

MPC Robustness Demo

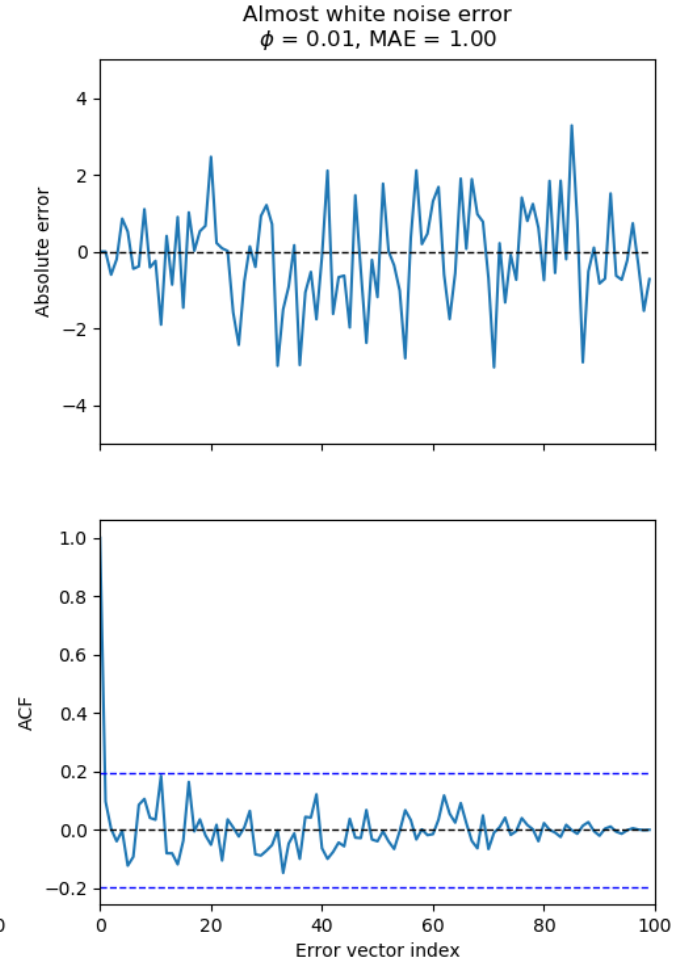
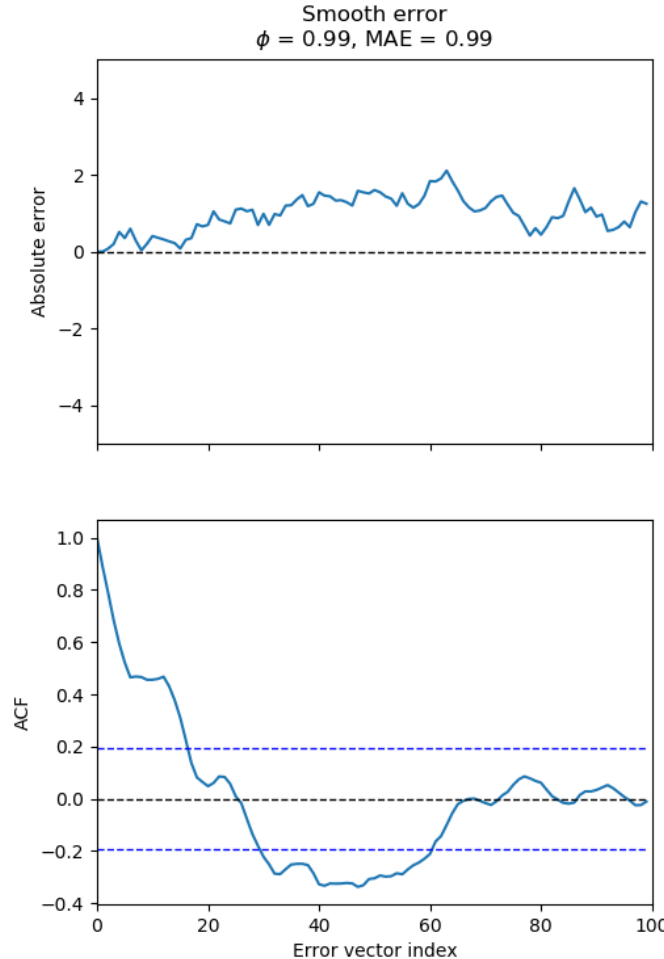
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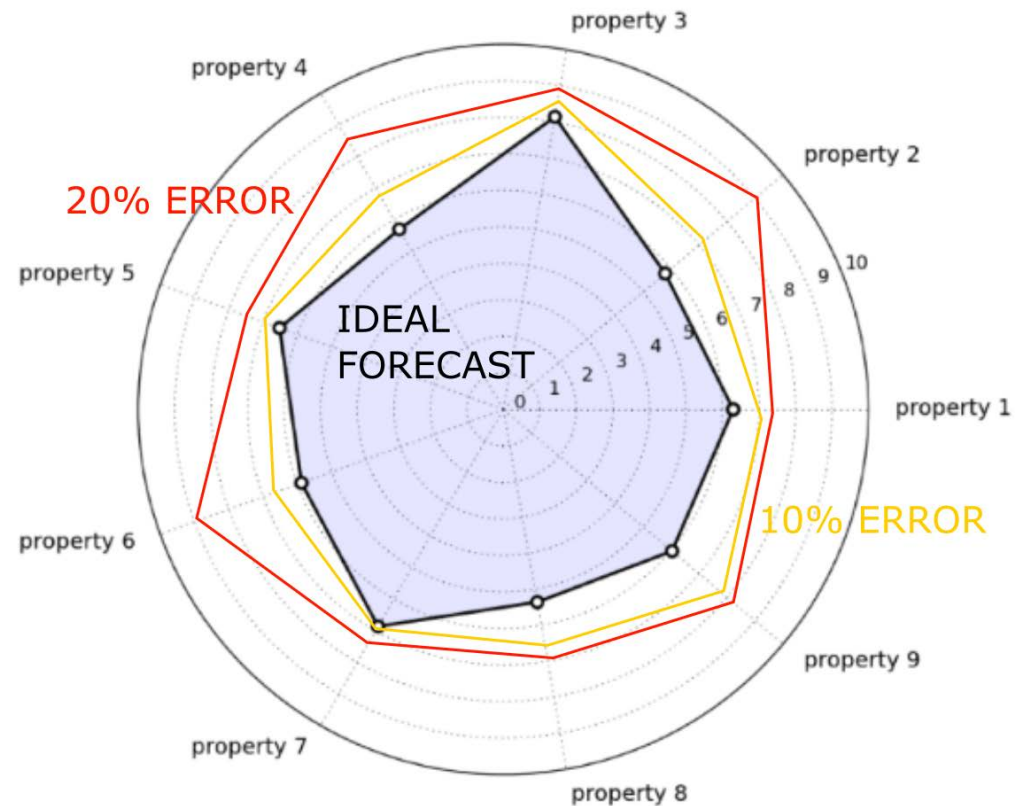
Python Package

- <https://github.com/krzysztofarendt/forecast-gen>
- Function generating error signals:
 - controllable autocorrelation
 - controllable mean absolute error (MAE)
 - controllable length (number of elements in a vector)



Integration With KPIs: Concept

Radar plot



Demo Case

- R1C1 model calibrated to a thermal zone in the OU44 building (SDU Campus, Odense, Denmark)
- Controllable input: - heating/cooling signal u
- Objective: - minimize energy consumption
- Constraints: - indoor temperature stay within (T_{min}, T_{max})
 - u within $(-100, 100)$
- Prediction horizon: - 6h
- MPC solver: - multiple shooting method,
 <https://github.com/sdu-cfei/mshoot>,
 paper presenting the solver accepted to BS2019
- Control model using inaccurate forecasts for outdoor temperature, solar radiation, and occupancy

Demo Case

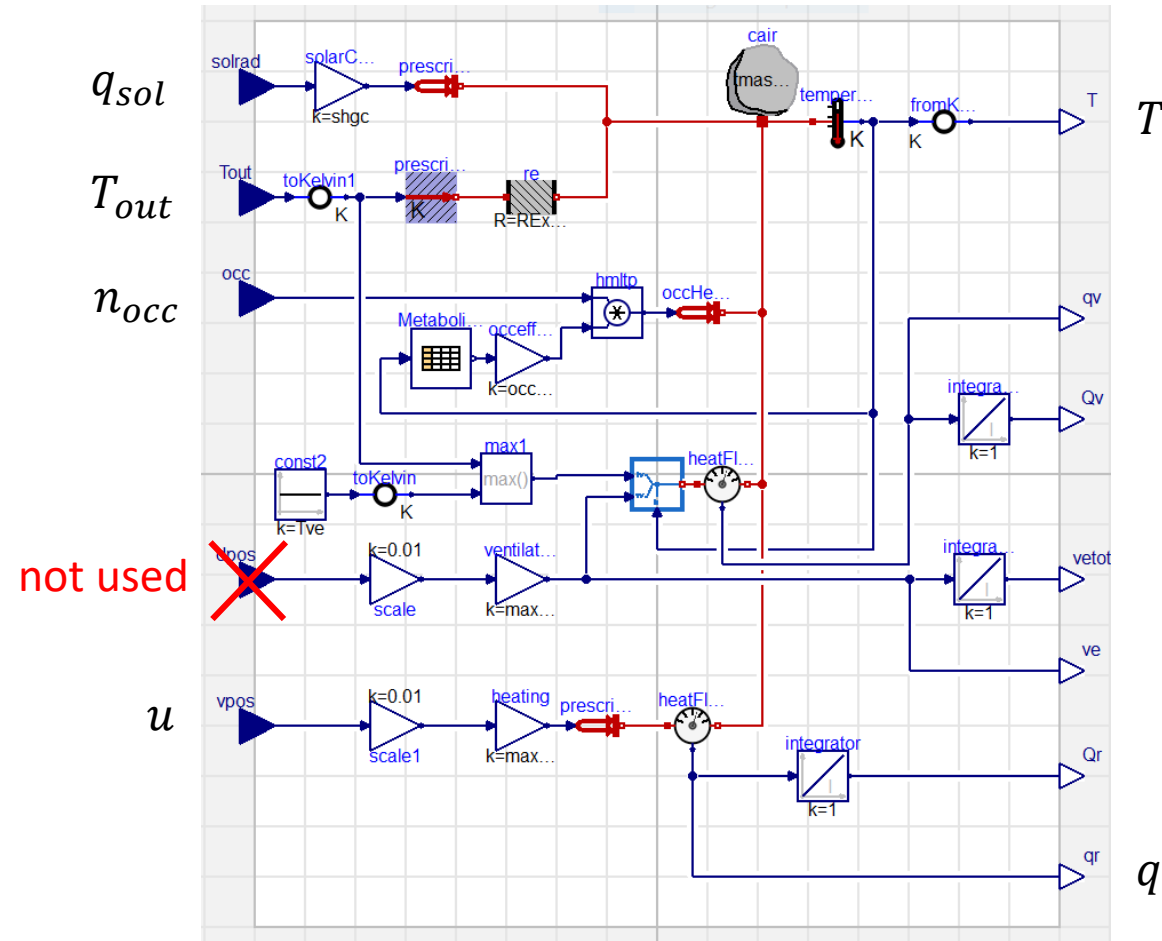
Optimization problem:

$$\arg \min_u \sum_{i=1}^N q_i^2$$

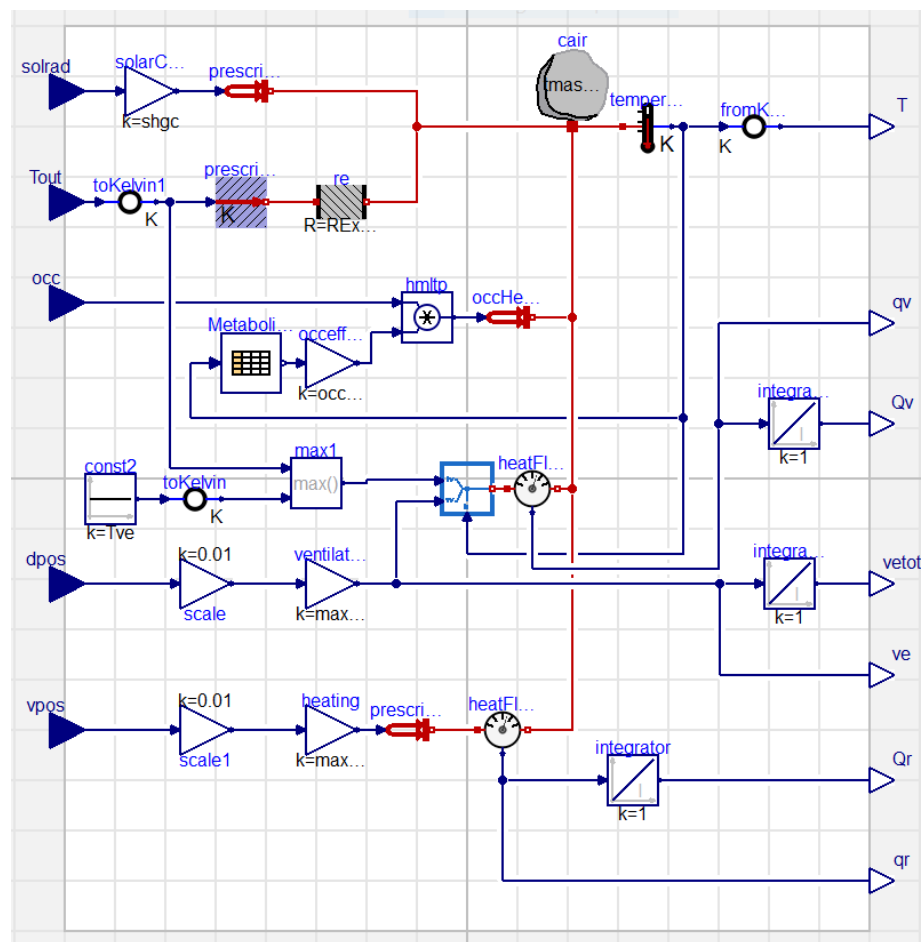
s. t.

$$T_{min,i} < T_i < T_{max,i}$$

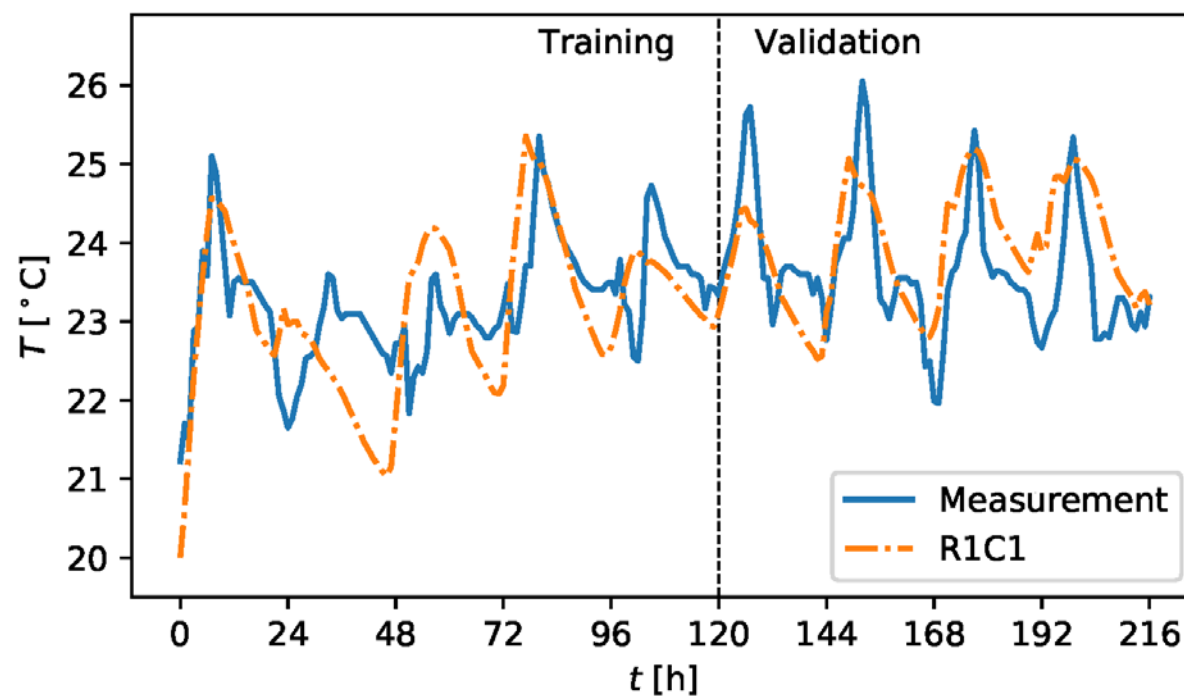
$$-100 < u_i < 100$$



Demo Case



R1C1 model

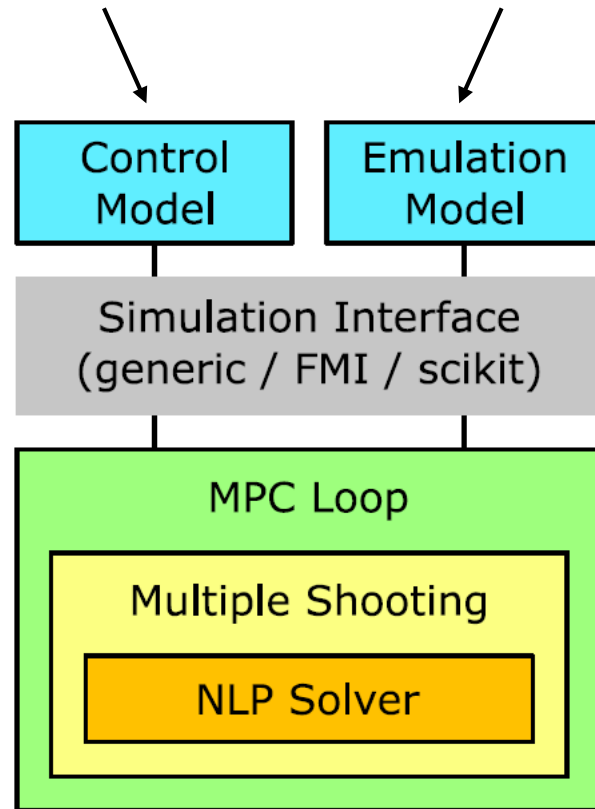


Model validation
(not very accurate, but physically reasonable)

Demo Case

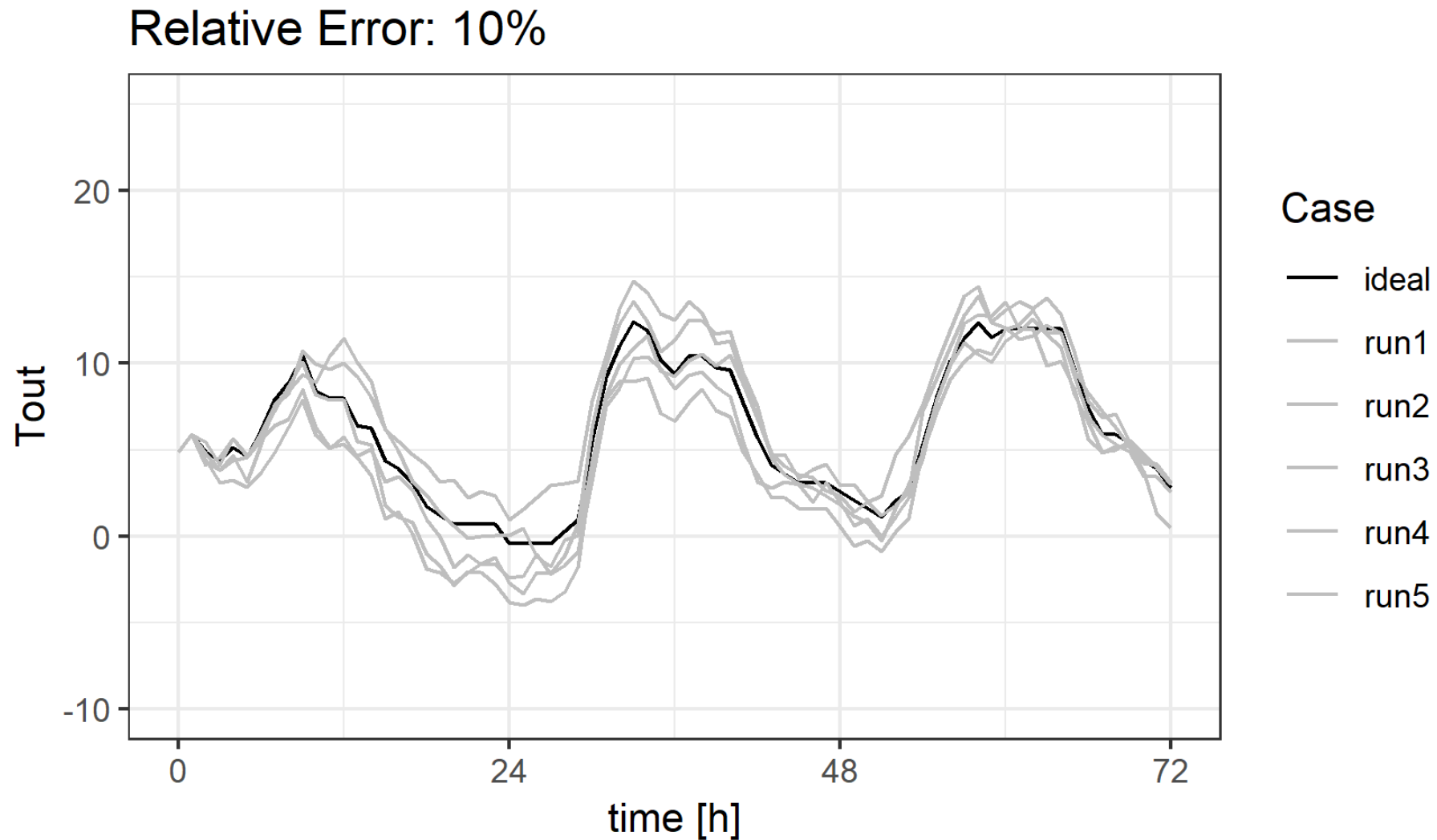
R1C1 + inaccurate forecasts
(rel. errors: 0%, 10%, 20%, 30%)

R1C1 + ideal forecasts

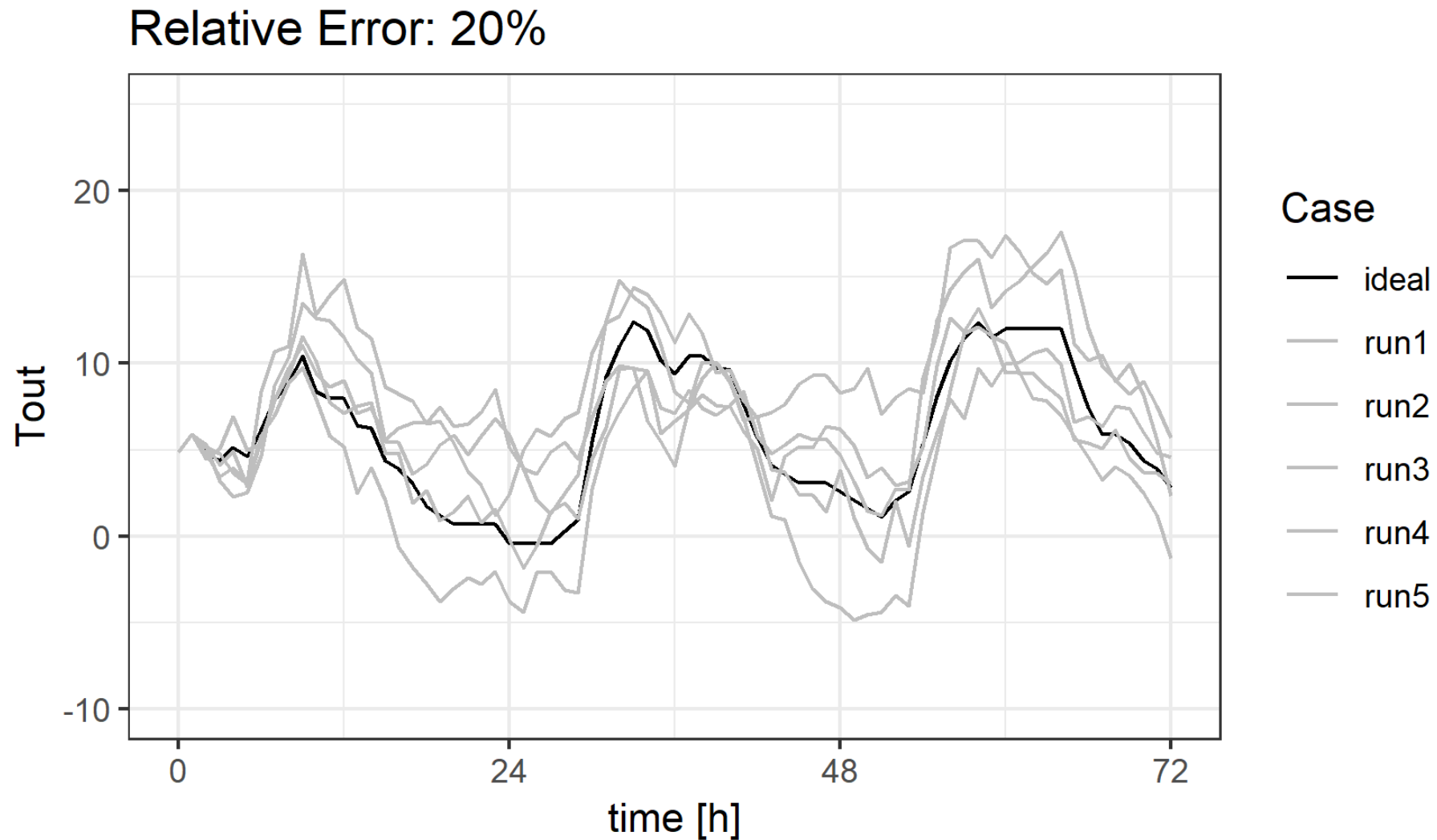


× 5 runs per each error level
(random errors in each run)

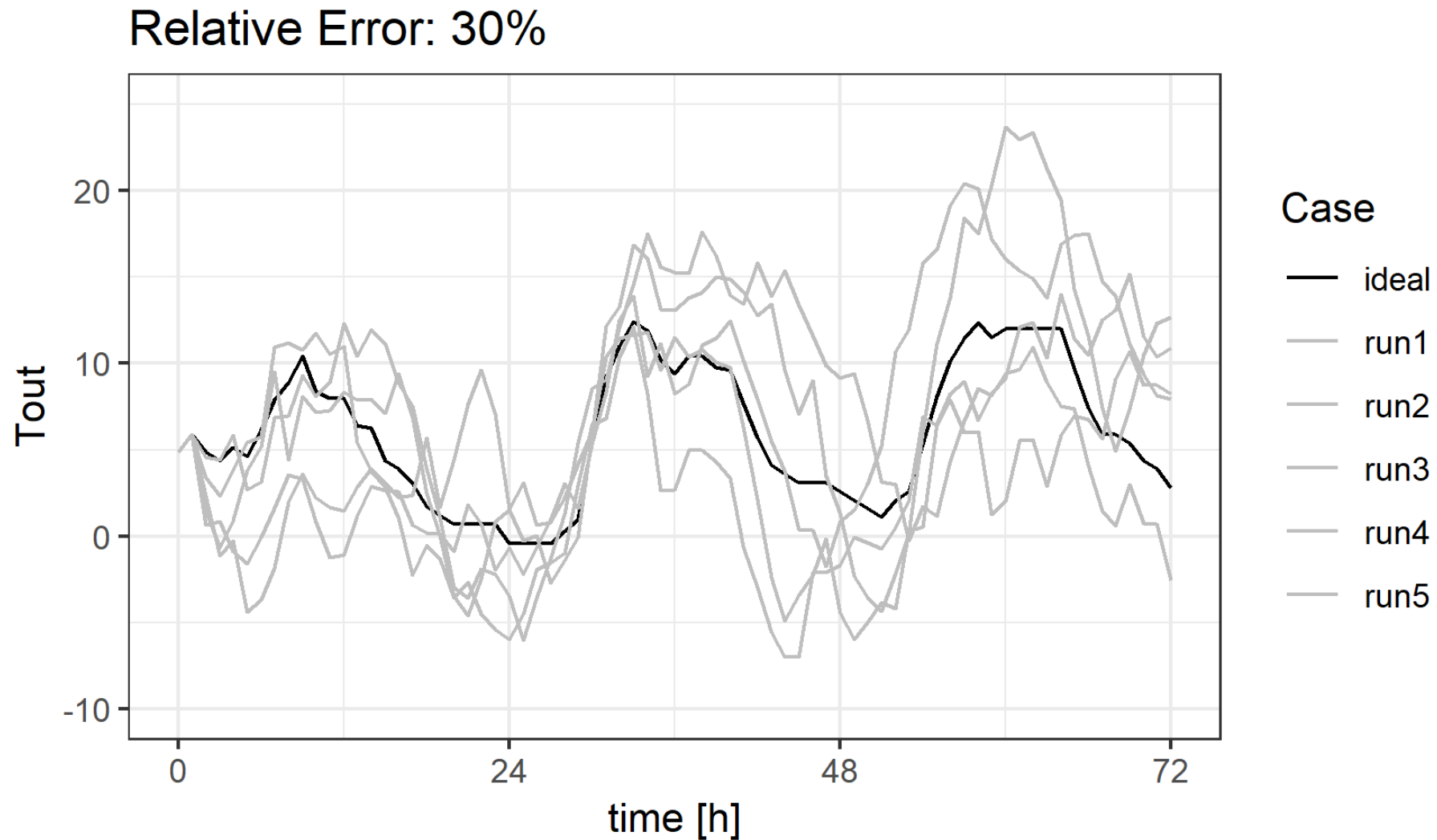
Inaccurate Forecasts (T_{out} profiles)



Inaccurate Forecasts (T_{out} profiles)



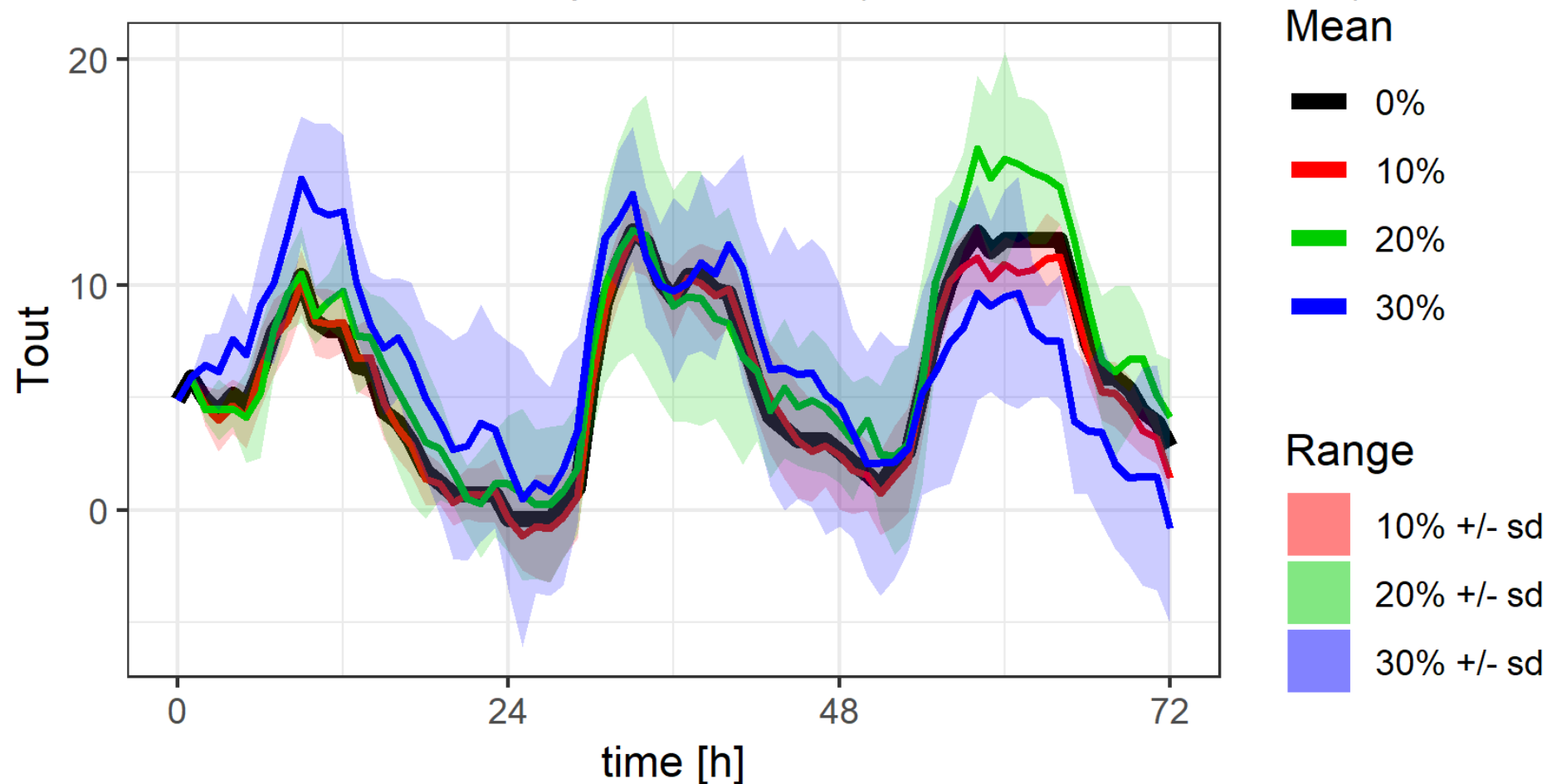
Inaccurate Forecasts (T_{out} profiles)



Inaccurate Forecasts: Ranges

Outdoor temperature

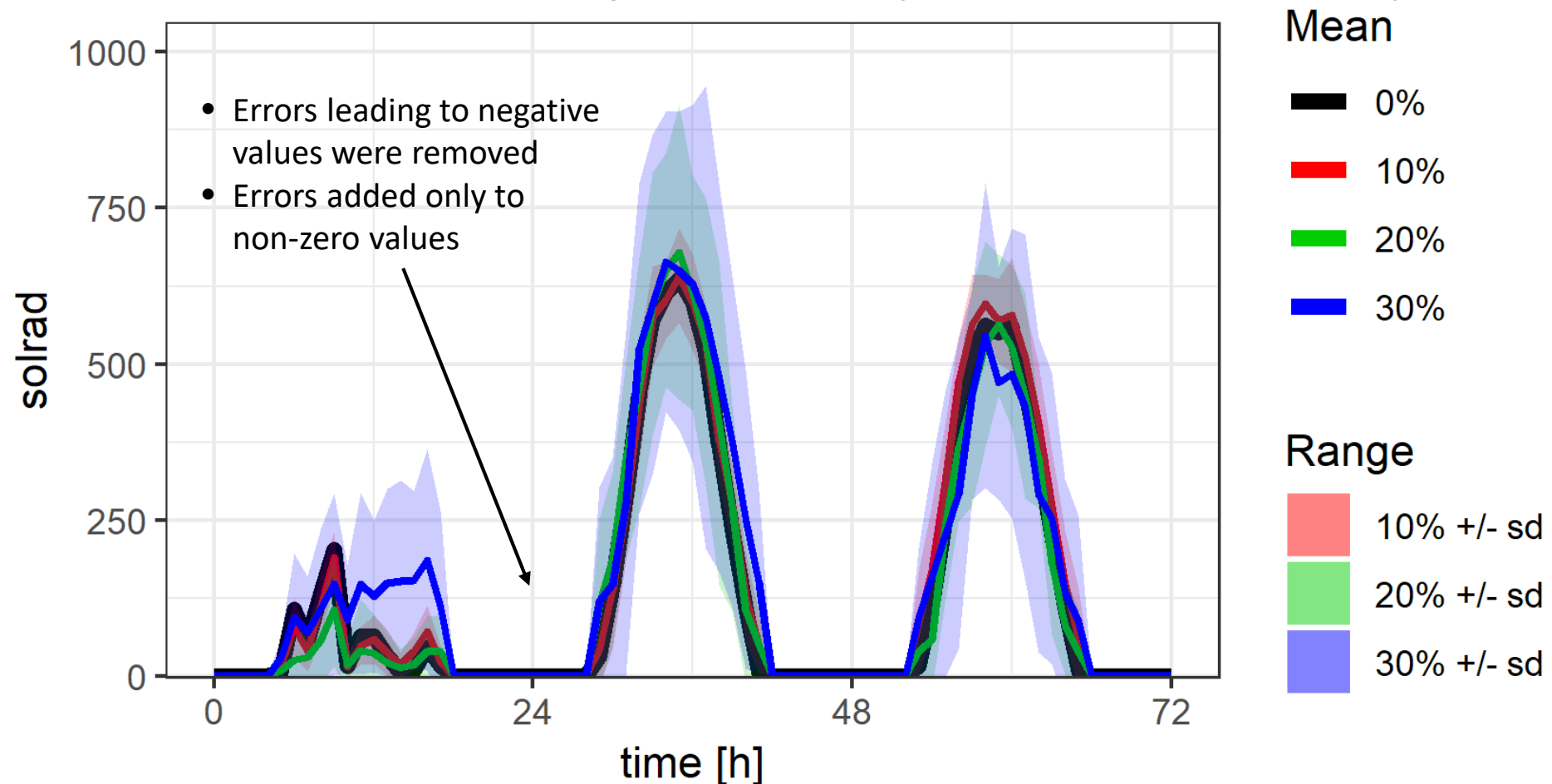
Based on 5 MPC runs per error level (random error in each run)



Inaccurate Forecasts: Ranges

Solar radiation

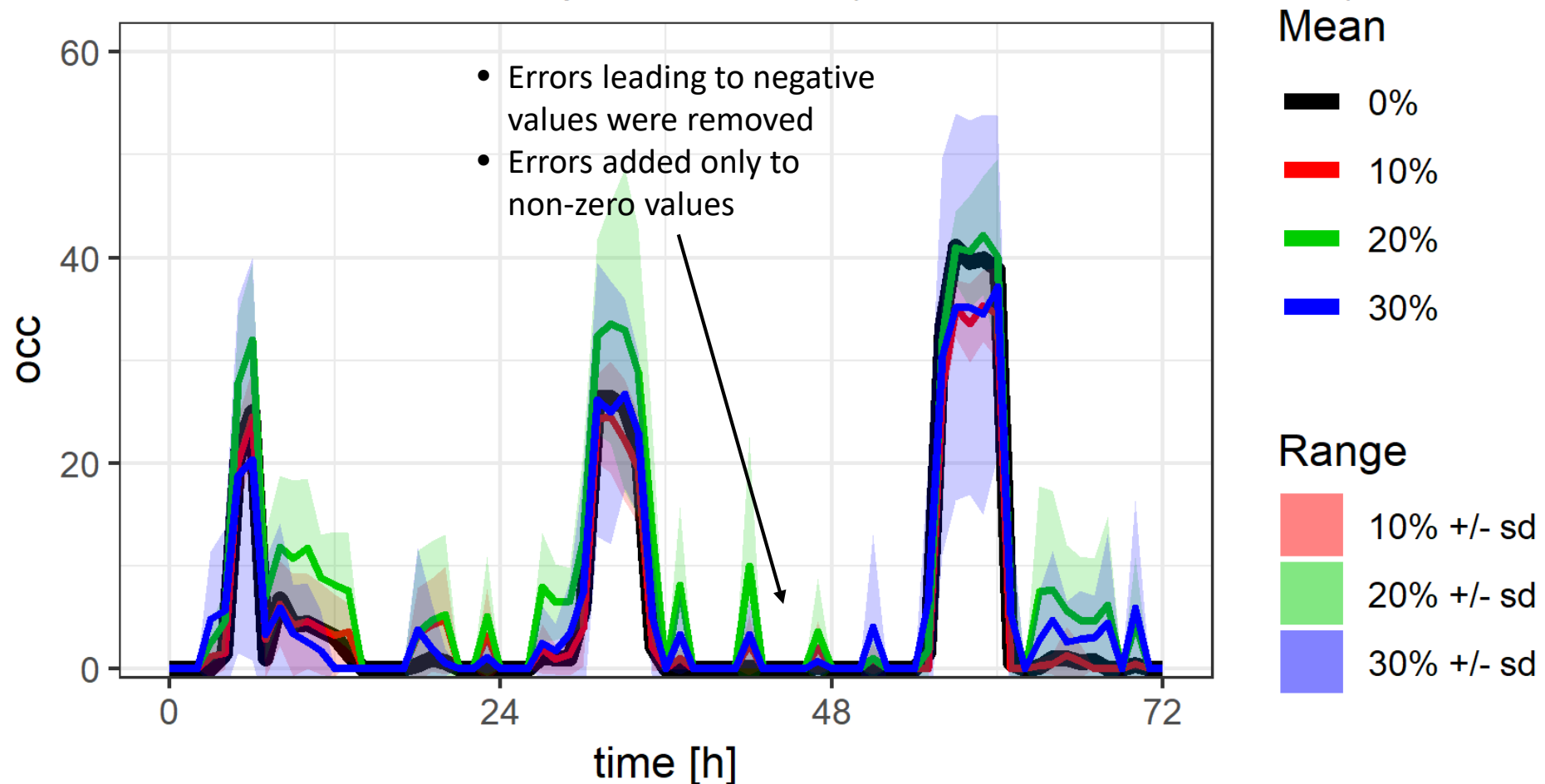
Based on 5 MPC runs per error level (random error in each run)



Inaccurate Forecasts: Ranges

Occupancy schedules

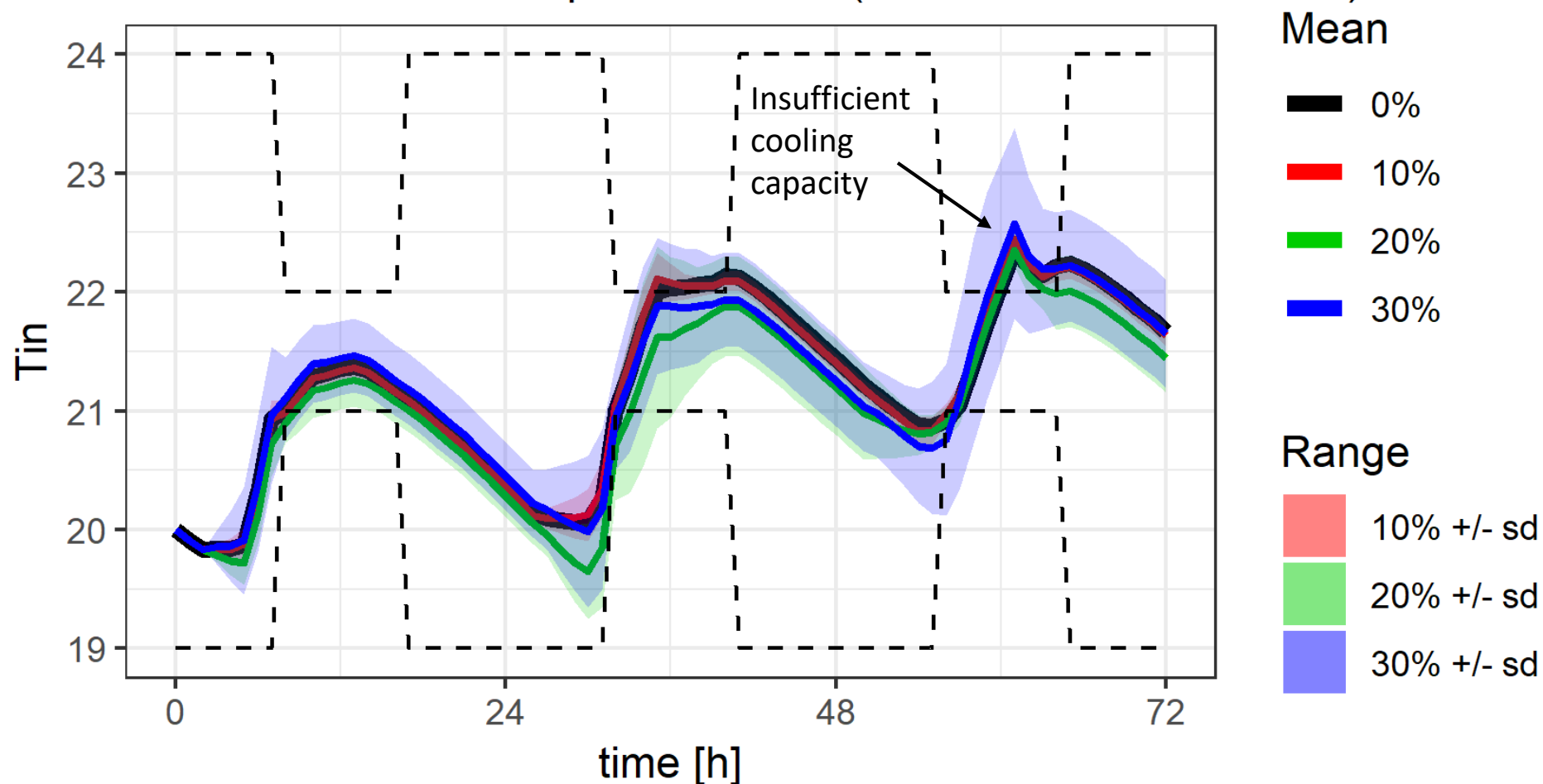
Based on 5 MPC runs per error level (random error in each run)



Results: Indoor Temperature

Indoor temperature vs. constraints

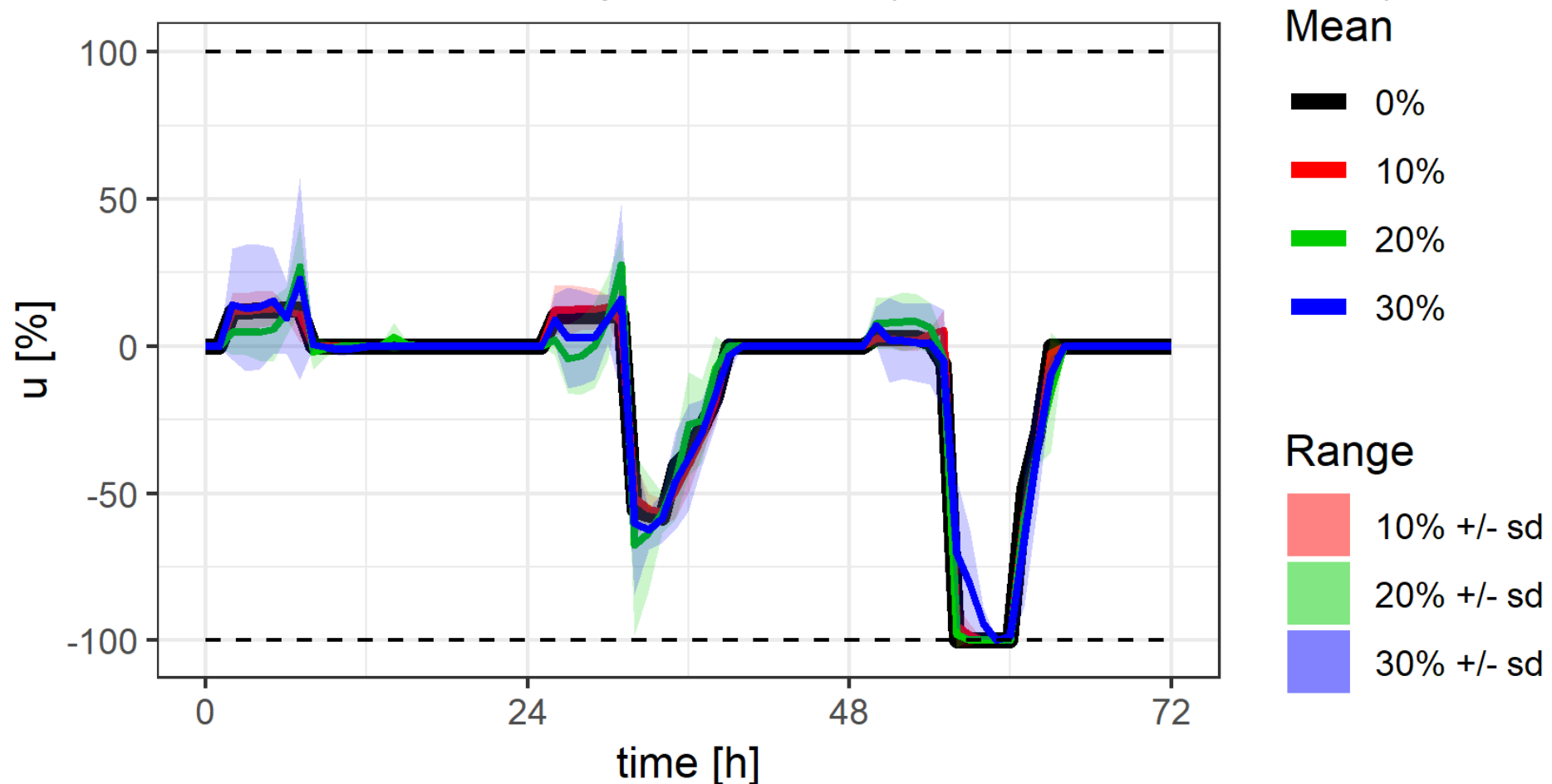
Based on 5 MPC runs per error level (random error in each run)



Results: Heating/Cooling Signal

Heating/cooling signal

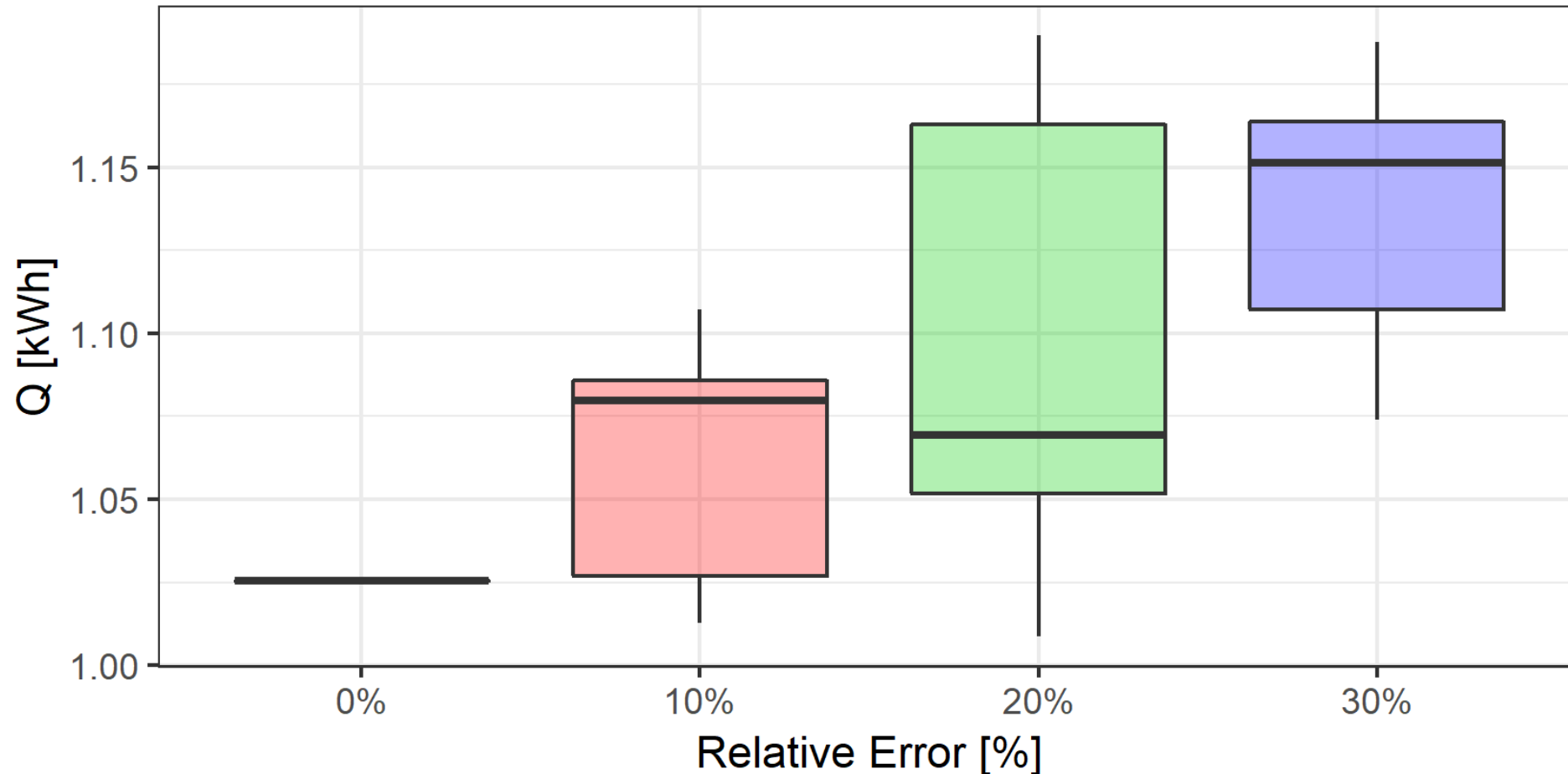
Based on 5 MPC runs per error level (random error in each run)



Results: Energy

Energy demand Q vs. forecast relative error

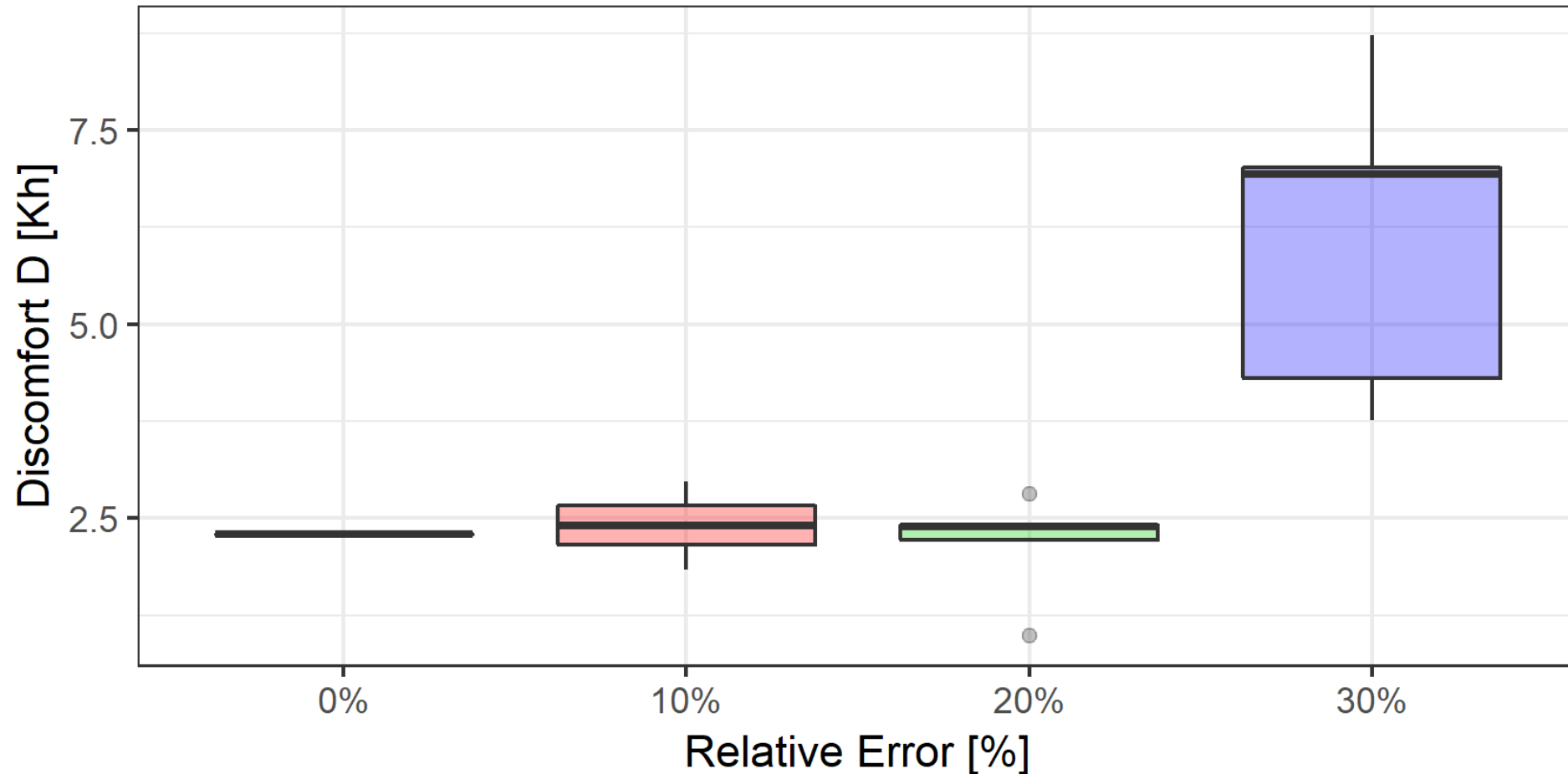
Based on 5 MPC runs per error level (random error in each run)



Results: Discomfort

Discomfort vs. forecast relative error

Based on 5 MPC runs per error level (random error in each run)



Summary

Importance with respect to BOPTEST:

- The higher error level, the worse MPC performance.
- The higher error level, the higher variation in the performance.
- Some MPC formulations are specifically designed to deal with uncertainty (e.g. Robust MPC). Using only accurate forecasts puts these approaches into a disadvantageous position.

Limitations of the analysis:

- The control model used the same static forecast for all optimization horizons. New forecasts should be made for each horizon instead.
- The current implementation of error generator is not suitable for variables which cannot go below zero (irradiance, occupancy).