

Example Development and Evaluation of MPC and RL Controllers

Workshop: Introduction to the BOPTEST Framework for Testing and
Benchmarking Advanced Controllers

RLEM

November 12, 2023

The logo for KU Leuven, featuring the text "KU LEUVEN" in white capital letters on a blue rectangular background.

KU LEUVEN

The logo for Thermal Systems Simulation, with the word "THERMAL" in small grey letters to the left of a large stylized "S" that is red on top and blue on the bottom. To the right of the "S" are the words "SYSTEMS" and "SIMULATION" in grey.

THERMAL **S**YSTEMS
SIMULATION

The logo for vito, featuring a stylized orange and blue swoosh to the left of the word "vito" in a bold, black, sans-serif font.

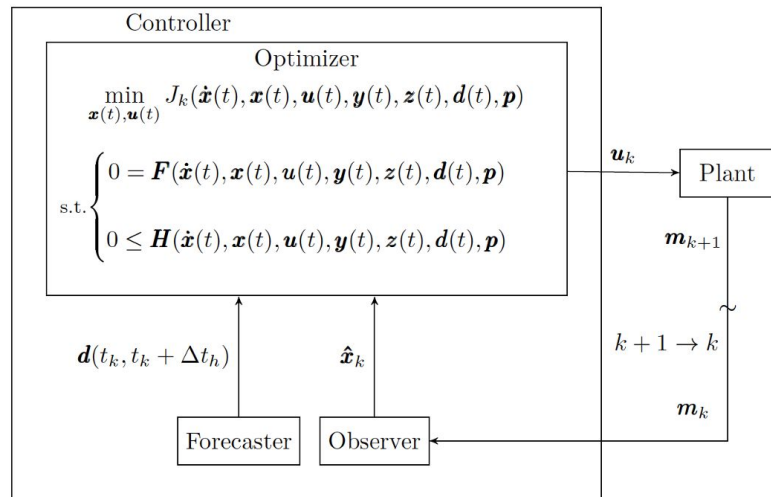
vito



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MPC working principle

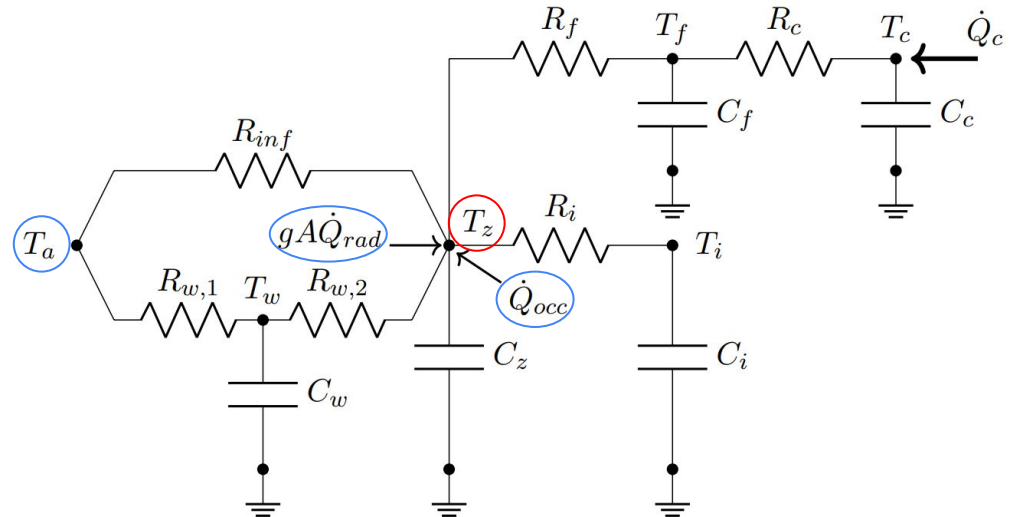
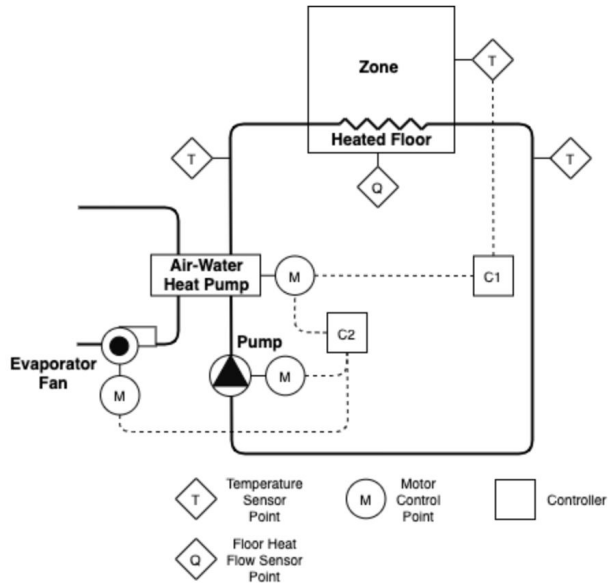


MPC: Optimizes the **future** from domain knowledge

$$J_k = \int_{t=t_k}^{t_k+\Delta t_k} l(\dot{\mathbf{x}}(t), \mathbf{x}(t), \mathbf{u}(t), \mathbf{y}(t), \mathbf{z}(t), \mathbf{d}(t), \mathbf{p}) dt$$

The diagram illustrates the cost function J_k as an integral of a loss function l over time. The loss function l is composed of two terms: **cost** (red) and **w * discomfort** (blue), where w is a weight factor.

System identification



$$\dot{Q}_c = (a_c + b_c(T_c - T_{c,n}) + c_c(T_a - T_{a,n}))k_e u_{hp}$$

$$\dot{Q}_e = (a_e + b_e(T_c - T_{c,n}) + c_e(T_a - T_{a,n}))k_e u_{hp}$$

$$P_{hp} = \dot{Q}_c - \dot{Q}_e$$

$$COP = \frac{\dot{Q}_c}{P_{hp}},$$

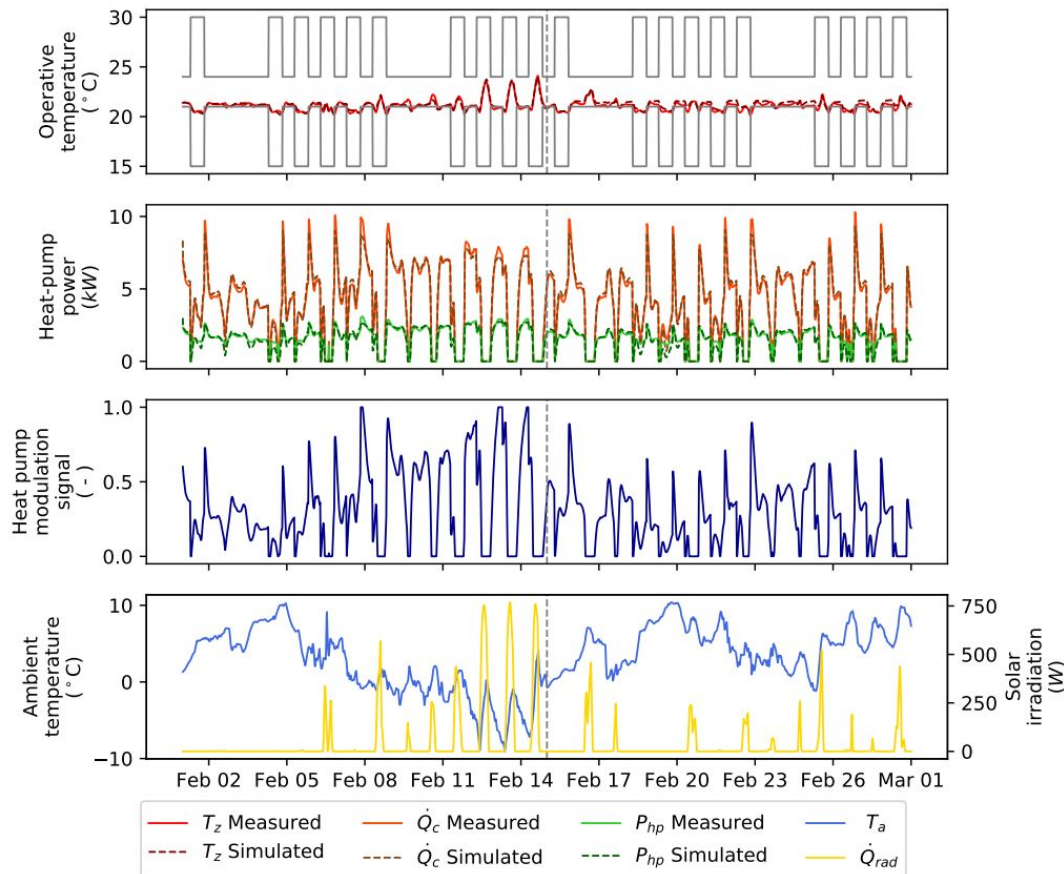
System identification

```

1 import requests
2
3 # url for the BOPTTEST service
4 url = "http://boptest-workshop.net"
5
6 # Select test case and get identifier
7 testcase = "bestest_hydrionic_heat_pump"
8 testid = \
9 requests.post("{0}/testcases/{1}/select".format(url, testcase)).json()["payload"]["testid"]
10
11 # Find all measurements and inputs of this emulator
12 inputs = requests.get("{0}/inputs/{1}".format(url, testid)).json()["payload"]
13 measurements = requests.get("{0}/measurements/{1}".format(url, testid)).json()["payload"]
14 all_points = measurements.keys() + inputs.keys()
15
16 # Set the emulator in the desired simulation period and initialize
17 requests.put("{0}/initialize/{1}".format(url, testid),
18             json={"start_time": 31*24*3600,
19                  "warmup_period": 7*24*3600}).json()["payload"]
20
21 # Simulate with baseline control for one month
22 for _ in range(28*24):
23     requests.post("{0}/advance/{1}".format(url, testid),
24                 json={}).json()["payload"]
25
26 # Gather data
27 res = requests.put("{0}/results/{1}".format(url, testid),
28                   json={"point_names": all_point,
29                        "start_time": int(0),
30                        "final_time": int(3.1536e7)}).json()["payload"]

```

The **Grey-Box Toolbox** [4] is used to prototype the model and train its parameters



MPC description

- Controlled variable: zone operative temperature
- Control variable: modulation signal for HP compressor frequency
- BOPTEST deterministic forecast
- Prediction horizon: 3, 6, 12, **24**, 48 hours
- Control step: 15, **30**, 60 minutes
- Direct collocation with JModelica
- Unscented Kalman filter

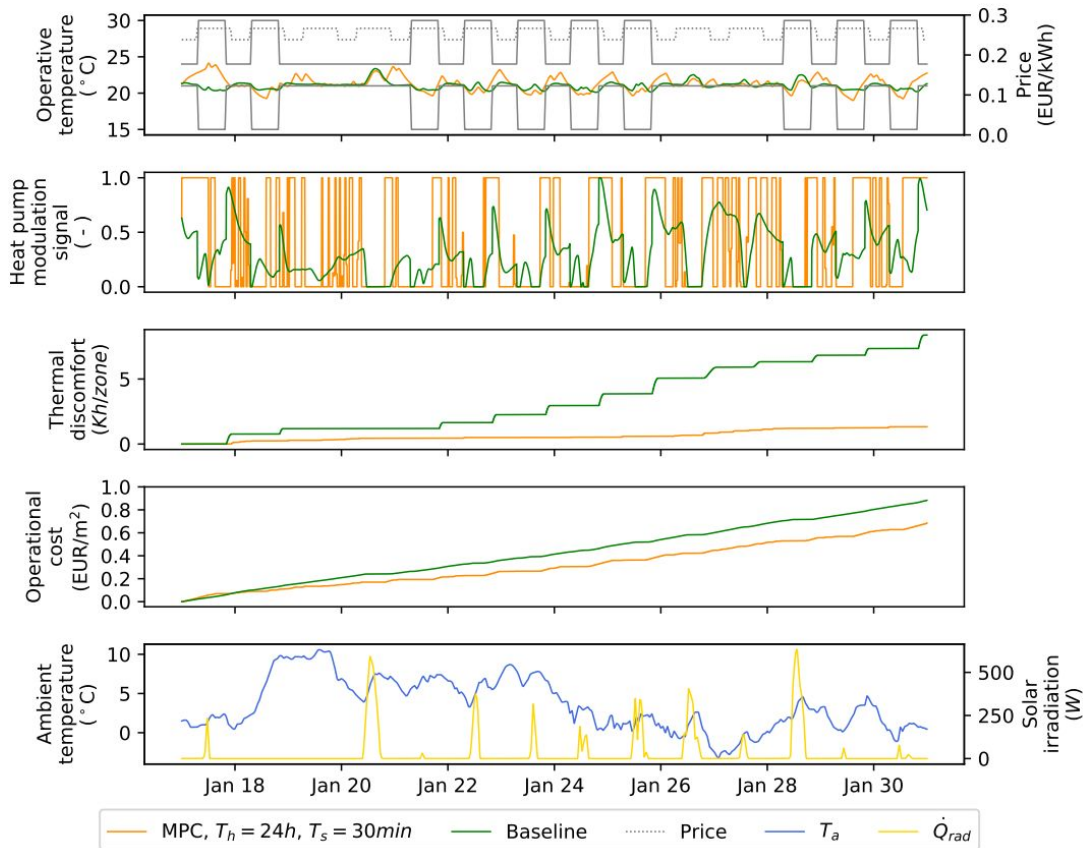
$$\begin{aligned} \min_{u_{HP}} \int_{t=t_i}^{t_h} (p^{e,\tau}(P_{hp} + P_{fan} + P_{pum}) + w\delta^{T_z}) dt \\ \dot{T}_z, P_{hp}, P_{fan}, P_{pum} = F(u_{hp}, \dot{Q}_{rad}, \dot{Q}_{occ}, T_a, T_z, T_c, T_f, T_i, T_w) \\ \underline{T}_z - \delta^{T_z} \leq T_z \leq \bar{T}_z + \delta^{T_z} \\ \delta^{T_z} \geq 0 \\ 0 \leq u_{hp} \leq 1. \end{aligned}$$

MPC results

```

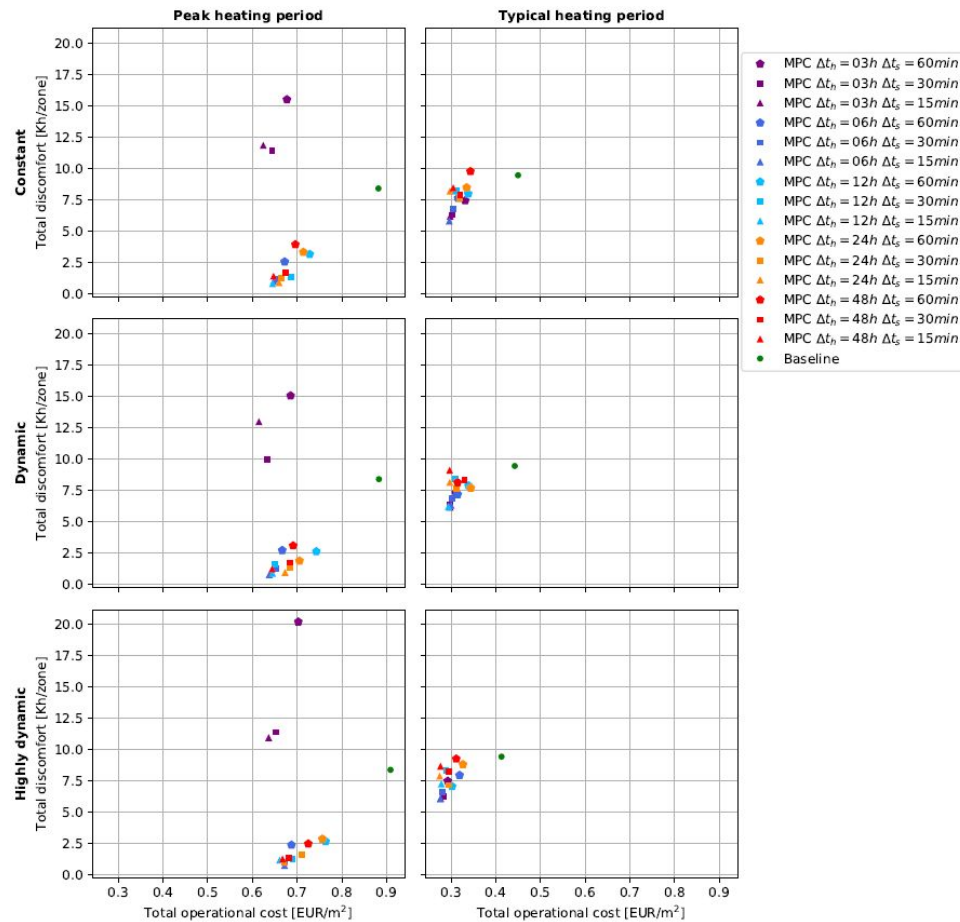
34 # -- Implement your MPC magic --
35
36
37
38 # Set step
39 requests.put("{0}/step/{1}".format(url, testid),
40             json={"step":30*60})
41
42 # Move to the peak heat testing period with dynamic pricing
43 y = requests.put("{0}/scenario/{1}".format(url, testid),
44                 json={"time_period":"peak_heat_day",
45                     "electricity_price":"dynamic"}).json()["payload"]
46
47 # Test your MPC magic
48 while y:
49     # Get forecast
50     f = requests.put("{0}/forecast".format(self.url),
51                     json={"point_names": ["TDryBul", "LowerSetp[1]"],
52                         "horizon":      int(24*3600),
53                         "interval":      int(3600)}).json()["payload"]
54
55     # Compute control signal
56     u = mpc.compute_control(y, f)
57
58     # Advance simulation with control signal
59     y = requests.post("{0}/advance/{1}".format(url, testid),
60                     json=u).json()["payload"]
61
62     # Get KPIs
63     kpi = requests.get("{0}/kpi/{1}".format(url, testid)).json()["payload"]

```



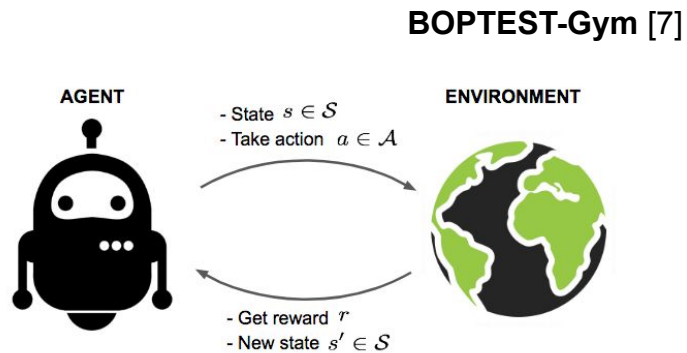
MPC results

```
61 # Get KPIs
62 kpi = requests.get("{0}/kpi/{1}".format(url, testid)).json()["payload"]
```

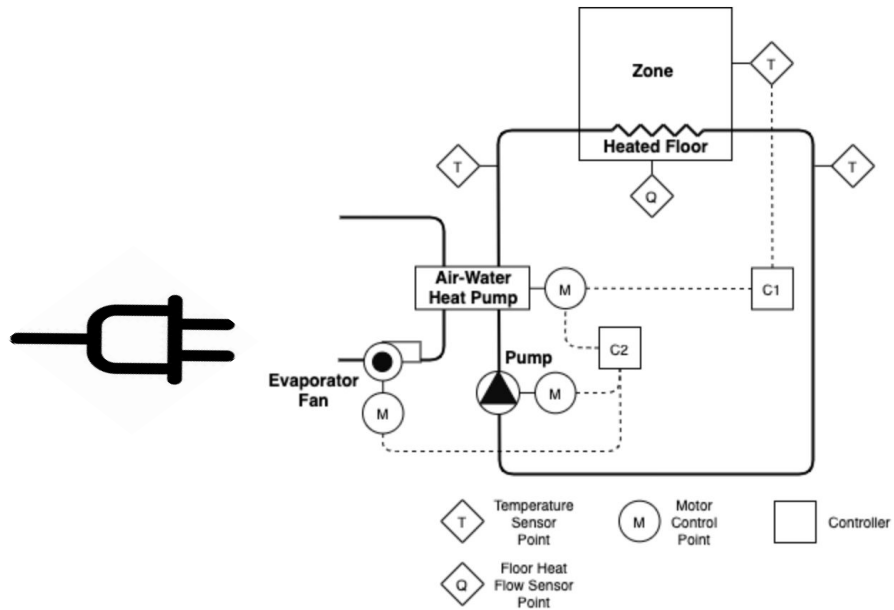


MPC vs. RL

RL be like:



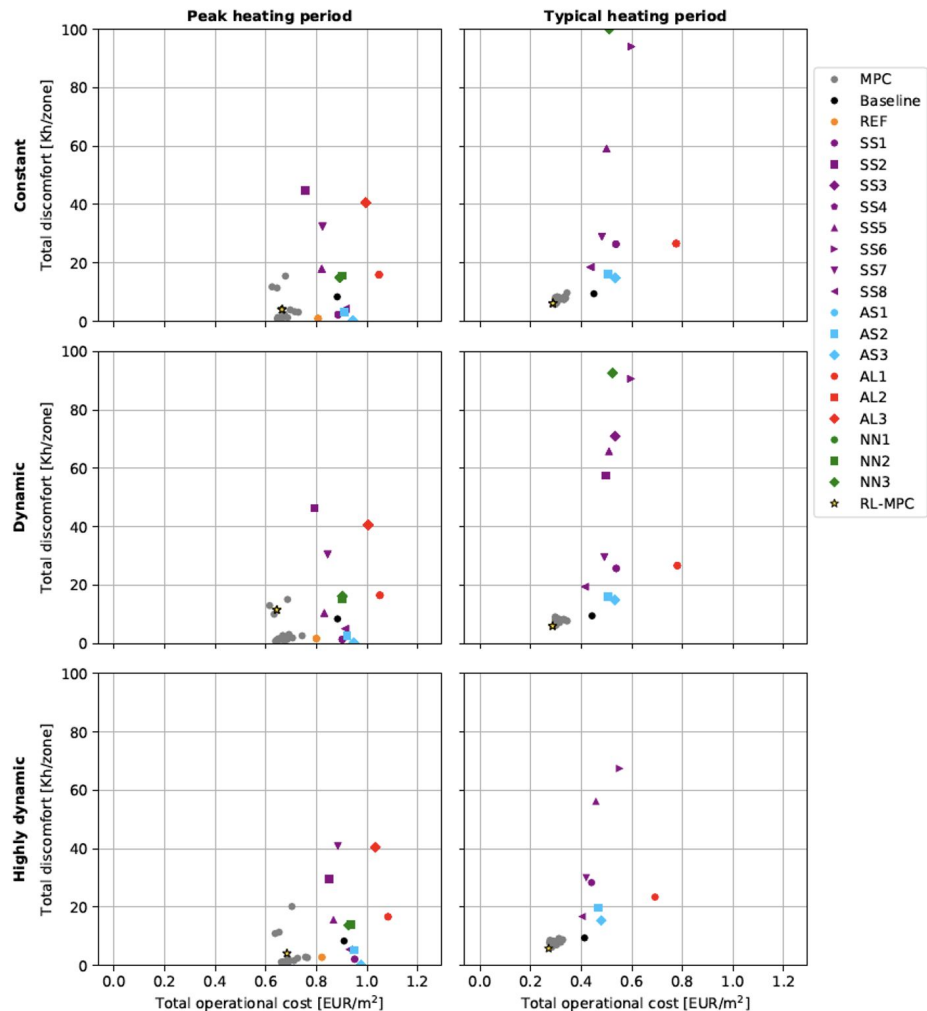
<https://lilianweng.github.io>



MPC vs. RL

```
61 # Get KPIs
62 kpi = requests.get("{0}/kpi/{1}".format(url, testid)).json()["payload"]
```

	\mathcal{S}_t	\mathcal{S}_m	\mathcal{S}_d														
Case	t_w	T_z	λ	\underline{T}	\overline{T}	T_a	\dot{Q}_{rad}	\dot{Q}_{occ}	Δt_s	Δt_h	Δt_r	$ \mathcal{S} $	\mathcal{A}	$ \mathcal{A} $	Alg	Net	Marker
REF	✓	✓	✓	✓	✓	✓	✓	✓	15m	24h	6h	608	u_{hp}	11	DDQN	2×64	●
SS1	✓	✓	✓	✗	✗	✗	✗	✗	15m	0h	0h	3	u_{hp}	11	DDQN	2×64	●
SS2	✓	✓	✓	✓	✓	✗	✗	✗	15m	0h	0h	3	u_{hp}	11	DDQN	2×64	■
SS3	✓	✓	✓	✓	✓	✗	✗	✗	15m	3h	0h	41	u_{hp}	11	DDQN	2×64	◆
SS4	✓	✓	✓	✓	✓	✓	✓	✓	15m	3h	6h	104	u_{hp}	11	DDQN	2×64	●
SS5	✓	✓	✓	✓	✓	✓	✓	✓	15m	6h	6h	176	u_{hp}	11	DDQN	2×64	▲
SS6	✓	✓	✓	✓	✓	✓	✓	✓	15m	12h	6h	320	u_{hp}	11	DDQN	2×64	▶
SS7	✓	✓	✓	✓	✓	✓	✓	✓	30m	24h	6h	308	u_{hp}	11	DDQN	2×64	▼
SS8	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	158	u_{hp}	11	DDQN	2×64	◀
AS1	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	2	DDQN	2×64	●
AS2	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	T_{set}	11	DDQN	2×64	■
AS3	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	T_{set}^1	11	DDQN	2×64	◆
AL1	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	∞	SAC	2×64	●
AL2	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	∞	A2C	1×64	■
AL3	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	∞	PPO	2×64	◆
NN1	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	11	DDQN	1×64	●
NN2	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	11	DDQN	1×32	■
NN3	✓	✓	✓	✓	✓	✓	✓	✓	60m	24h	6h	608	u_{hp}	11	DDQN	2×64 ²	◆



References

BOPTTEST Github Repository: <https://github.com/ibpsa/project1-boptest>

BOPTTEST OpenAI-Gym Interface: <https://github.com/ibpsa/project1-boptest-gym>

- [1] D. Blum, F. Jorissen, S. Huang, Y. Chen, J. Arroyo, K. Benne, Y. Li, V. Gavan, L. Rivalin, L. Helsen, D. Vrabie, M. Wetter, and M. Sofos. (2019). “Prototyping the BOPTTEST framework for simulation-based testing of advanced control strategies in buildings.” In *Proc. of the 16th International Conference of IBPSA*, Sep 2 – 4. Rome, Italy.
- [2] D. Blum, J. Arroyo, S. Huang, J. Drgona, F. Jorissen, H. T. Walnum, T. Chen, K. Benne, D. Vrabie, M. Wetter, and L. Helsen (2021). “Building Optimization Testing Framework (BOPTTEST) for Simulation-Based Benchmarking of Control Strategies in Buildings.” *Journal of Building Performance Simulation*, Accepted.
- [3] X. Pang, M. A. Piette, and N. Zhou (2017). “Characterizing variations in variable air volume system controls.” *Energy and Buildings*, vol. 135, pp. 166–175.
- [4] R. D. Coninck, F. Magnusson, J. Akesson, and L. Helsen (2016). “Toolbox for development and validation of grey-box building models for forecasting and control” *Journal of Building Performance Simulation*, vol. 9, no. 3, pp. 288–303.
- [5] S. Lucia, A. Tatulea-Codrean, C. Schoppmeyer, and S. Engell. Rapid development of modular and sustainable nonlinear model predictive control solutions. *Control Engineering Practice*, 60:51-62, 2017
- [6] Picard, D., Jorissen, F., & Helsen, L. (2015). Methodology for obtaining linear state space building energy simulation models. In *11th international modelica conference* (pp. 51-58).
- [7] Arroyo, J., Manna, C., Spiessens, F., and Helsen, L (2021). “An OpenAI-Gym environment for the Building Optimization Testing (BOPTTEST) framework.” In *Proc. of the 17th International Conference of IBPSA*, Sep. 1 – 3 Bruges, Belgium.

Example Development and Evaluation of Optimal Control

Thank you!

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vito



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