)_{\mathbf{X}_k}$  that minimizes the energy consumed in specific configurations  $\mathbf{X}_k$  drawn in  $\mathbf{X}$ .
- Cost function  $f^E$ : Minimize the energy consumed.
- Constraints  $\mathcal{C}$ : Security, punctuality, and comfort.
- Method: Discretization, penalization with augmented Lagrangian, and CMA-ES iterative algorithm.

$$u_k^*(\mathcal{E}, \mathbf{X}_k) = \arg \min_{u \in \mathcal{C}} f^E(u, \mathbf{X}_k, \mathcal{E}) \quad (4)$$

## Second optimization method

- Idea: Find a deterministic  $u^*$  easy to transform to be robust to the fluctuations of  $\mathbf{X}$ .
- Cost function: Minimize the mean of the energy.
- Constraints: Transform  $u^*$  as  $\tilde{u}^*(\mathbf{X}_k)$  to be robust to  $\mathbf{X}$ .
- Method: Two discretizations, Principal Component Analysis, and CMA-ES iterative algorithm.

$$u^* = \arg \min_{u \in \mathcal{D}} \frac{1}{K} \sum_{k=1}^K F^E(\tilde{u}(\mathbf{X}_k), \mathbf{X}_k, \mathcal{E}). \quad (5)$$



## References

- [1] Nespoulous J. et al. Optimisation of train speed to limit energy consumption. *Veh. Syst. Dyn.*, 2021.