

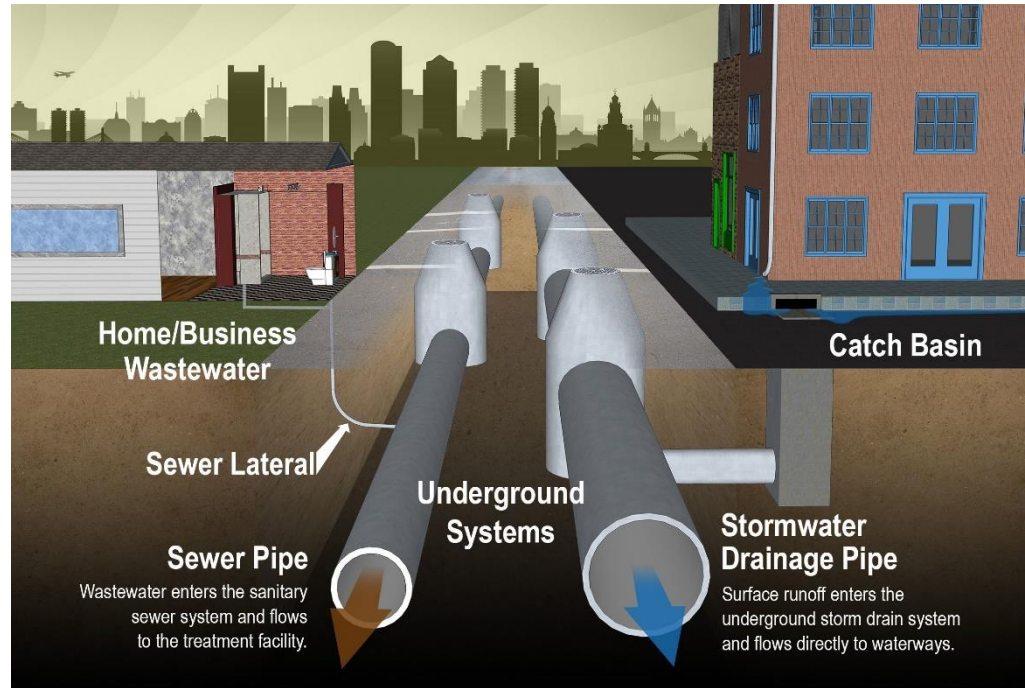
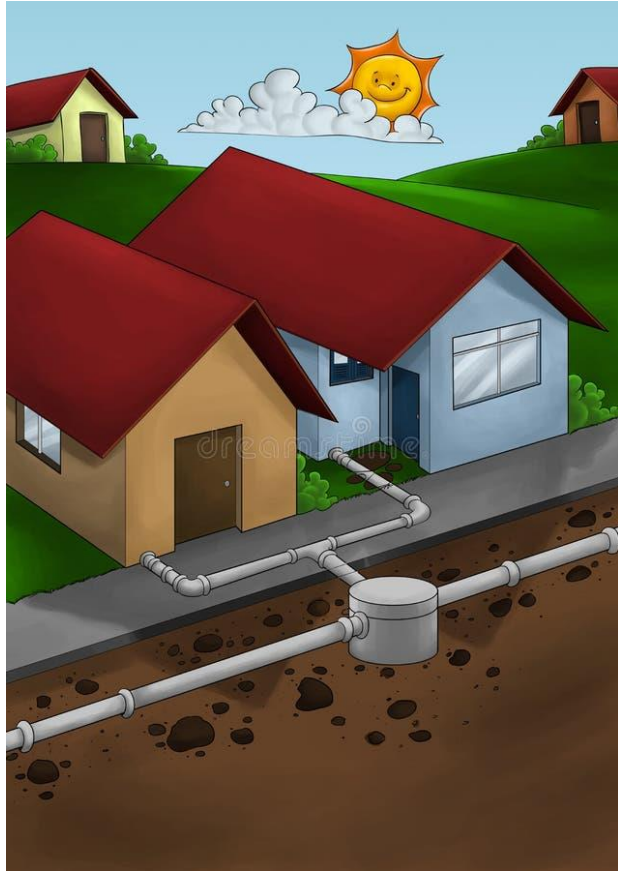
# Graph neural networks for urban drainage systems metamodeling

Alexander Garzón et al.

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# Urban drainage systems



[AASHTO Specs for Storm Water Drainage Pipes, Manholes ETC. | LinkedIn](#)



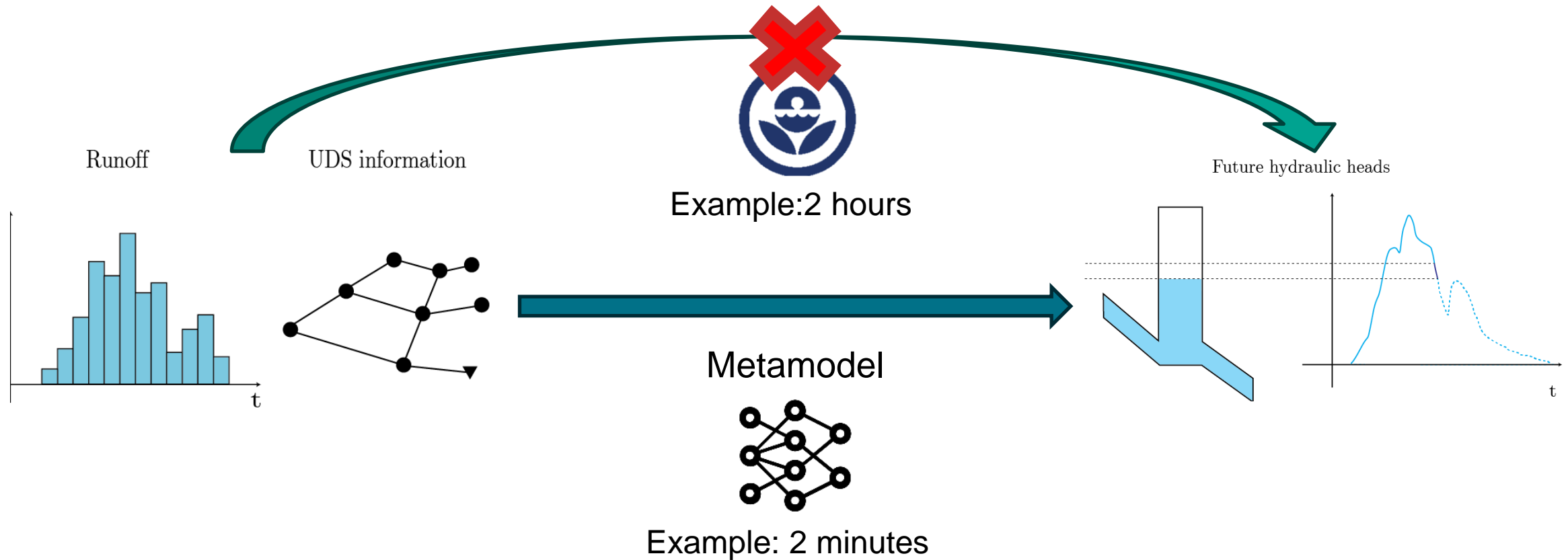
[Infiltrerende stad - HVA](#)



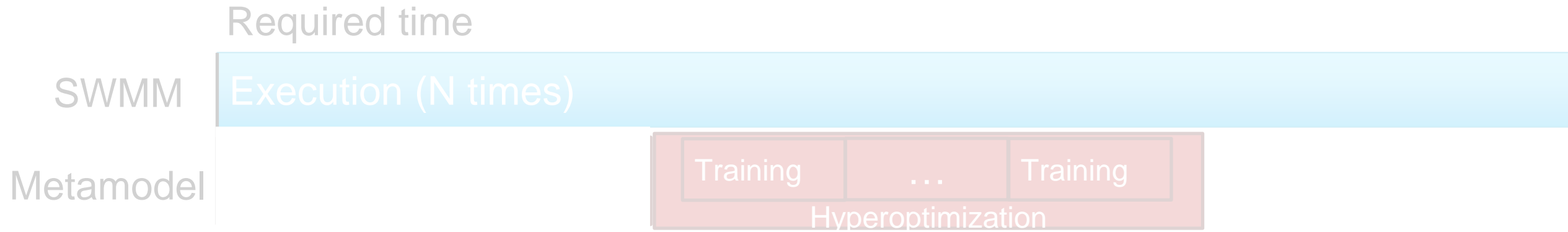
[Rioolwater in watergang Hullerweg Nunspeet - Al het nieuws uit Nunspeet \(nunspeethuisaanhuis.nl\)](#)

# Metamodels

Computer intensive applications require fast simulators

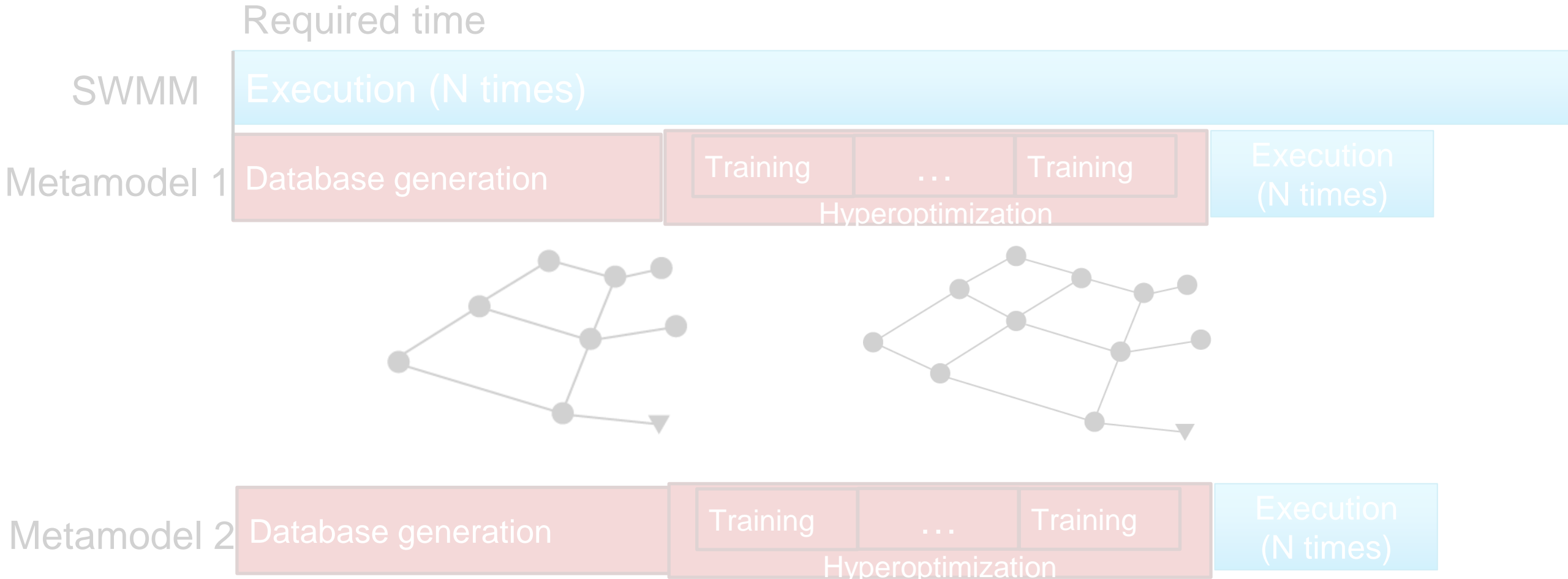


# Problem 1: Developing a metamodel can be time-consuming



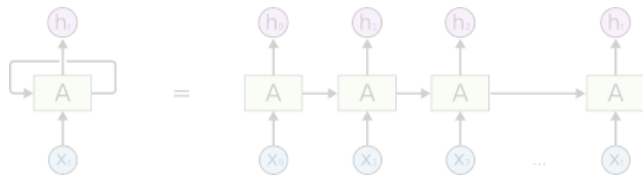
# Problem 2:

Metamodels are system specific

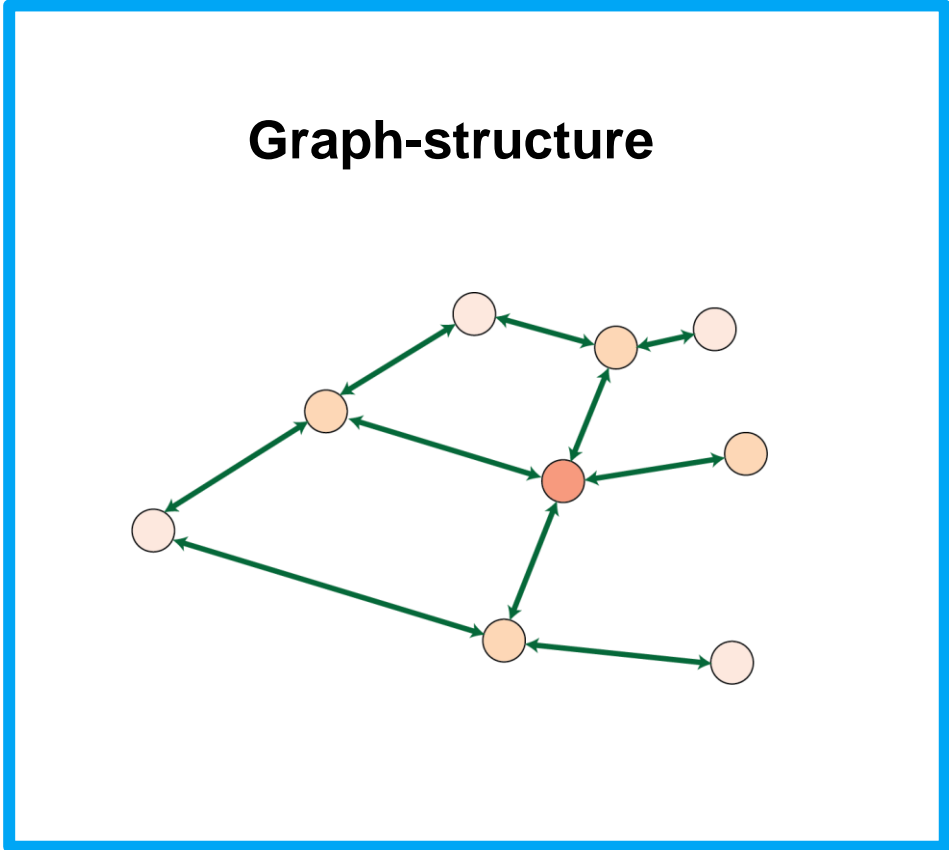


# Proposal: Inductive biases in Machine Learning algorithms

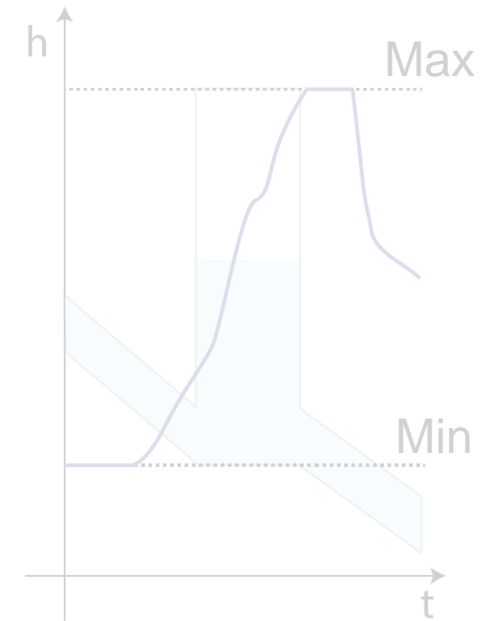
Temporal



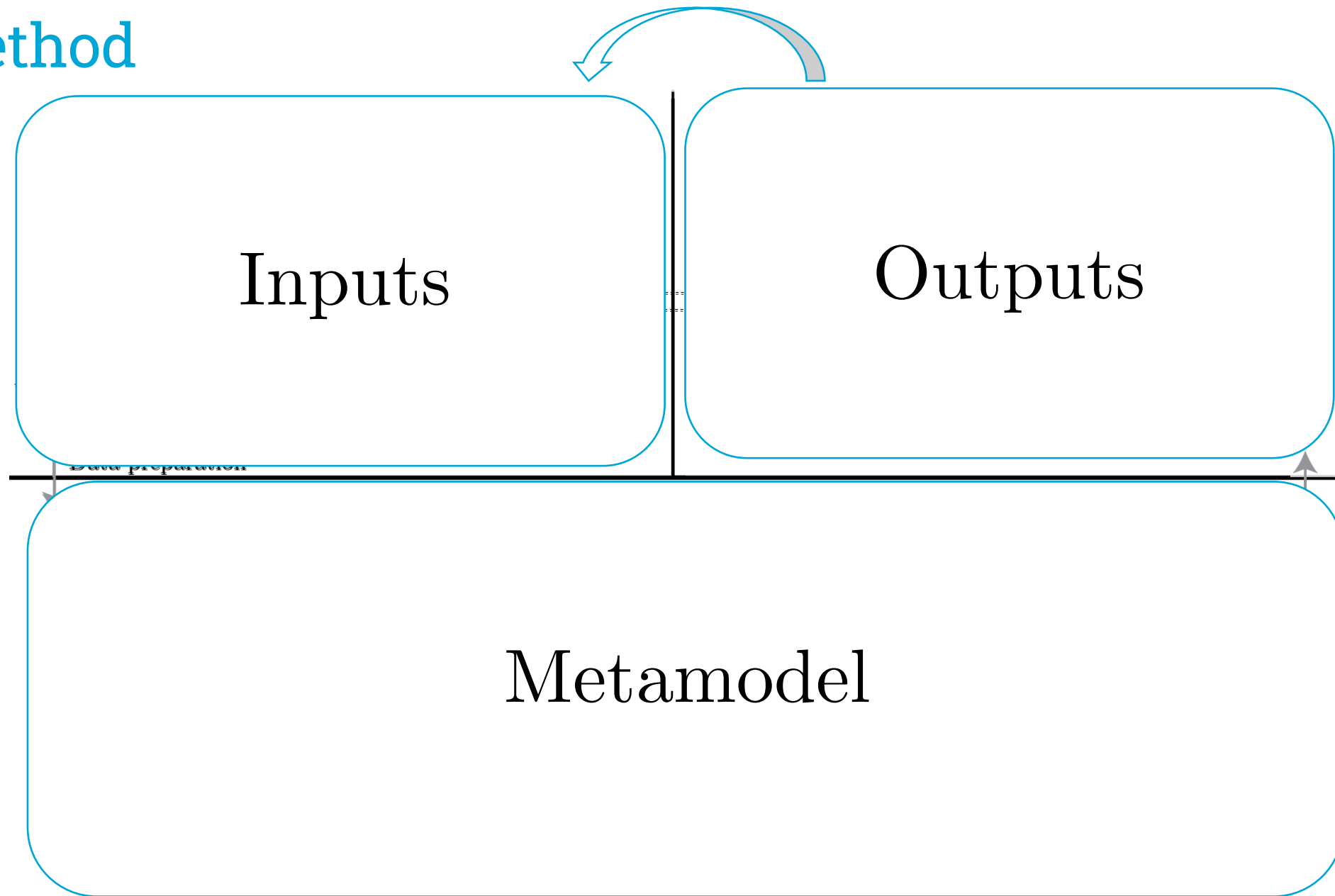
Graph-structure



Physical



# Method





# Case study

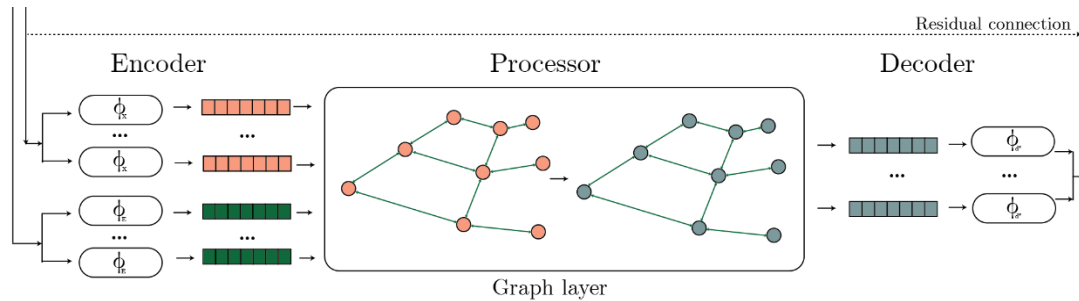
- Tuindorp.
  - Utrecht, The Netherlands
  
- Combined drainage system
  - Storm water drainage for this study





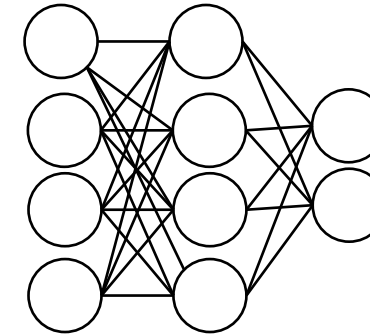
# Metamodels

- GNN metamodel



~6.000 parameters

- MLP metamodel

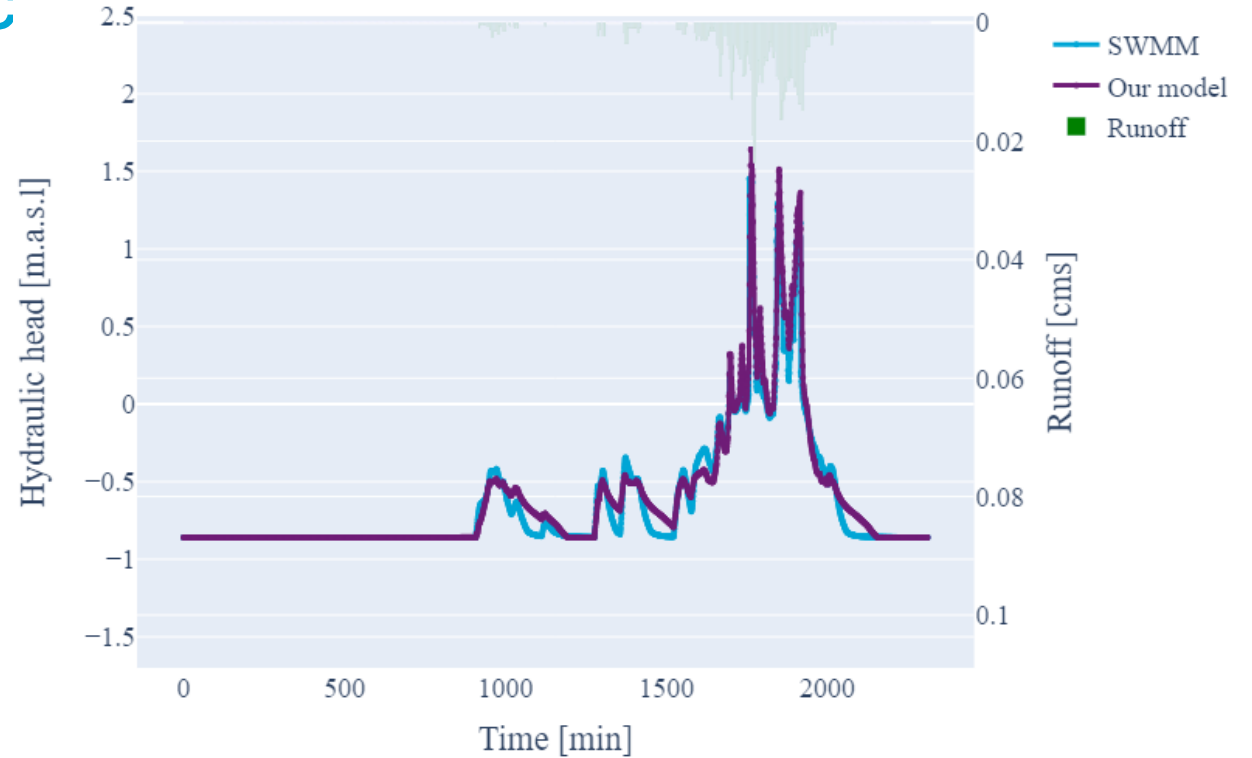
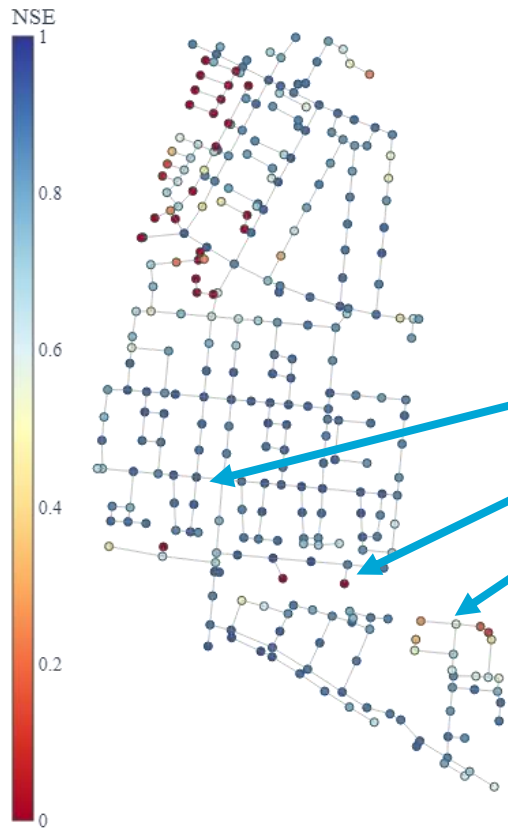


~700.000 parameters

Same dynamic inputs

Same temporal and physical bias

# Results – Test Performanc



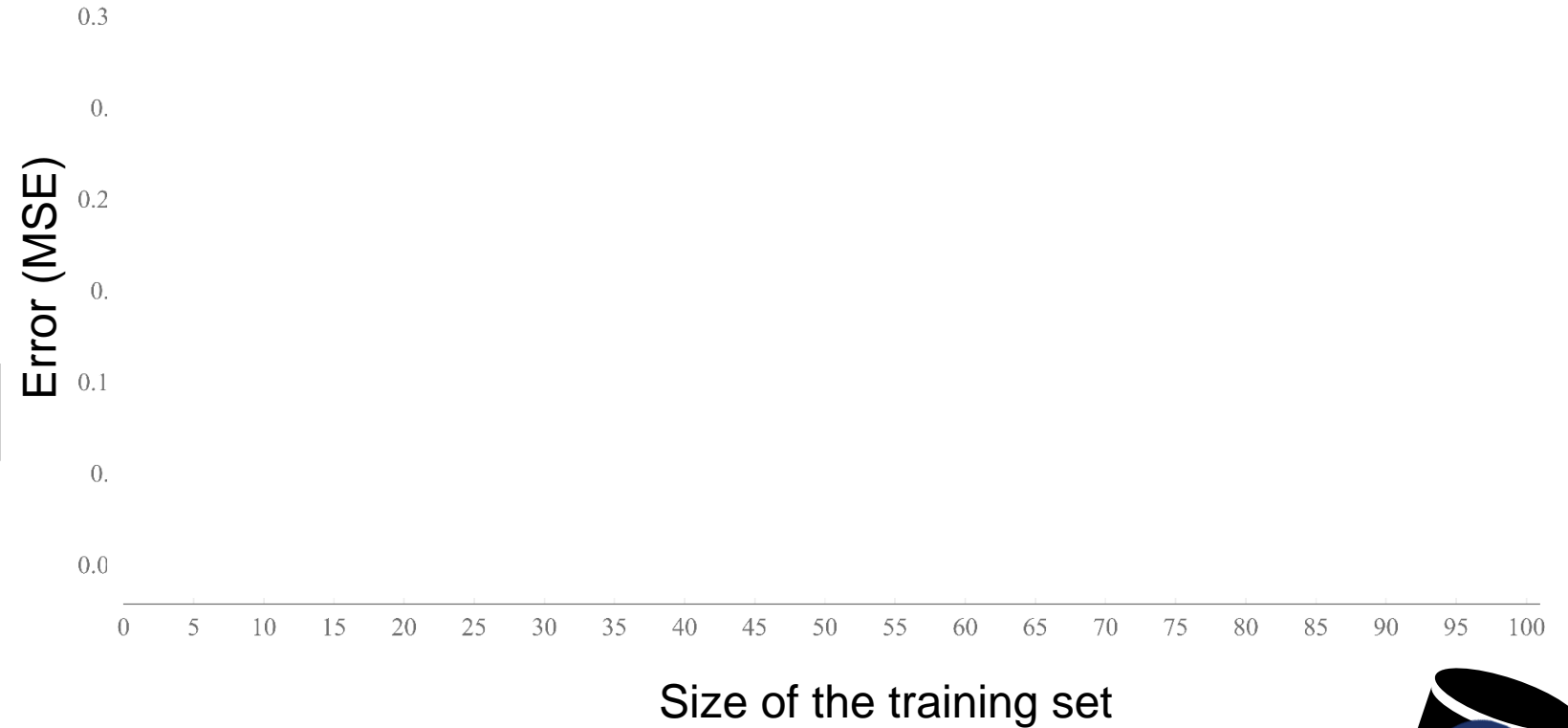
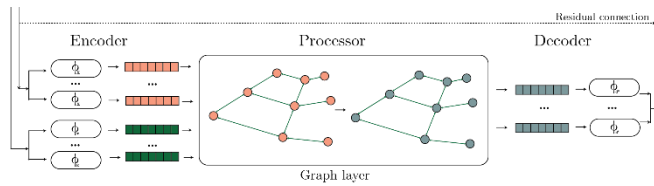
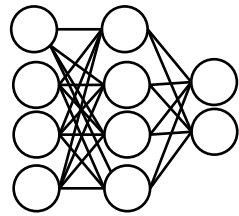
Accuracy

MLP			GNN		
No Flow	Flow	Overall	No Flow	Flow	Overall
1.0	0.908	0.976	0.997	0.924	0.981

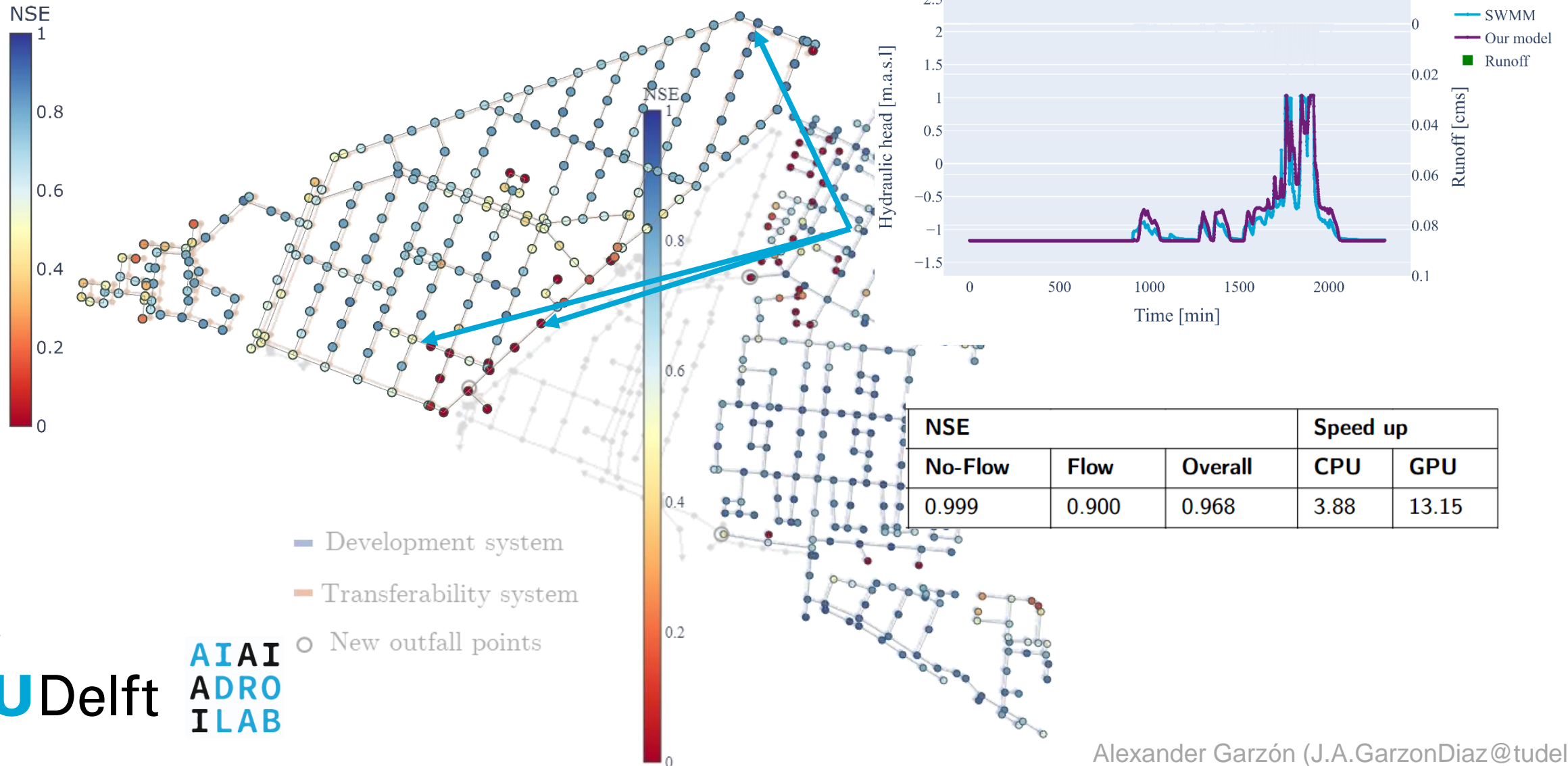
Speed

MLP (CPU)	MLP (GPU)	GNN (CPU)	GNN (GPU)
Speed-up	Speed-up	Speed-up	Speed-up
11.96	35.04	3.69	17.39

# Results – Data Efficiency



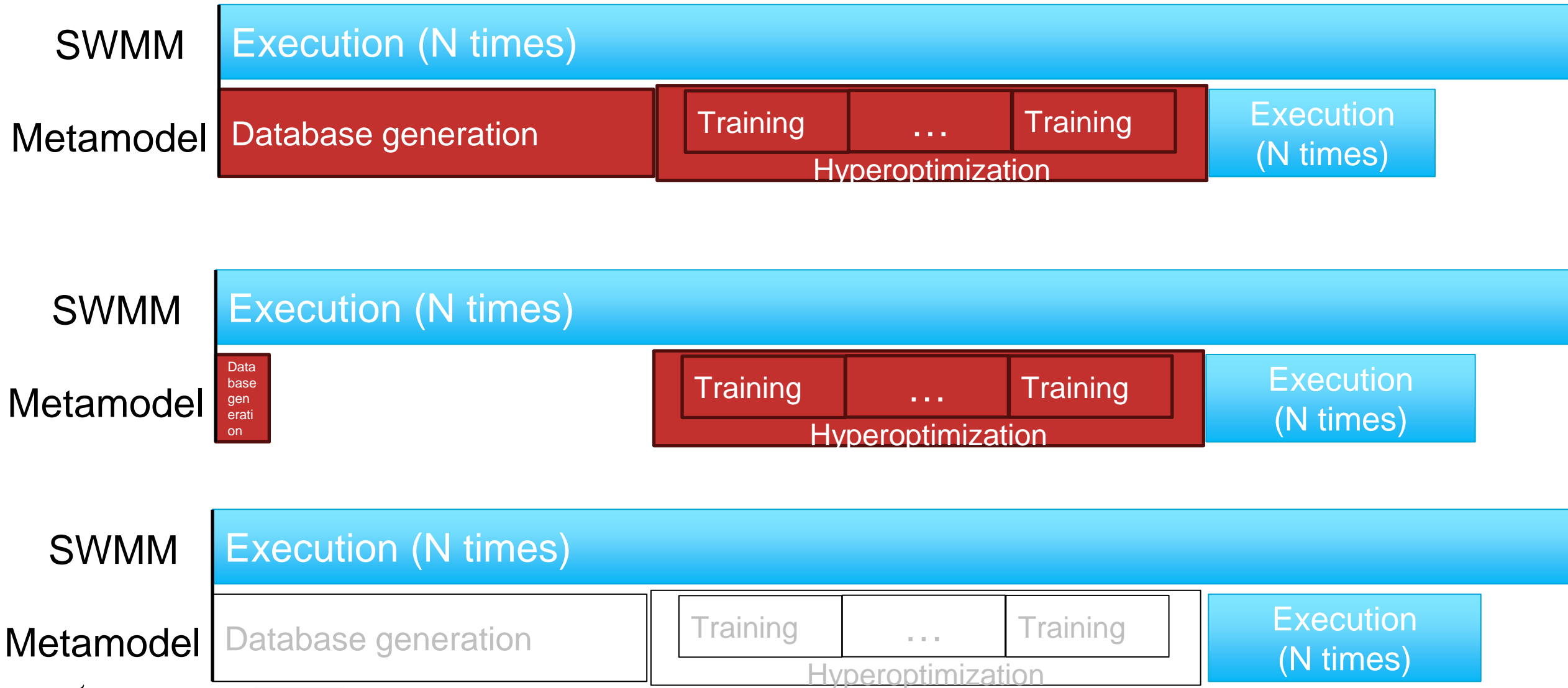
# Results - Transferability



# Limitations – Future work

- Special components (Pumps, tanks, orifices, weirs, etc.)
- Dry weather Flow
- Flow estimation
- Other catchments
- Hyperparameter pre-selection

# Required time





# Thank you for your attention

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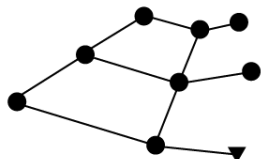
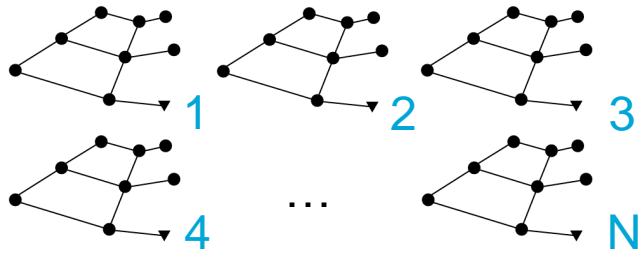
# References

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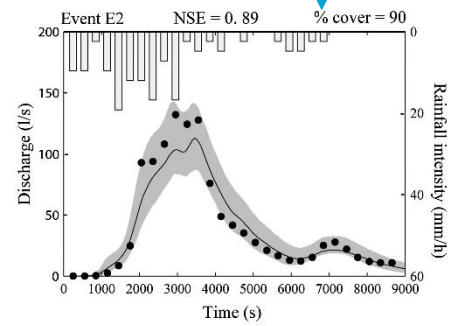
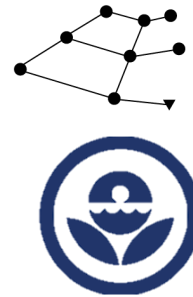
# Motivation

Some critical applications require a fast simulator

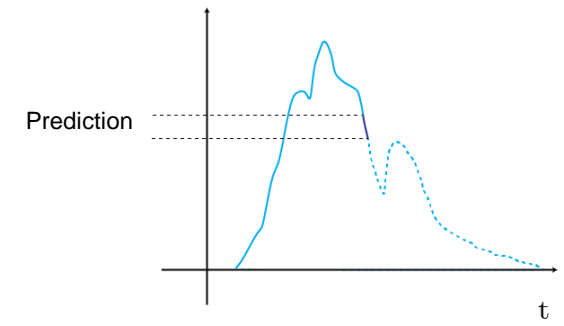
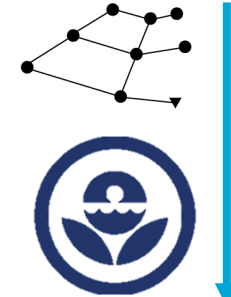
- Optimisation



- Uncertainty analysis



- Real-time forecast



# Problem 1 Developing a metamodel can be time-consuming

Required time

SWMM

Execution (N times)

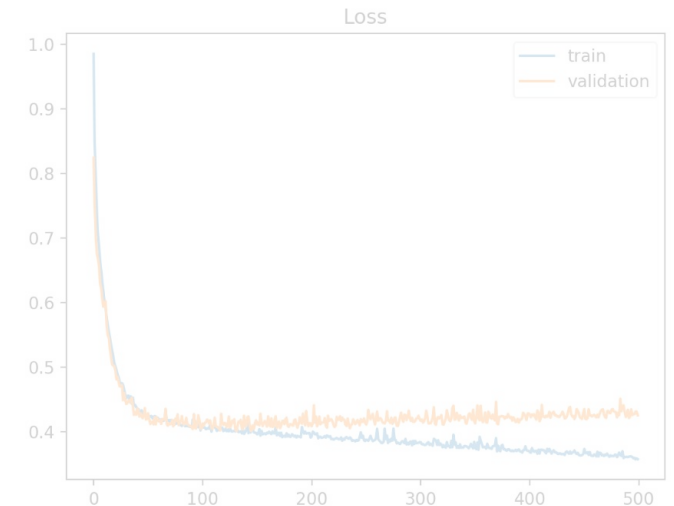
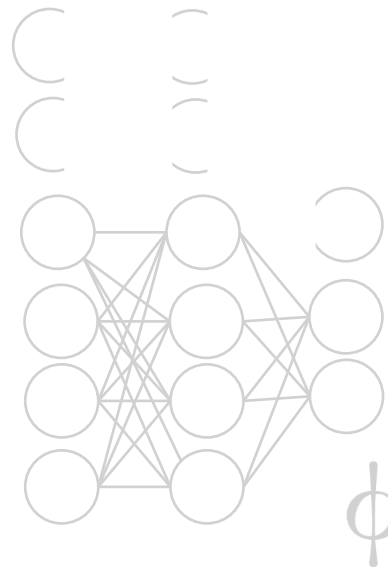
Metamodel

Training

...

Training

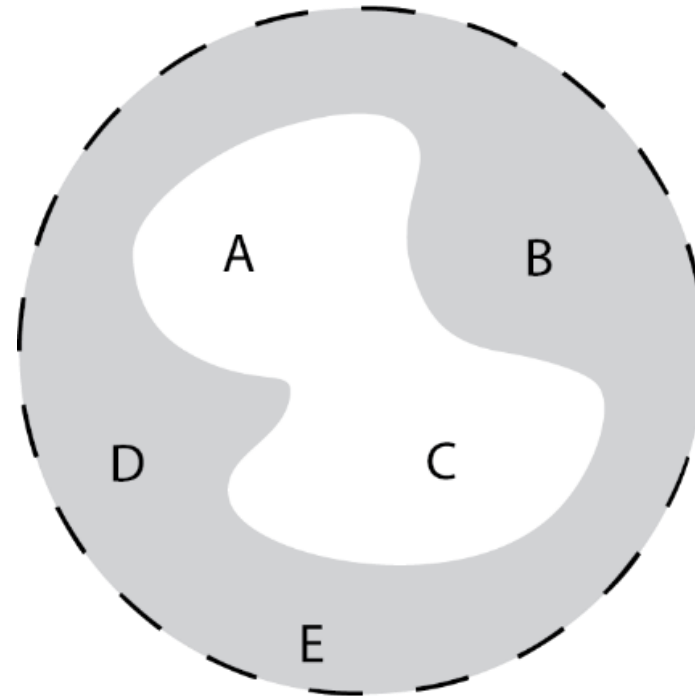
Hyperoptimization



# Proposal: Inductive biases in Machine Learning models

Expression of **assumptions** about either the data-generating process or the space of solutions.

It **prioritizes** some solutions over others.



- Practical implications:**
- Requires less training examples
  - Generalization of learned features