

Moving horizon estimation based model predictive control (MHE-MPC) for offset-free control of a single canal pool

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Introduction

- Model predictive control (MPC) is a powerful control option which is increasingly used by operational water managers for managing water systems.
- Closed loop performance of MPC is directly related to model accuracy. The **mismatch** between **real system** and **model** from **unknown disturbances** will result in **steady state offset** in the controlled output.
- Unknown Disturbances
 - Unknown water offtakes
 - Water loss (seepage, leakage, evaporation)
- Moving horizon estimation (MHE) is an optimization-based state estimation method working on an estimation window covering a certain number of past measurements.
- Moving horizon estimation based model predictive control (MHE-MPC) uses the past predictions of the model and the past measurements of the system to estimate unknown disturbances.

Offset Due to Unknown Disturbances

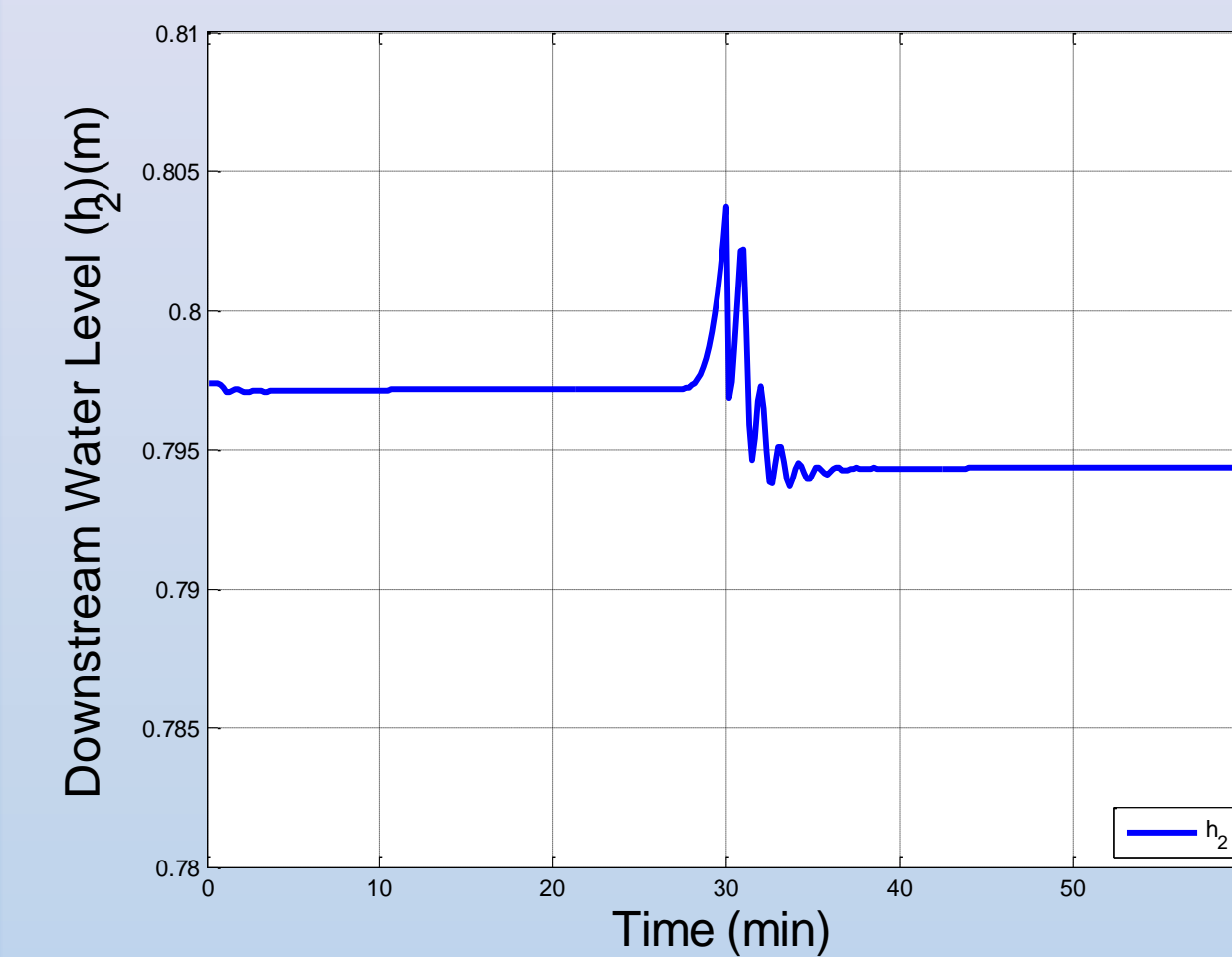


Fig. 2 Downstream water level (h_2) using MPC without offset free method

Set-point is 0.8 m, however due to unknown disturbances in the system, MPC can not achieve offset-free control. The water level reached a steady state but with an undesired offset.

Moving Horizon Estimation and MHE-MPC

The internal model and the available output measurements over a given estimation window are used by MHE, for estimating the unknown disturbances (Q_{ext}) at the current time. The estimation is obtained by solving a least-squares problem over the estimation window, N , with an objective function that minimize the differences between the predicted water levels, $h_p(k)$ and the measured water levels $h_m(k)$:

$$\min_{Q_{ext}} \sum_{i=1}^N (h_m(i) - h_p(i))^2$$

Where

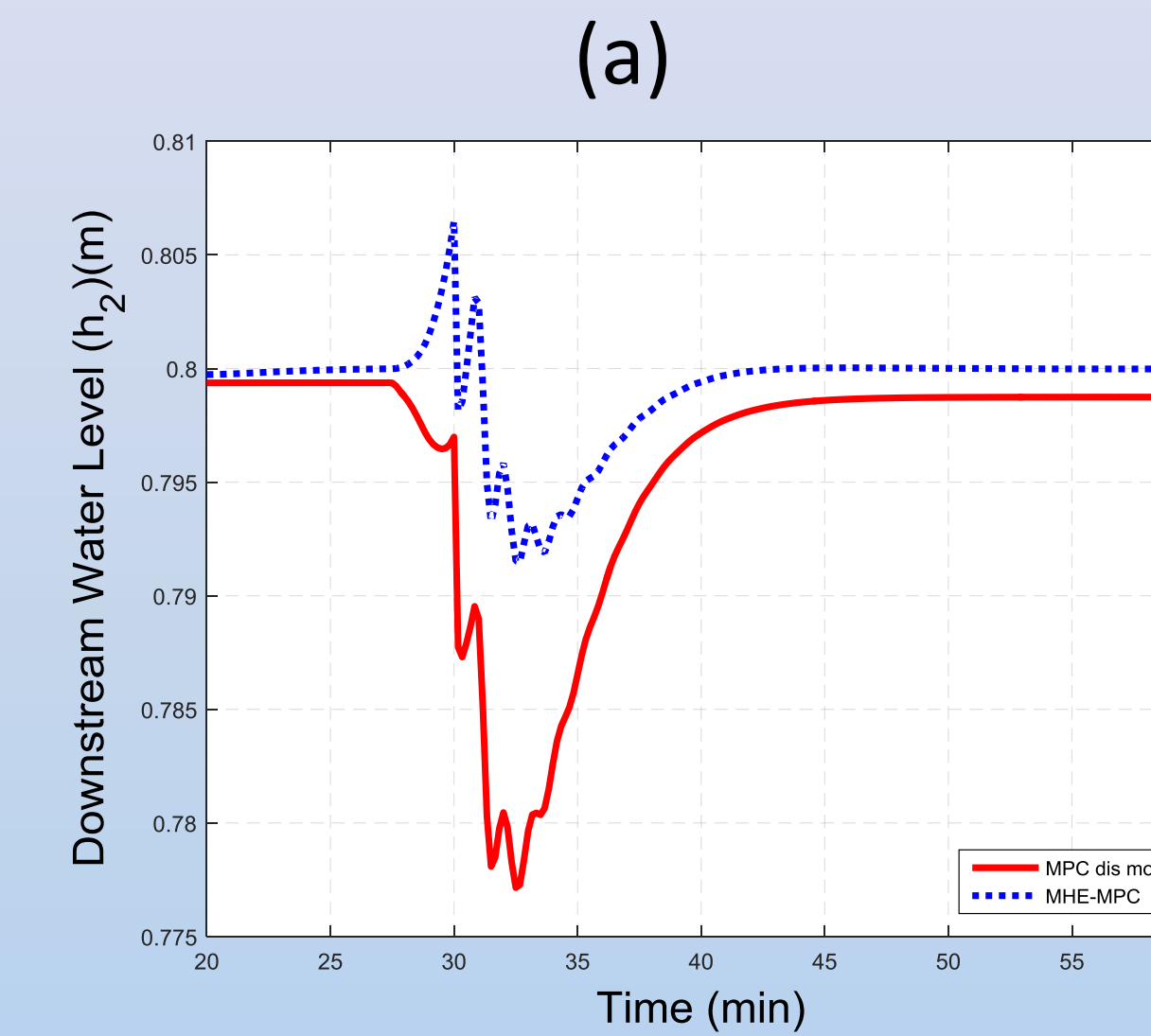
$$h_p(s) = \frac{\omega_0^2}{A_s s^3 + \frac{s^2}{M} + A_s \cdot \omega_0^2 \cdot s} \cdot Q_1(s) - \frac{2 \cdot s^2 + \frac{2}{A_s \cdot M} \cdot s + \omega_0^2}{A_s s^3 + \frac{s^2}{M} + A_s \cdot \omega_0^2 \cdot s} \cdot (Q_2(s) + Q_{ext}(s))$$

Estimation of Q_{ext} is added to the known disturbances in the MPC optimization

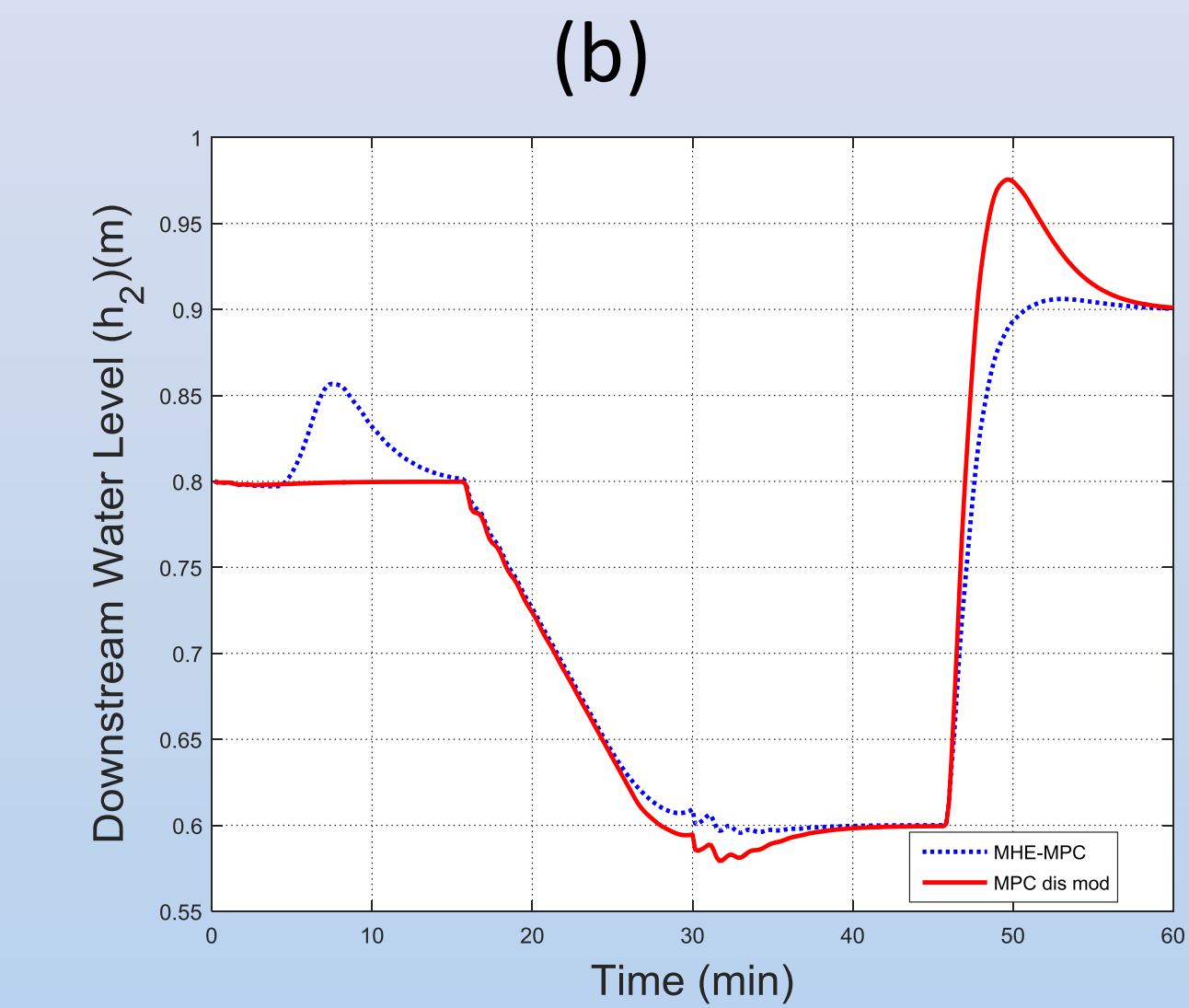
For comparison, an alternative offset-free scheme (MPC with disturbance modelling) is used. Disturbances were modelled by a disturbance model which augments the system states with integrating disturbances [3].

Simulation Results

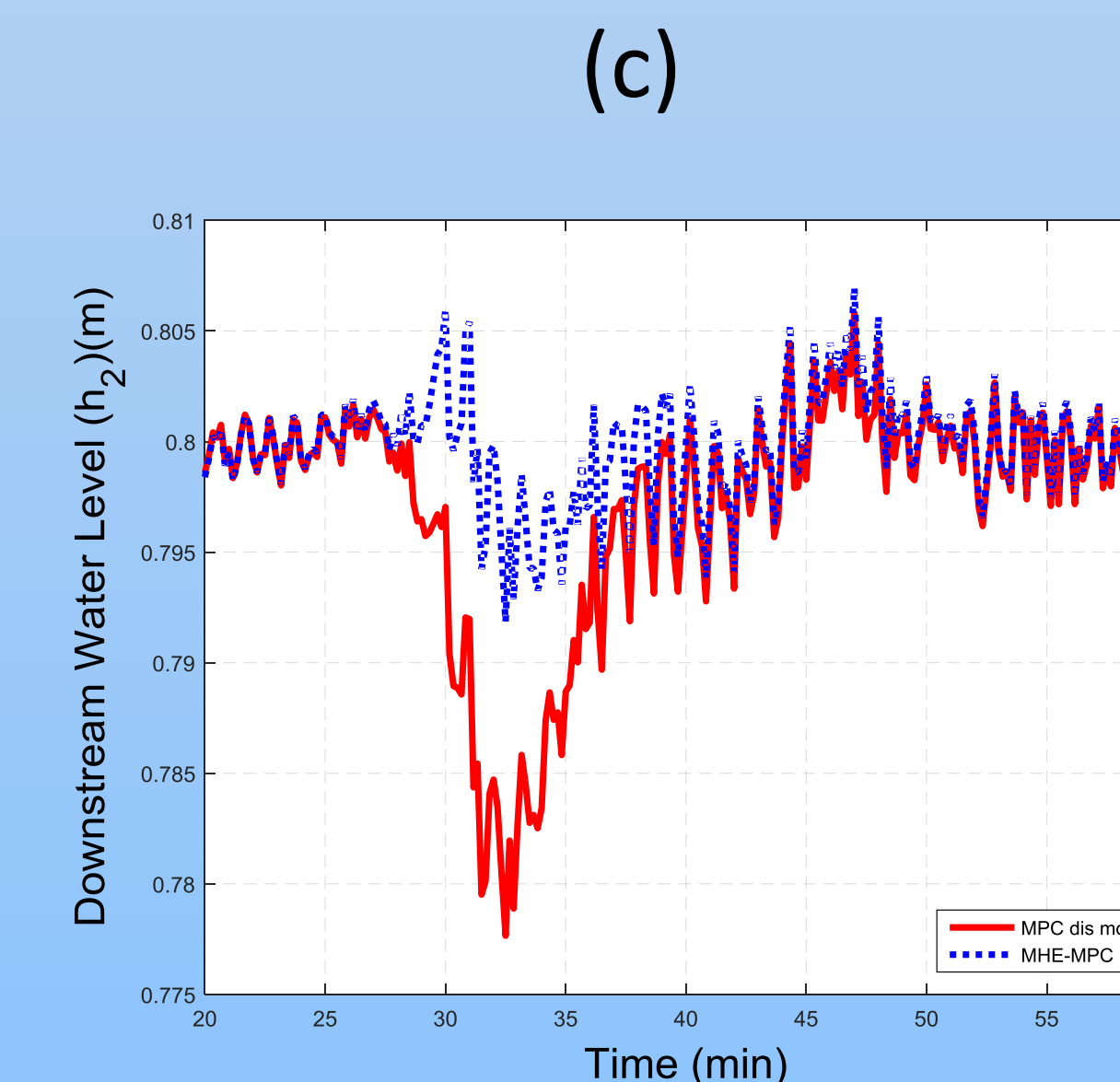
- 30 % Unknown Disturbance in Q_2



-Set-point change & 10 % unknown disturbance in Q_2



- Uncertain Unknown Disturbance in Q_2



-Set-point change & 200 % unknown disturbance in Q_2

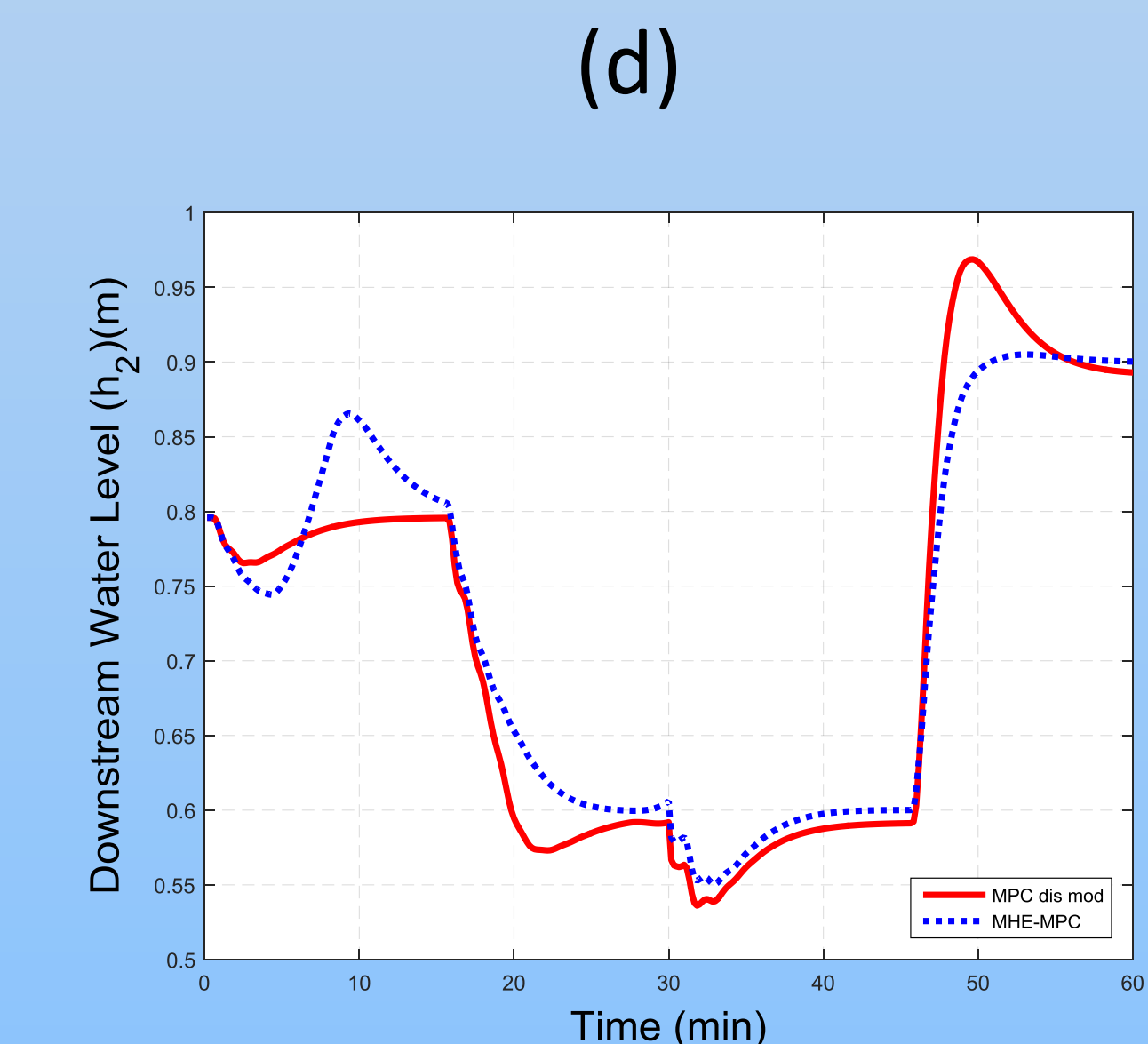


Fig. 3 Downstream water level (h_2) for 30% unknown disturbance in Q_2 (a); set point change & 10 % unknown disturbance in Q_2 (b); uncertain unknown disturbance in Q_2 (c); set point change & 200 % unknown disturbance in Q_2 (d)

Conclusion

- MHE-MPC estimates the unknown disturbance(s) with MHE and it is used to update the known disturbance for the MPC and the offset is removed.
- Simulation results show that the MHE-MPC is achieving better offset-free performance than the MPC with disturbance modeling scheme.
- MHE-MPC needs a certain amount of simulation time steps, length of the estimation window, to start unknown disturbance estimation. Therefore, the first step ($N+1$), MHE-MPC starts estimating the unknown disturbance, steep water level increases observed due to overshoots of the controlled inflow (Fig.3-b, Fig.3-d). Once the overshoots of inflow are resolved, MHE-MPC responds better to the known disturbances than the MPC with disturbance modeling.
- MHE-MPC achieves offset-free control in case of uncertain unknown disturbances (Fig.3-c).

Test Canal – Internal Model

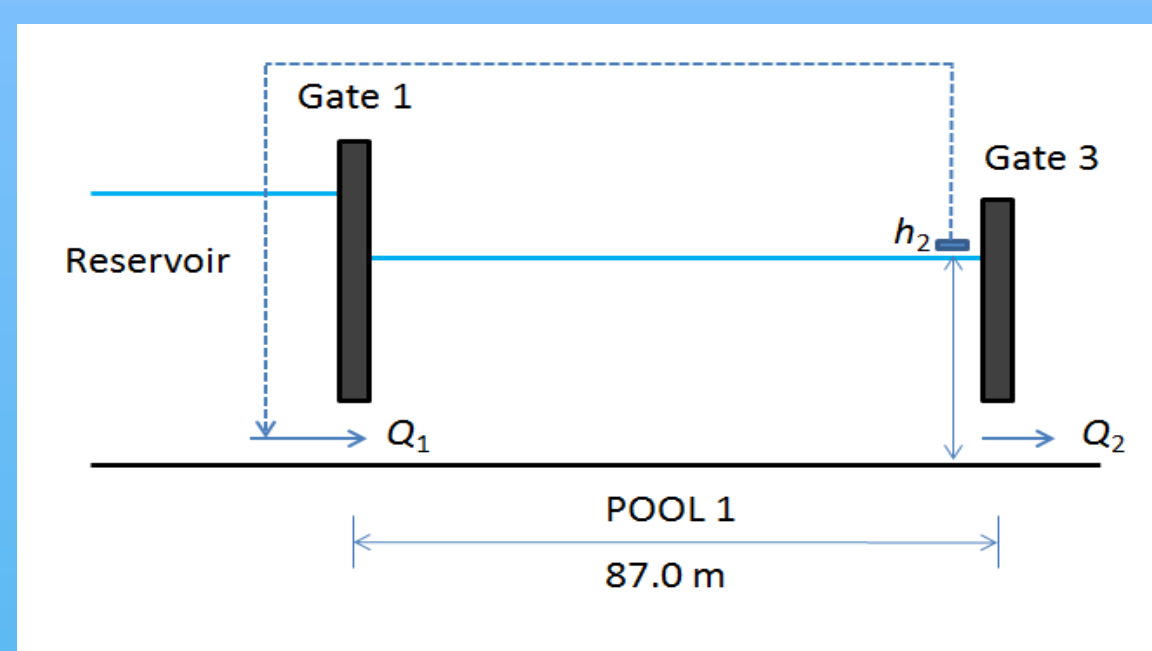


Table.1 Test canal parameters

Q_1 (m ³ /s)	0.010
A_s (m ²)	38.28
ω_0 (rad/s)	0.101
M (-)	35.09

Fig.1 Schematic of a single canal pool with distant downstream controller

Integrator Resonance Model [2]

$$h_2(s) = \frac{\omega_0^2}{A_s s^3 + \frac{s^2}{M} + A_s \cdot \omega_0^2 \cdot s} \cdot Q_1(s) - \frac{2 \cdot s^2 + \frac{2}{A_s \cdot M} \cdot s + \omega_0^2}{A_s s^3 + \frac{s^2}{M} + A_s \cdot \omega_0^2 \cdot s} \cdot Q_2(s)$$

References

- [1] Aydin, B.E., van Overloop, P.J., Rutten M., Tian X. Offset-Free Model Predictive Control of Open Water Channel Based on Moving Horizon Estimation. submitted to *Journal of Irrigation and Drainage Engineering*
- [2] van Overloop, P.J., Horváth, K., and Aydin, B. E. (2014). 'Model predictive control based on an integrator resonance model applied to an open water channel.' *Control Engineering Practice*, Elsevier, 27, 54–60.
- [3] Aydin, B.E., van Overloop, P.J., Tian, X. (2014). 'Offset-free Model Predictive Control of an Open Water Reach.' *International Conference on Hydroinformatics*, New York.

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