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Per-Link Modeling Towards Self-Adaptive Network Digital Twins

MSWiM 2025

Universitat Politècnica de Catalunya (UPC)

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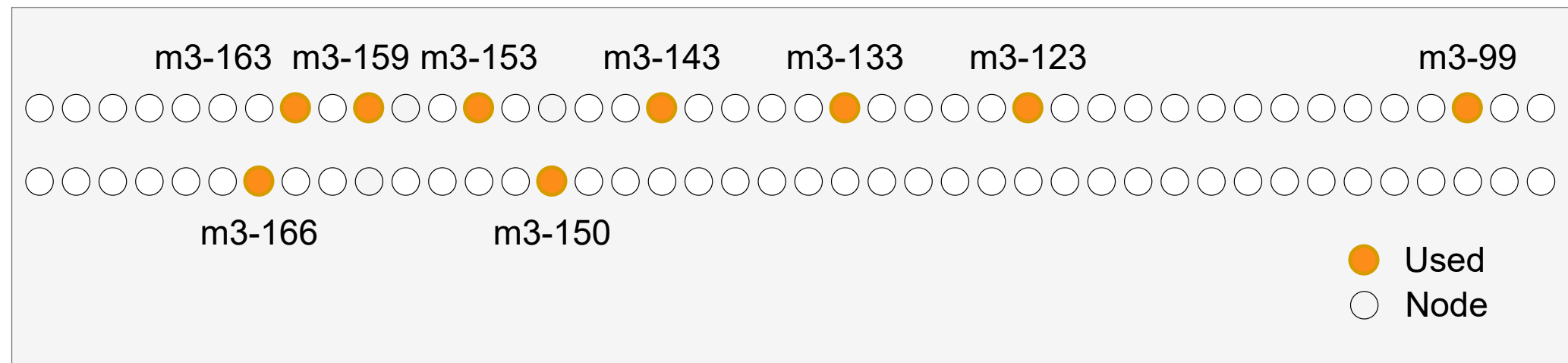
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28 October 2025

Problem Statement

❖ Case Study:

- Network architect deploying IoT solutions
- Industrial deployment on the experimental platform *FIT IoT-Lab* [1]
 - **Smart metering in an indoor environment**



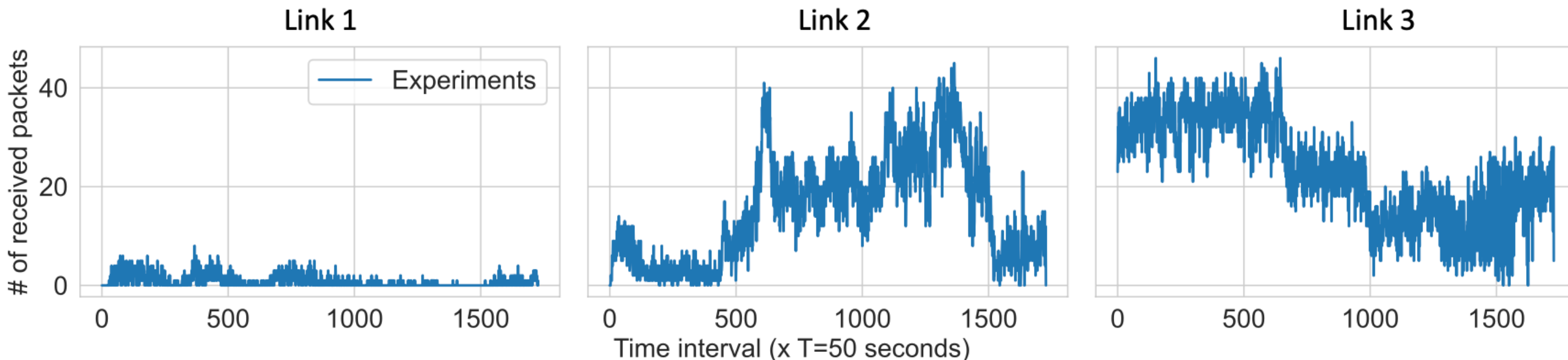
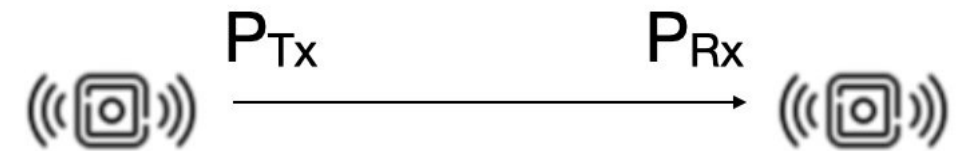
- 9 nodes on the Grenoble site, 24 hours deployment
- Communication technology: 802.15.4, on *Contiki-NG*
- Traffic: Broadcast, periodic (1 packet/s)
- **Objective:** Reproduce and predict the evolution of the PDR (Packet Delivery Ratio)

[1] Adjih, Cedric, et al. "FIT IoT-LAB: A large scale open experimental IoT testbed." 2015 IEEE 2nd World Forum on Internet of Things (WF-IoT). IEEE, 2015.

Problem Statement

❖ Compare between the PDR evolution:

- From real measures
- From the Cooja simulator [2]:
 - Calibrated with an average PDR value per link (fixed)

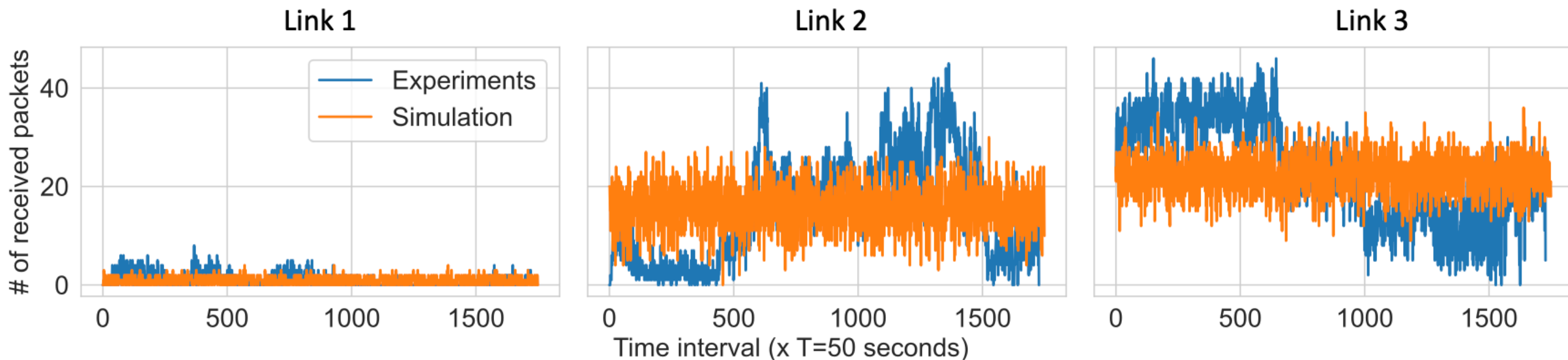
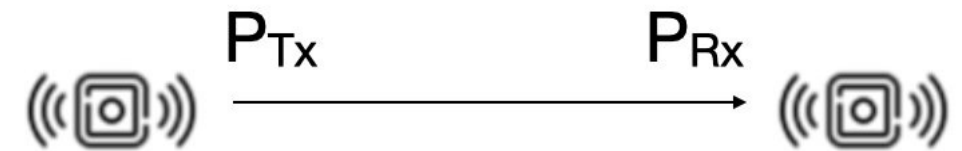


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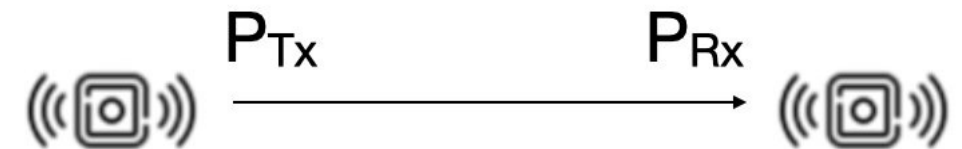


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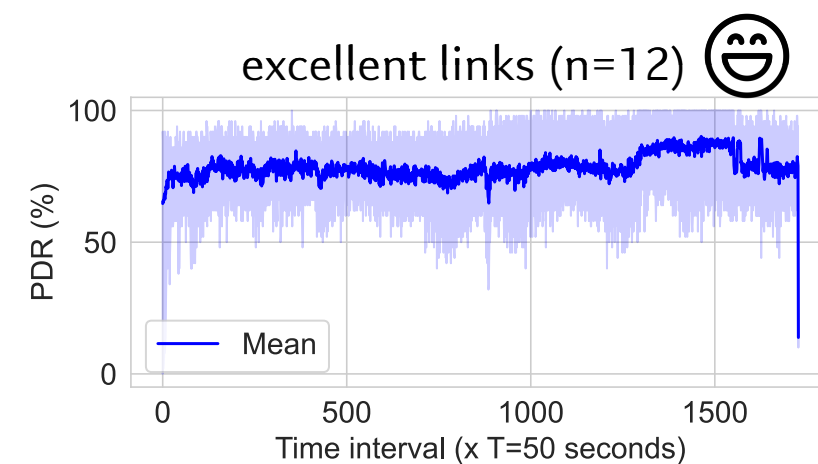
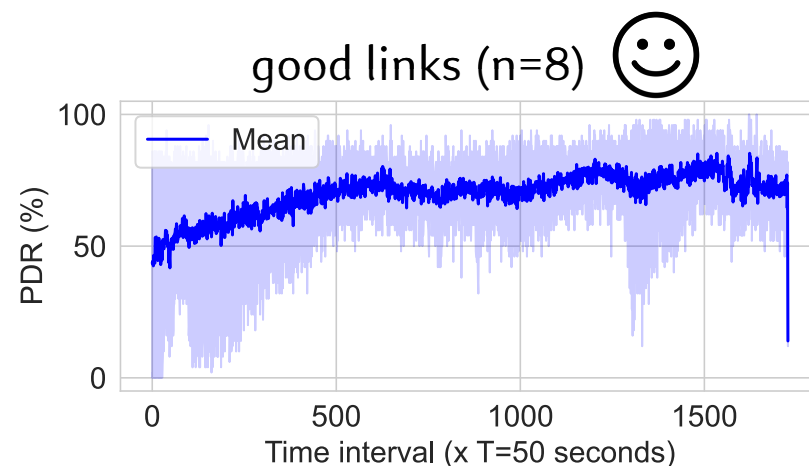
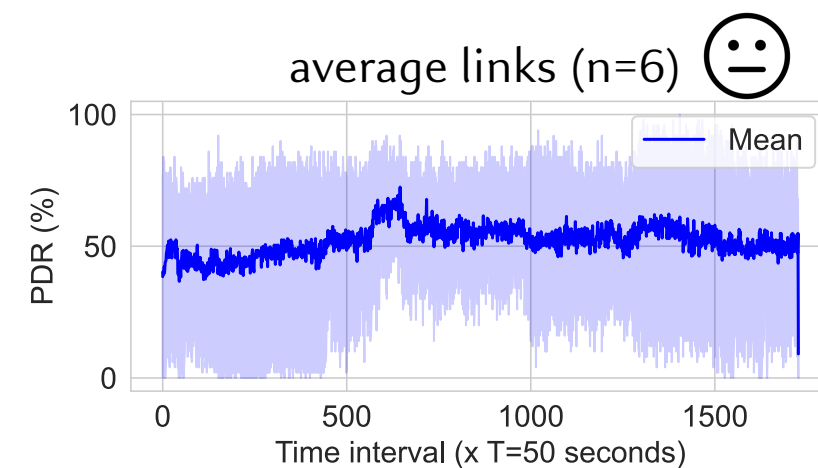
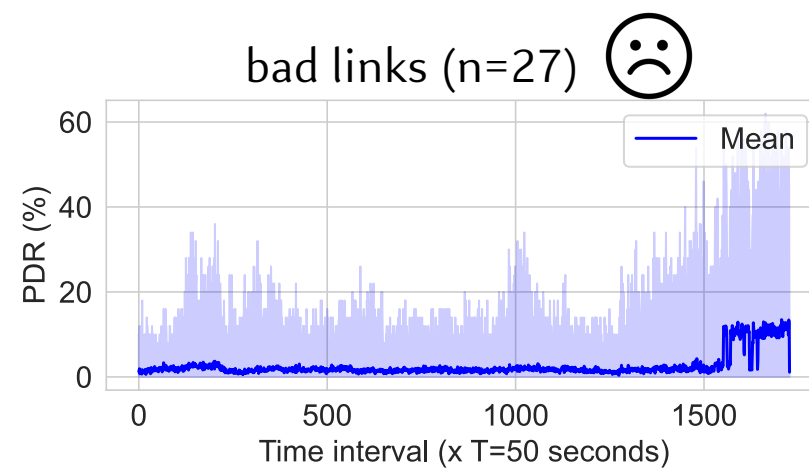
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➤ **Classification according to the mean PDR: 0% - 20% - 60% - 75% - 100%**

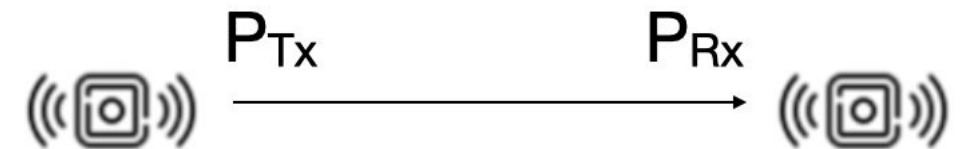


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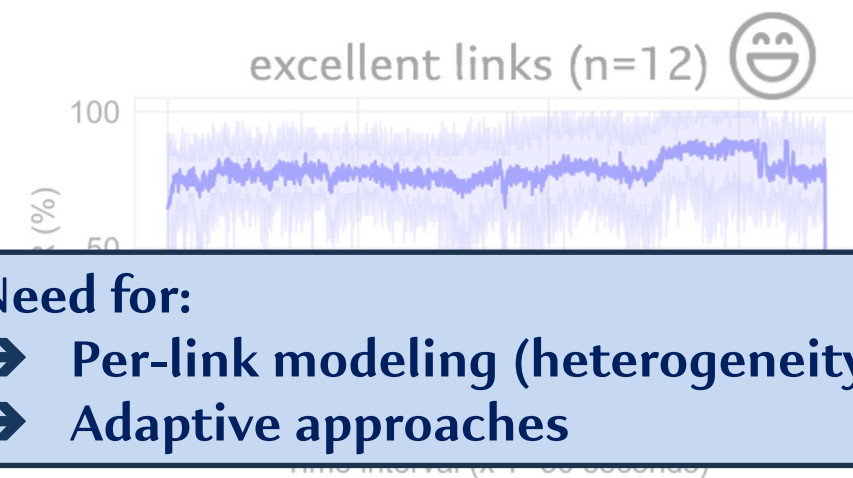
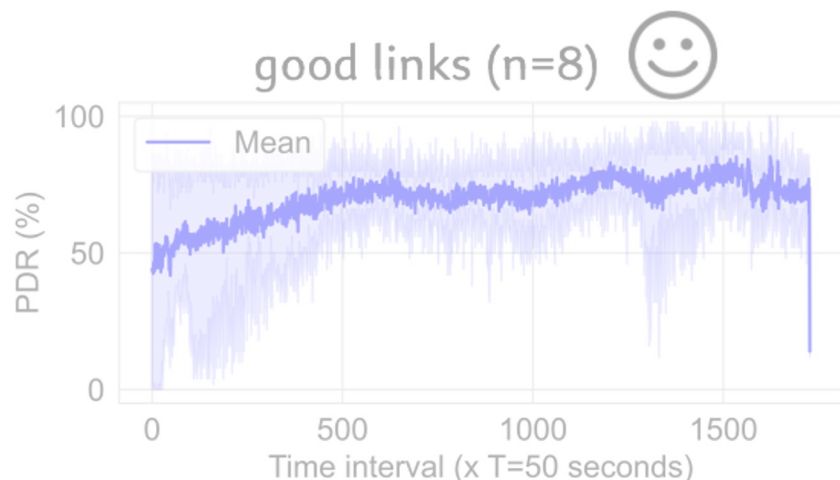
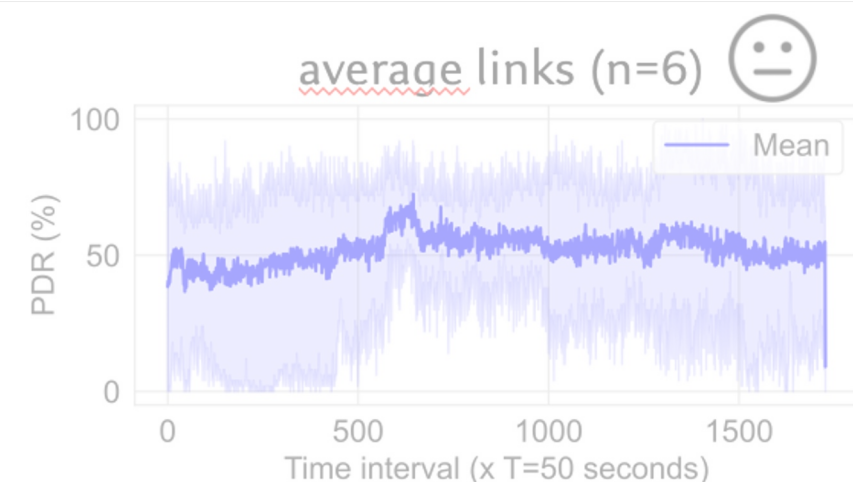
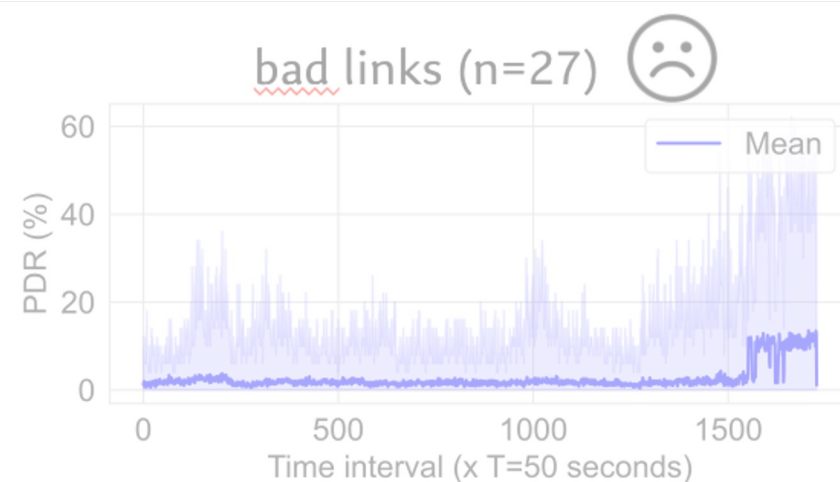
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Need for:

- ➔ **Per-link modeling (heterogeneity)**
- ➔ **Adaptive approaches**

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Problem Statement

- ❖ **How to accurately predict the evolution of a KPI (PDR) in a network?**
- ❖ **State of the Art:**
 - **Ray-tracing** approaches [3];
 - **Empirical studies** [4, 5] to generate propagation models;
 - **ML-based heuristics** [6, 7] for predicting performances.
- ❖ **Limits:**
 - Ignoring the links heterogeneity – **Modeling the network as a whole**
 - Absence of adaptivity in the modeling.

[3] Ruah, C. et al., (2024). Calibrating wireless ray tracing for digital twinning using local phase error estimates. *IEEE Transactions on Machine Learning in Communications and Networking*.

[4] Brun-Laguna, K., Minet, P., Watteyne, T., & Gomes, P. H. (2019). Moving beyond testbeds? Lessons (we) learned about connectivity. *IEEE Pervasive Computing*, 17(4), 15-27.

[5] Baccour, N. et al., (2012). Radio link quality estimation in wireless sensor networks: A survey. *ACM Transactions on Sensor Networks (TOSN)*, 8(4), 1-33.

[6] Sindjoun, M. L. F., & Minet, P. (2019, November). Wireless link quality prediction in iot networks. In 2019 PEMWN (pp. 1-6). IEEE.

[7] Benadji, H., Zitoune, L., & Vèque, V. (2023, December). Predictive modeling of loss ratio for congestion control in IoT networks using deep learning. In GLOBECOM 2023. IEEE.

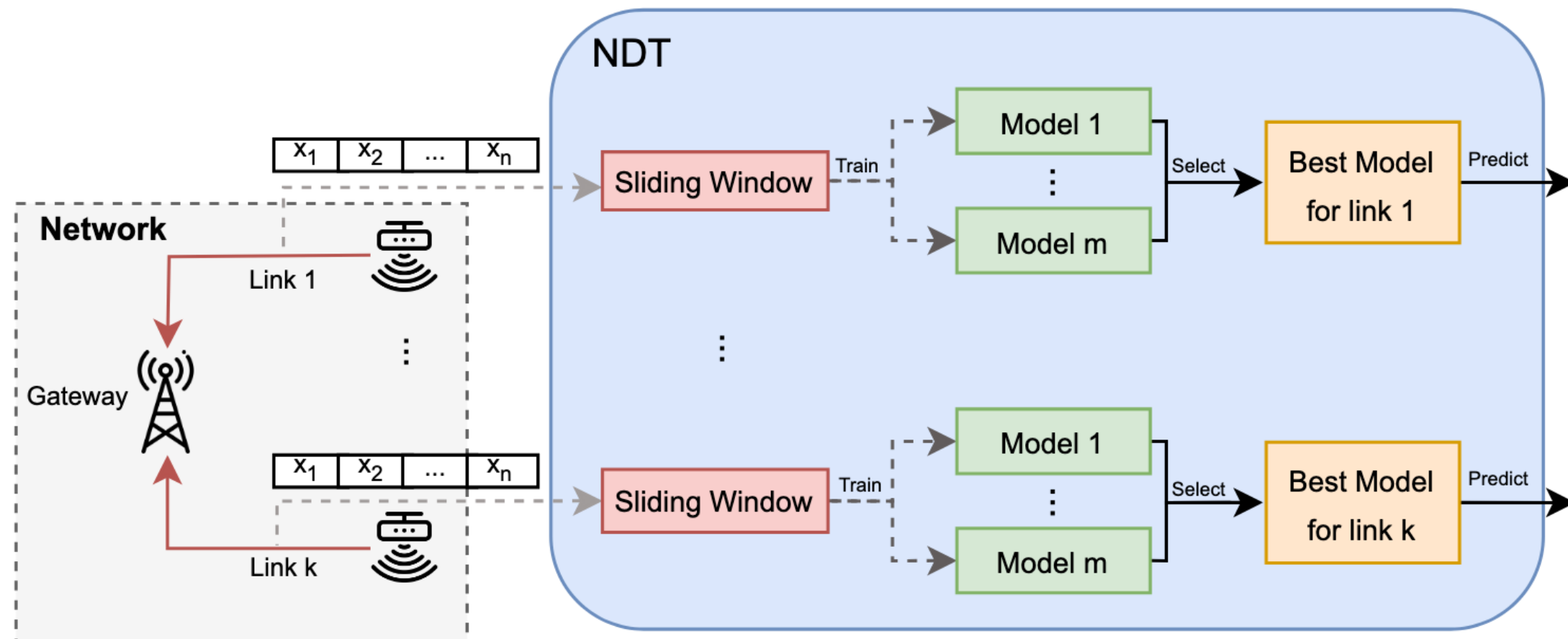
Proposed Solution

❖ Contribution: Data-driven approach

1. Replace the **physical layer** simulation by measurements
2. **Separately** model each radio link
3. **Continuously** select the best model for each link

❖ Interest:

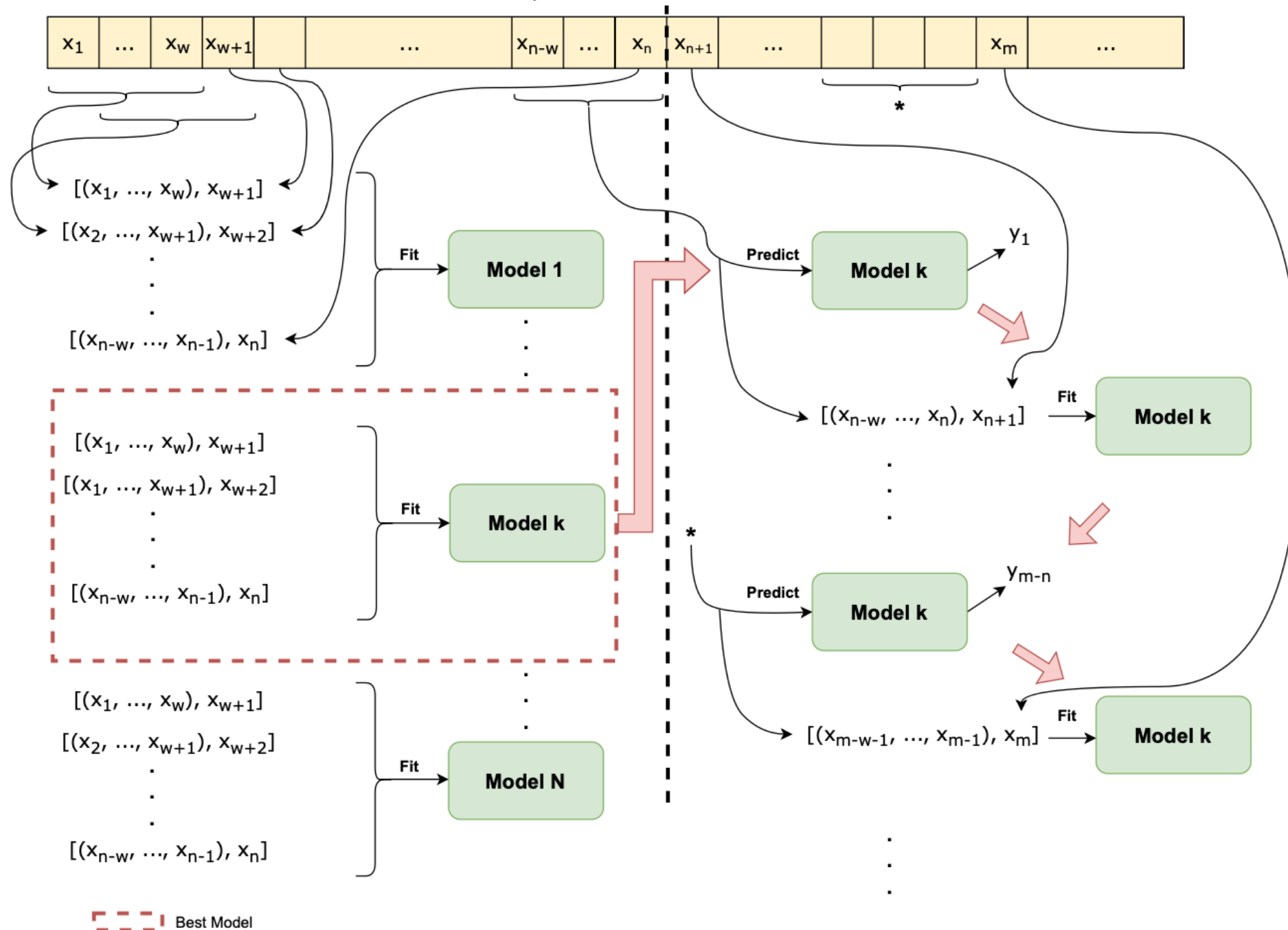
- Flexibility
- Temporality (anomaly Detection, etc.)



Proposed Solution – One Step

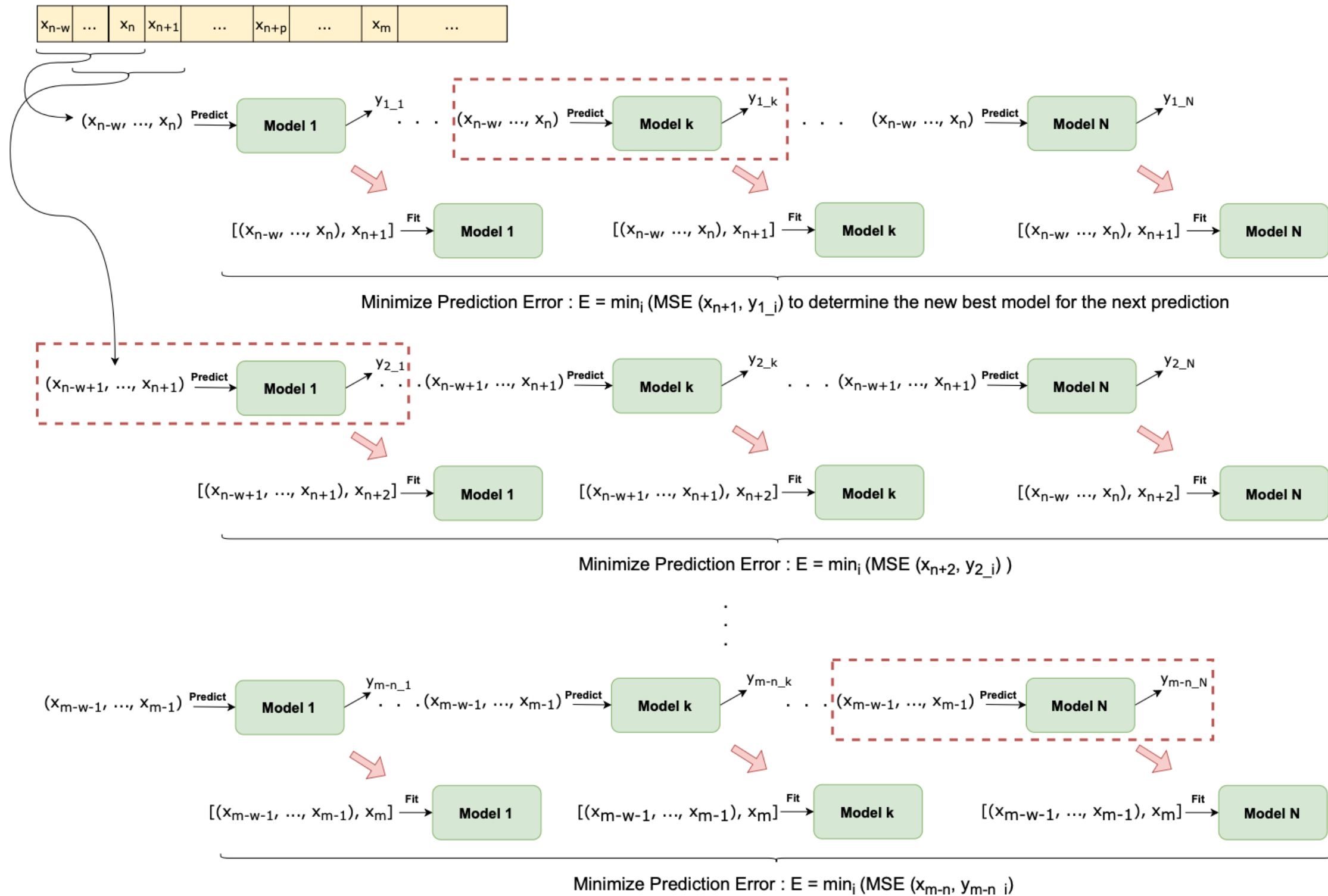
❖ **Fixed Model:** For each model:

- Split between Training and Testing data
- Create sliding windows for the regression
- Select the best model to use for predictions



Proposed Solution – One Step

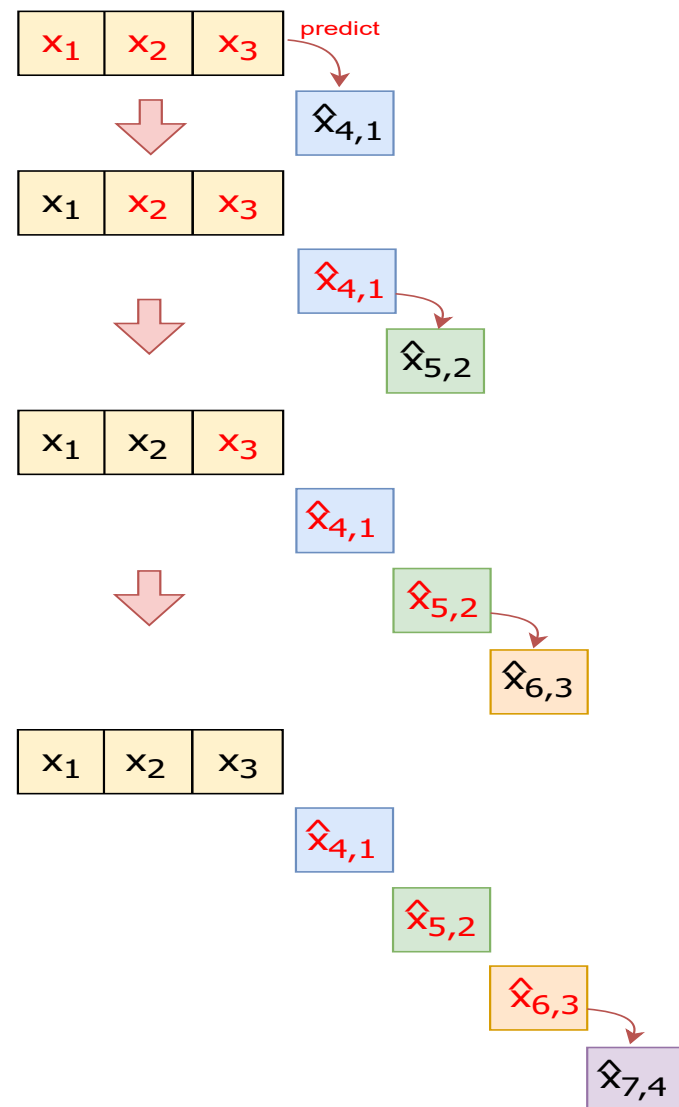
❖ Adaptive Model:



Proposed Solution – Multi-Step

❖ Multi-step prediction:

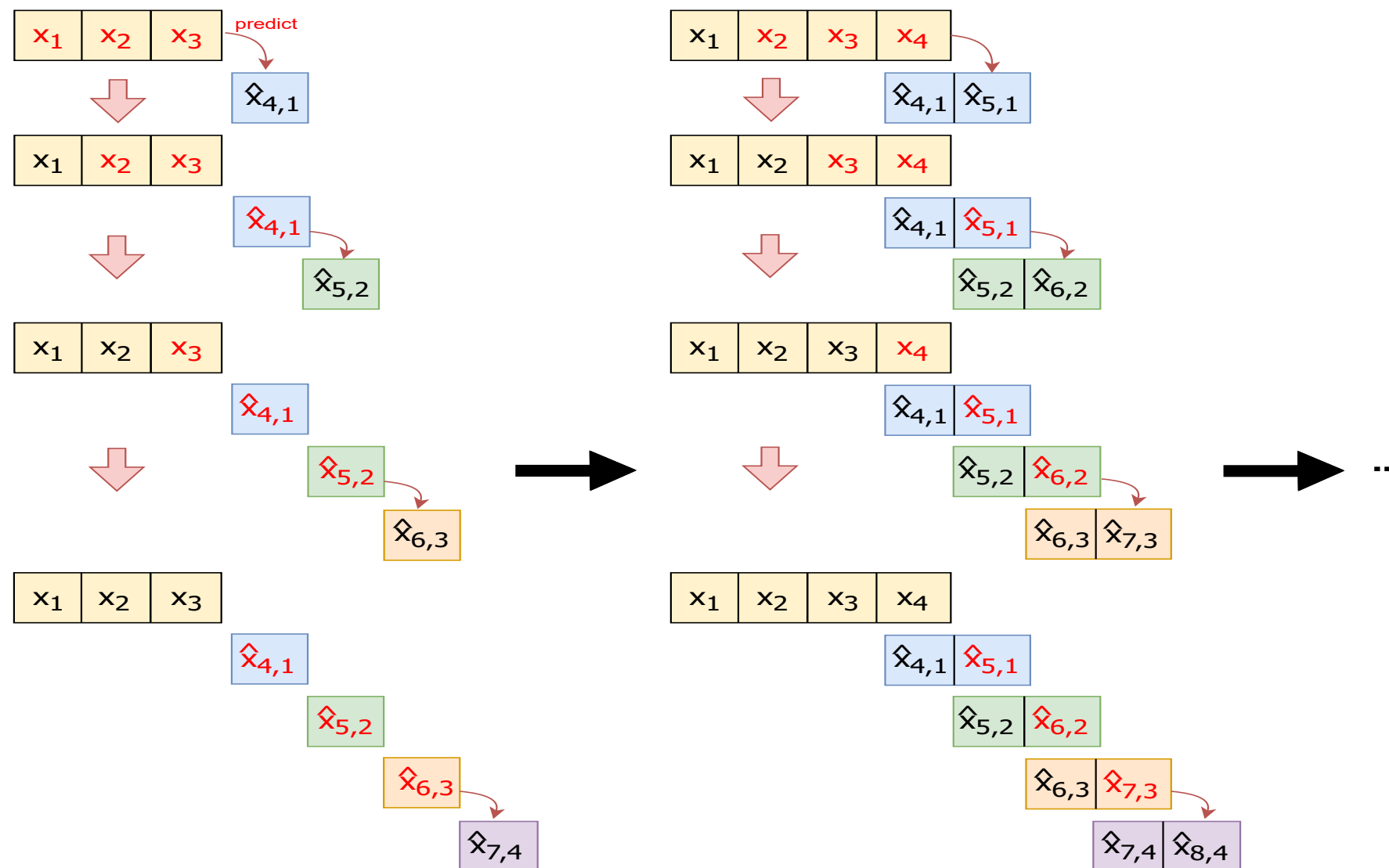
- Distinguish between the steps predictions for the evaluation



Proposed Solution – Multi-Step

❖ Multi-step prediction:

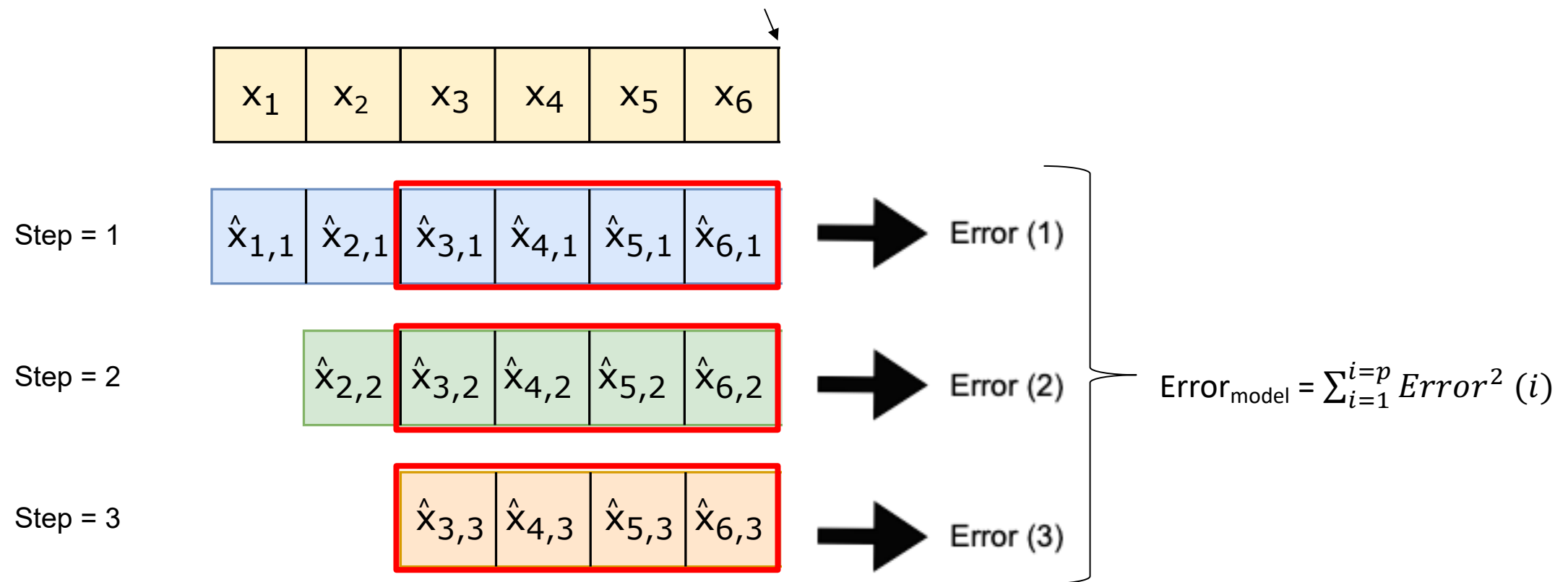
- Distinguish between the steps predictions for the evaluation



Proposed Solution – Multi-Step

❖ Error computing for the Adaptive approach:

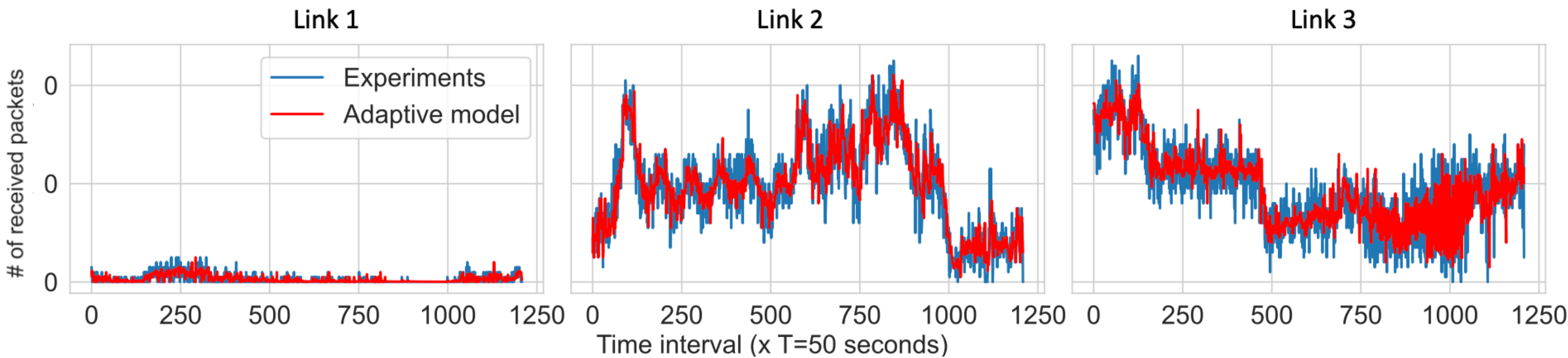
- At each interval, compute the error for the last q (e.g., $q=4$) predictions, according to the different steps errors of each model:



❖ Choose Best Model = $argmin (Error_{model})$ for the next prediction

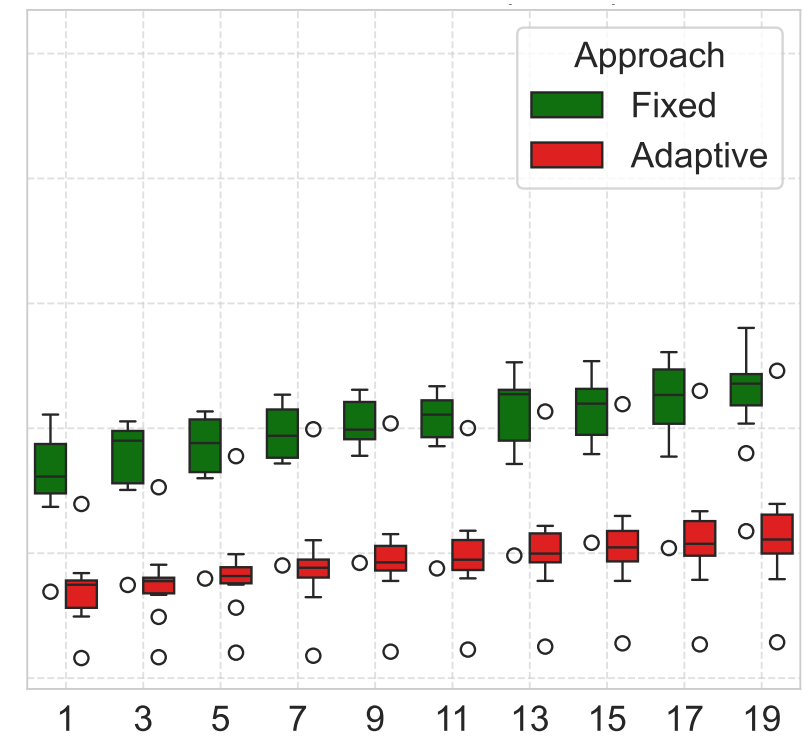
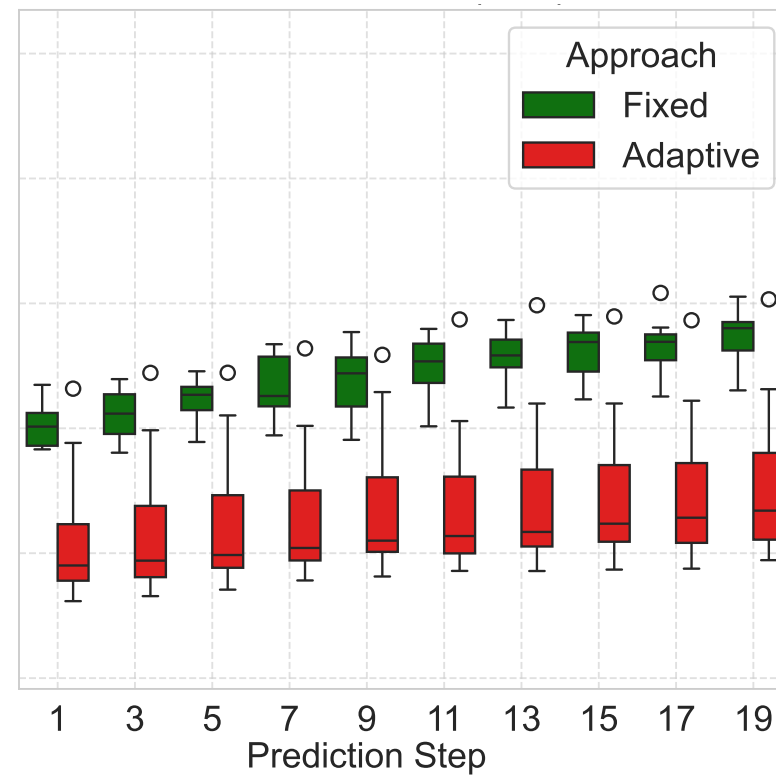
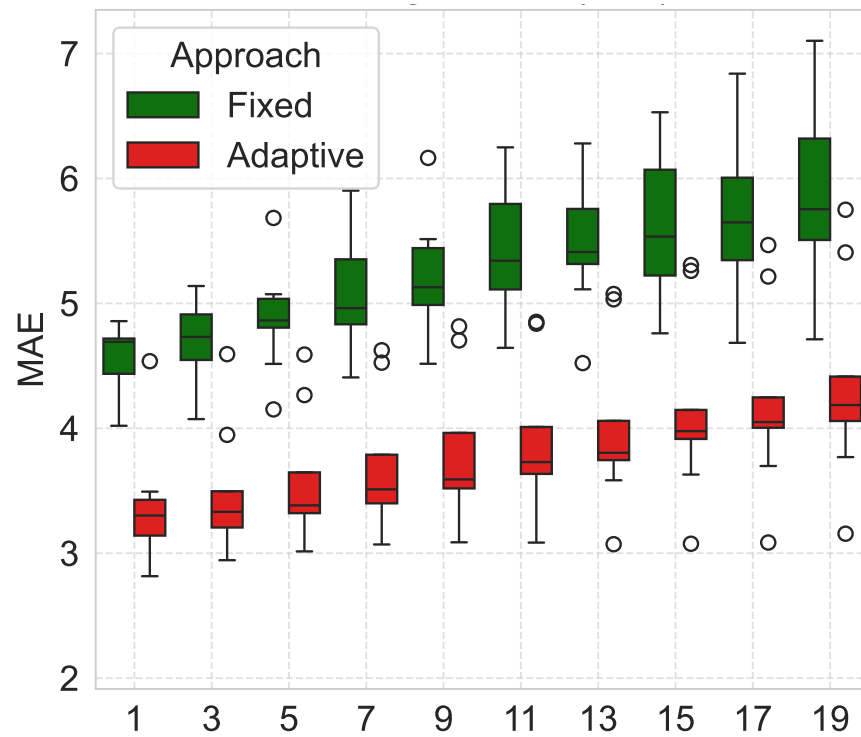
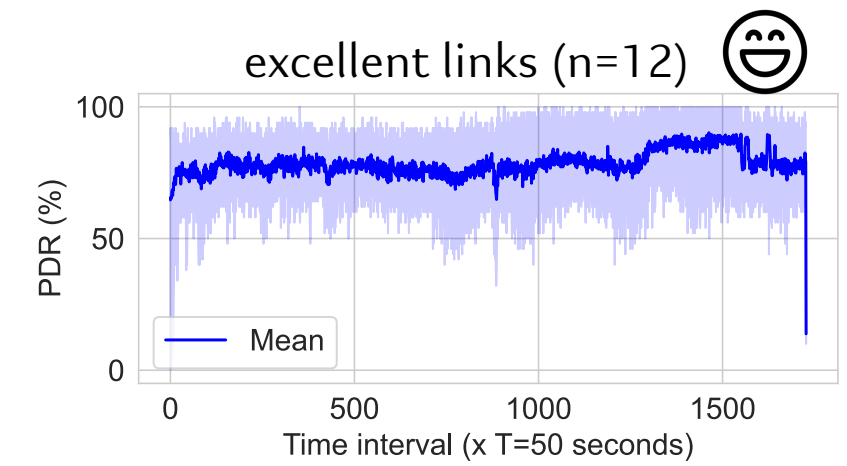
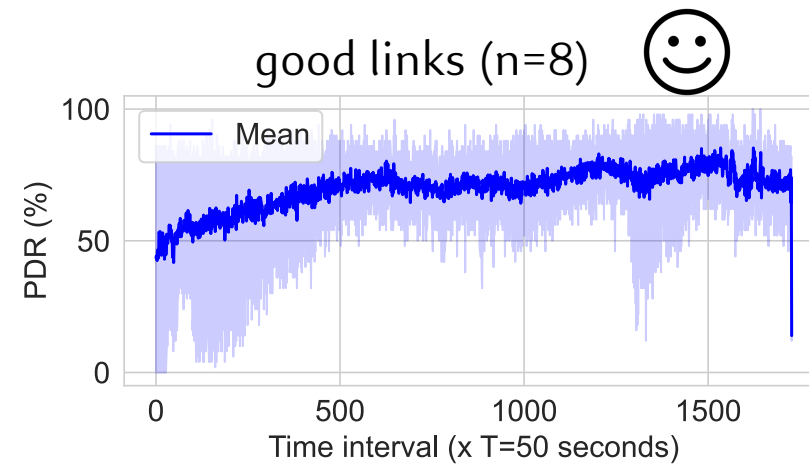
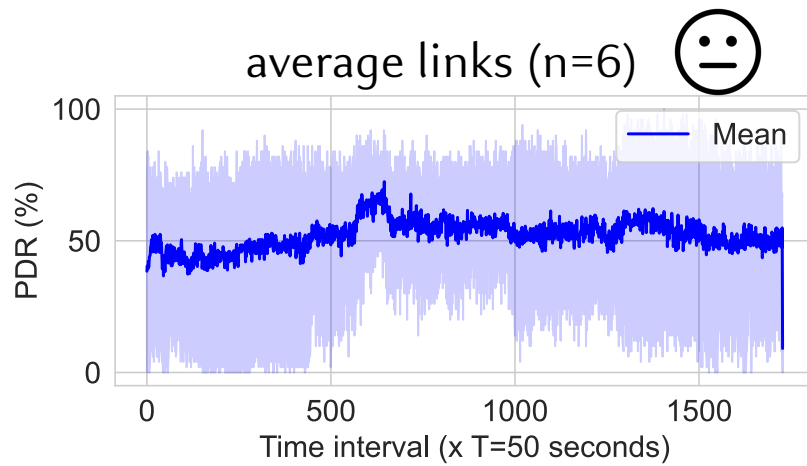
Results

❖ Short-term predictions ($p=1$):



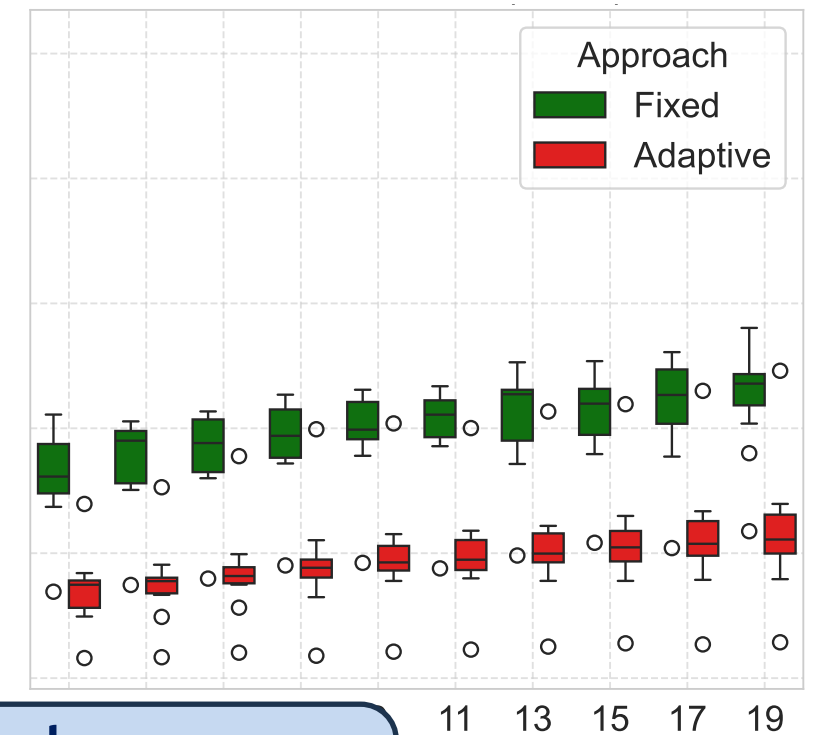
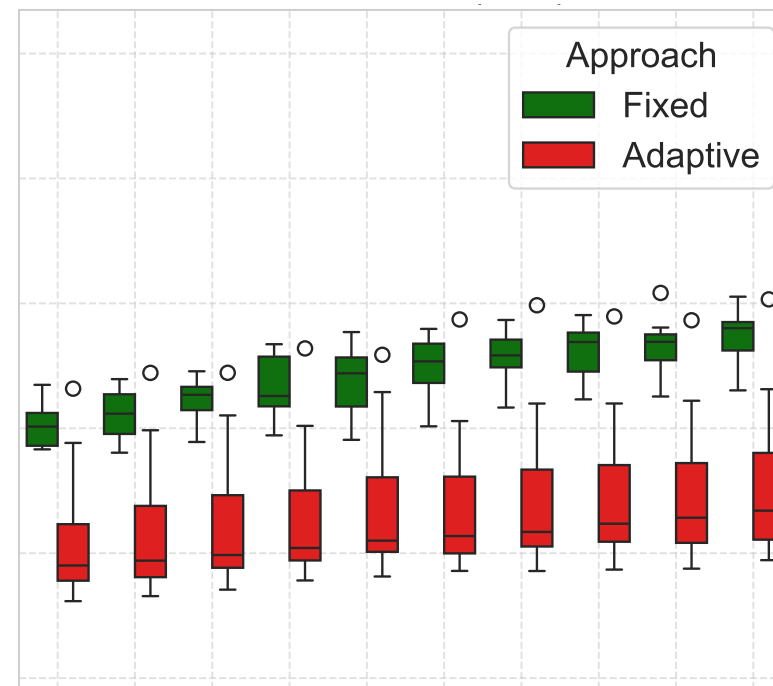
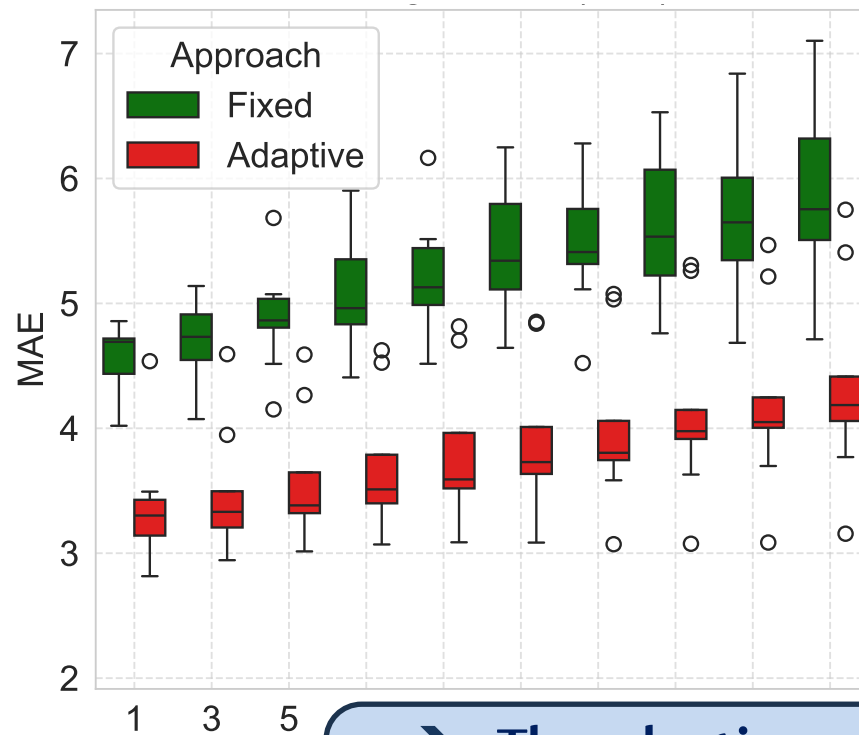
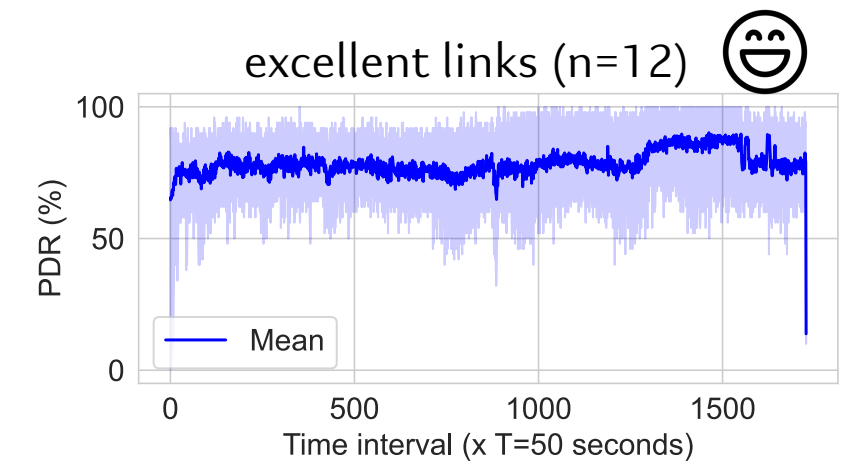
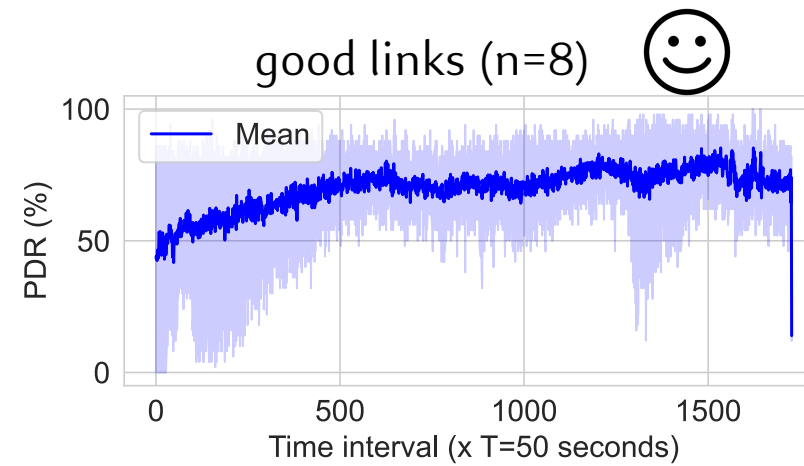
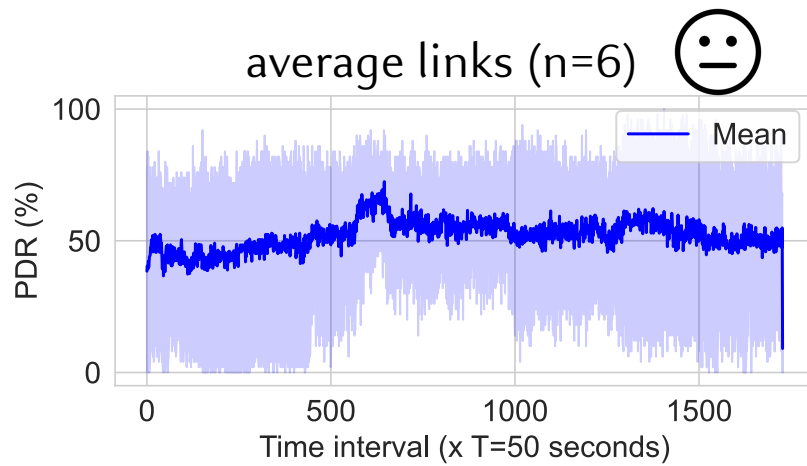
Results

❖ Mid-term predictions ($p > 1$):



Results

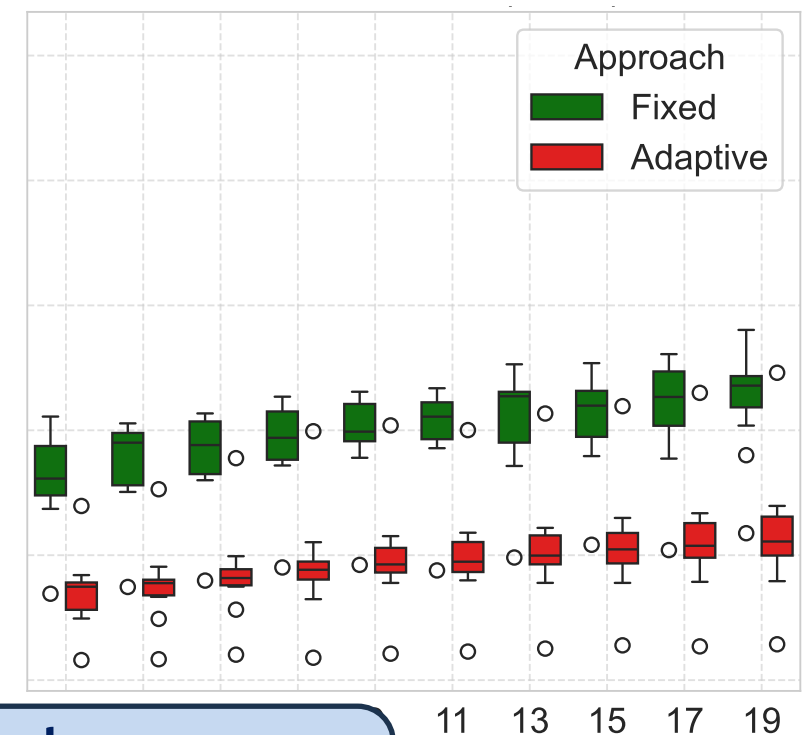
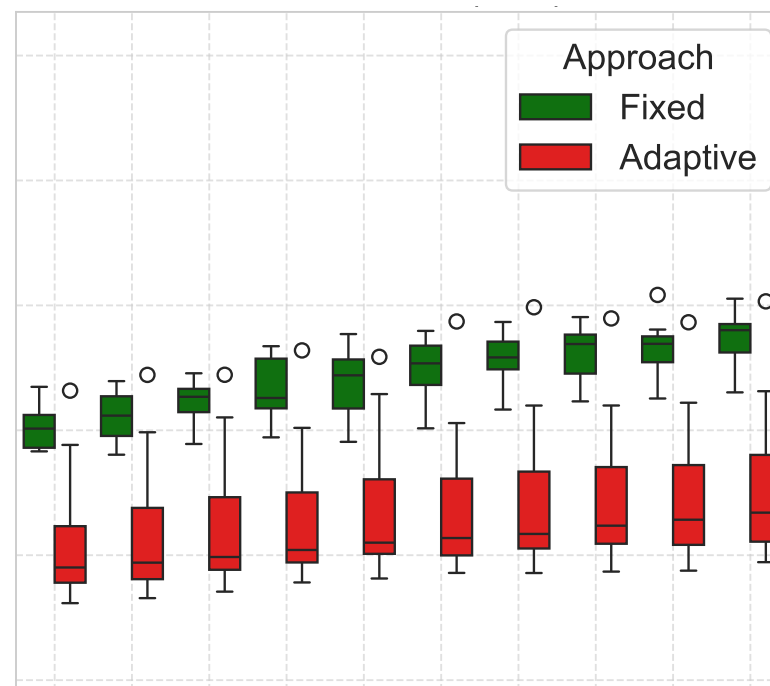
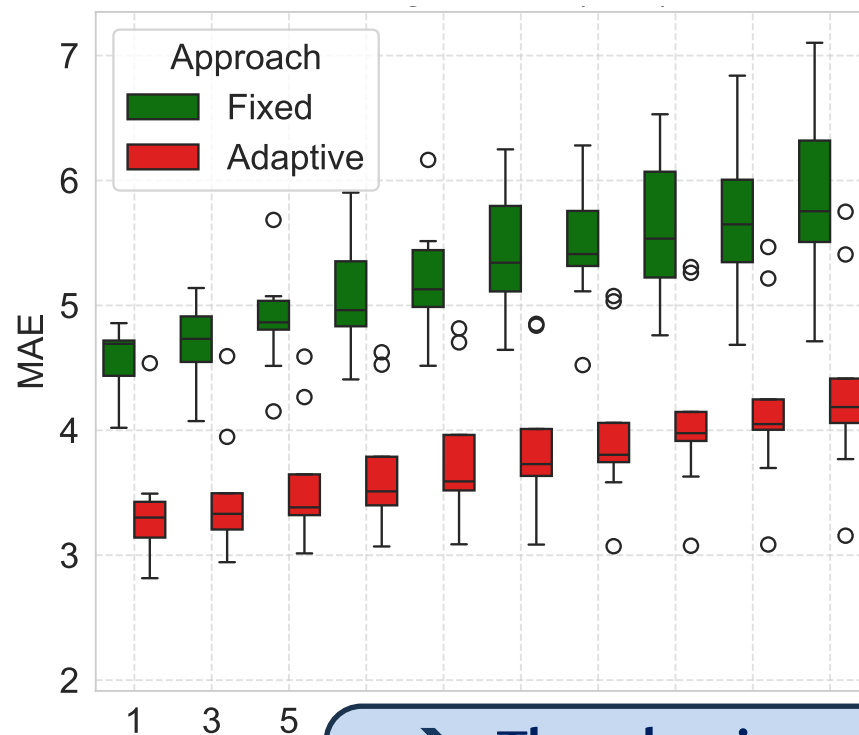
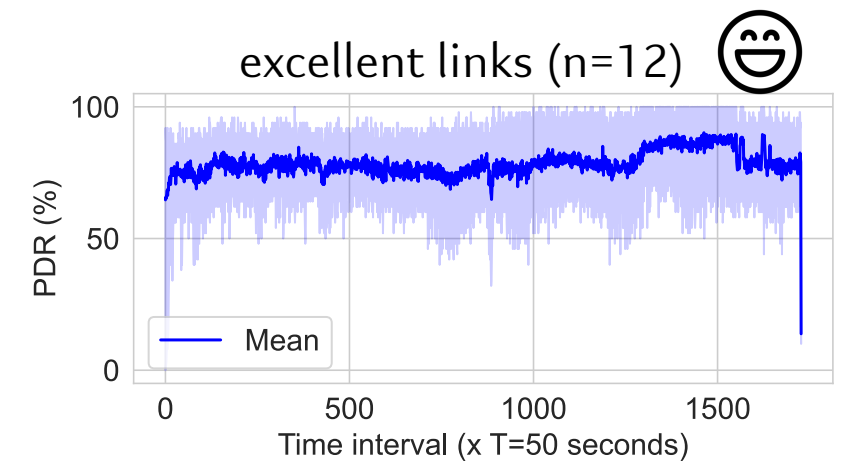
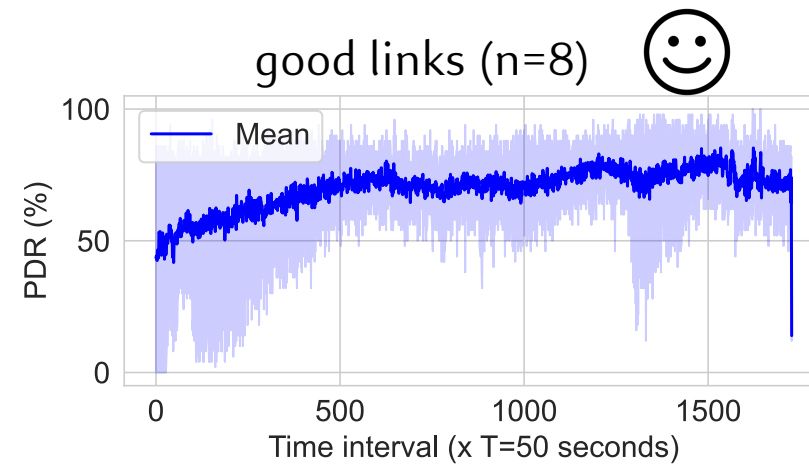
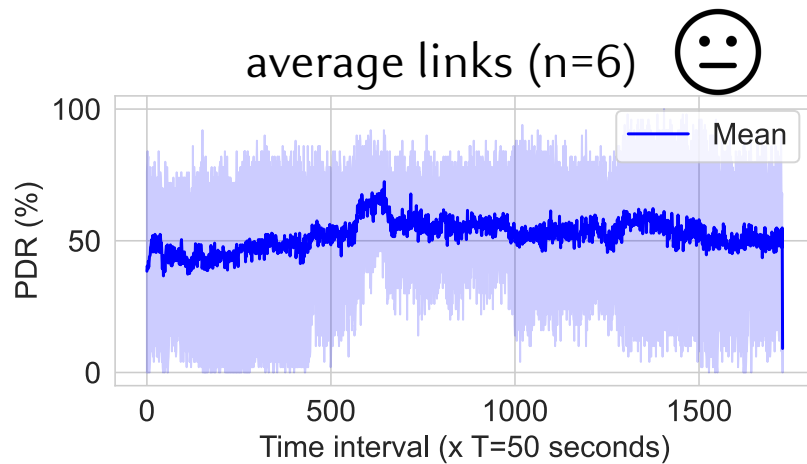
❖ Mid-term predictions ($p > 1$):



➔ The adaptive approach always outperforms the fixed one

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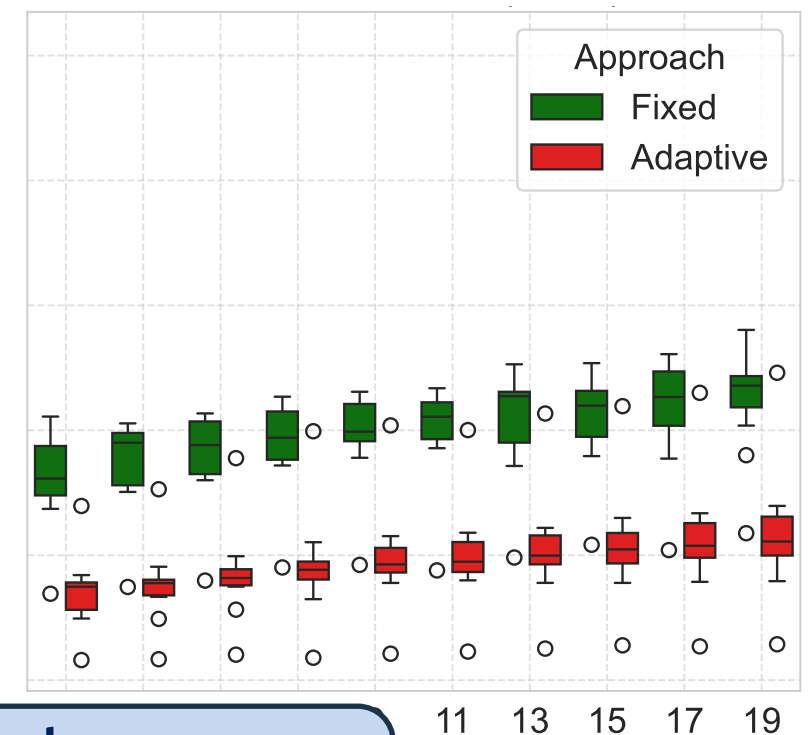
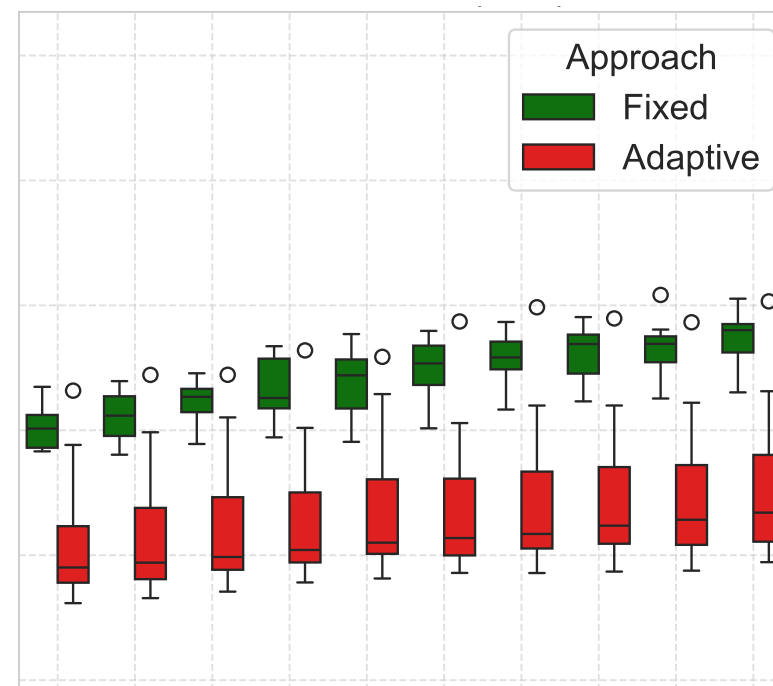
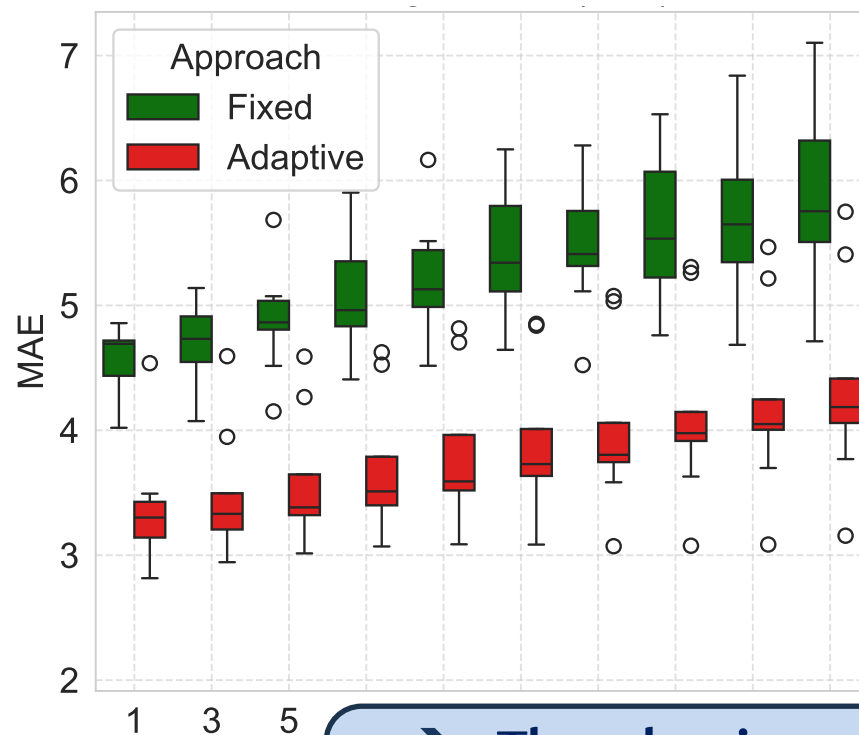
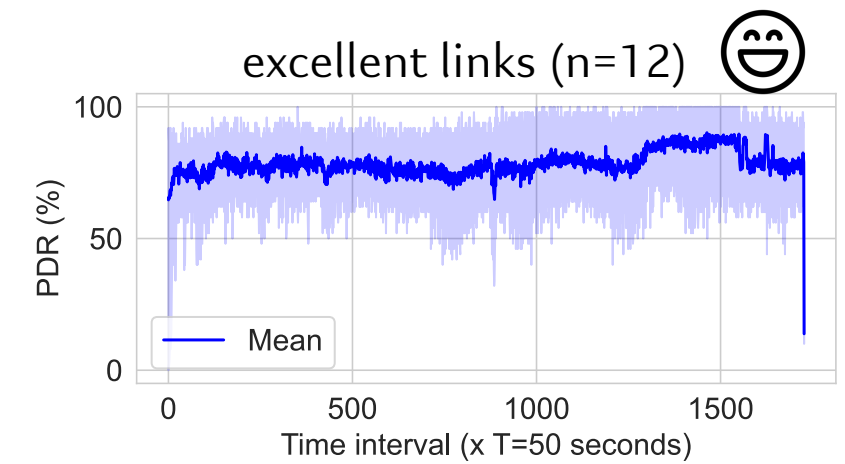
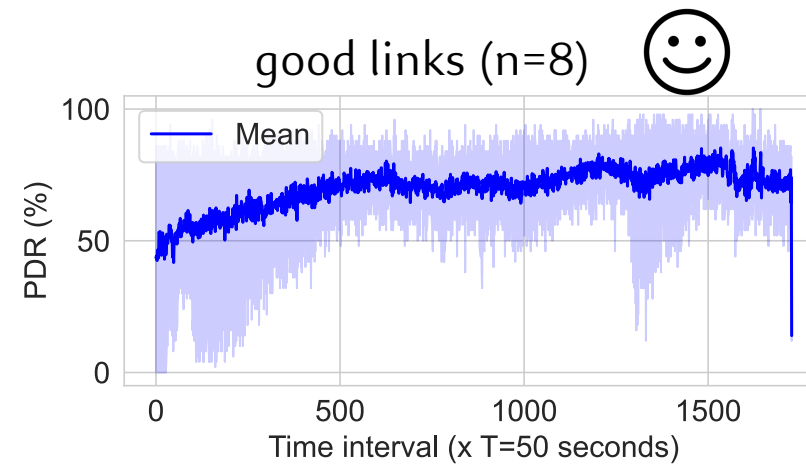
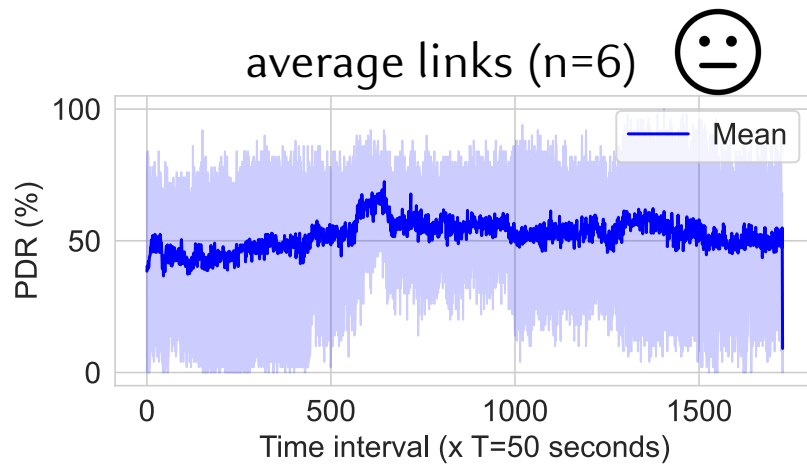
❖ Mid-term predictions ($p > 1$):



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- ➔ Low prediction errors for short/mid term, with a linear increasing

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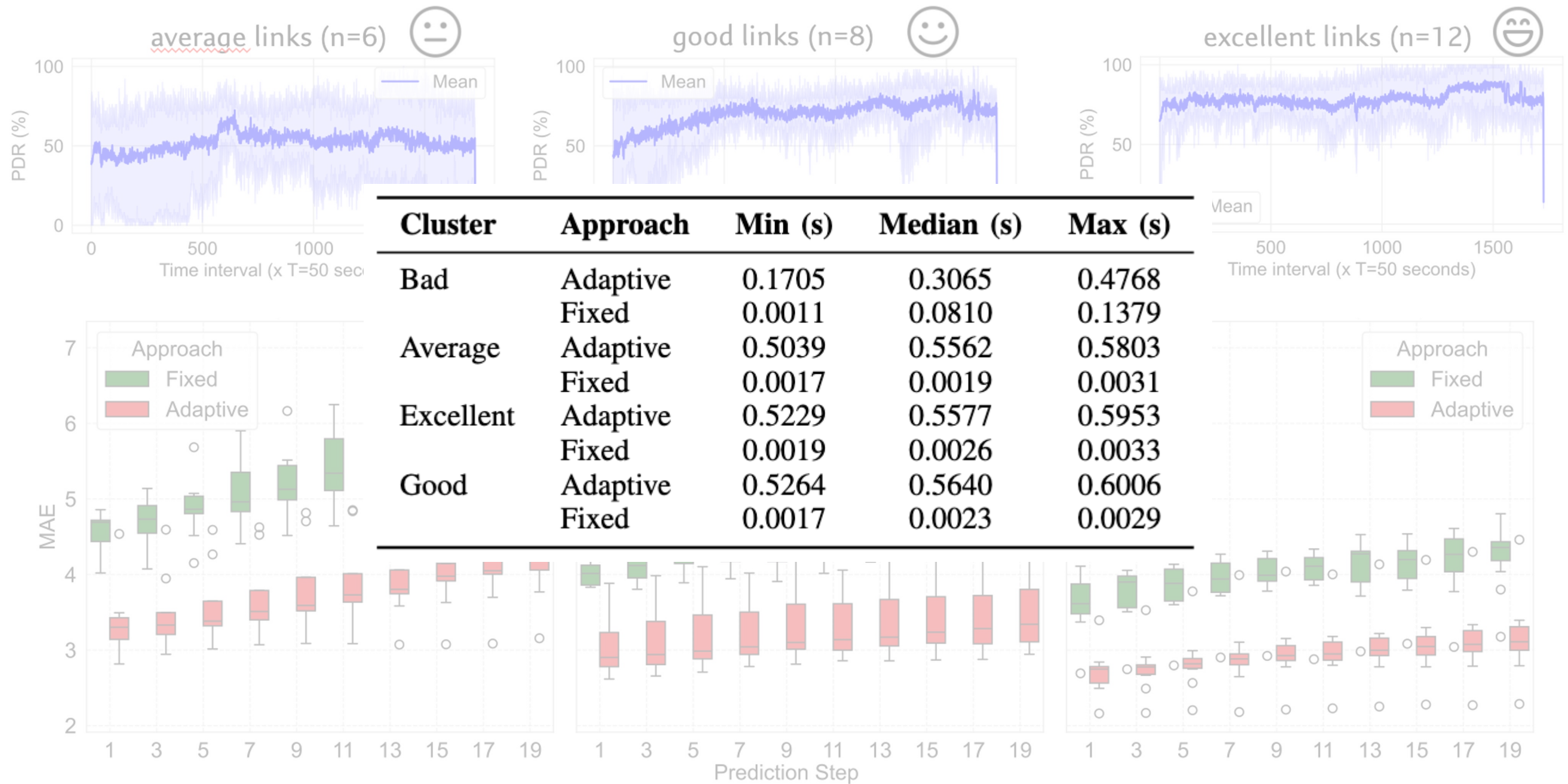
❖ Mid-term predictions ($p > 1$):



- ➔ The adaptive approach always outperforms the fixed one
- ➔ Low prediction errors for short/mid term, with a linear increasing
- ➔ Better predictions for stable links

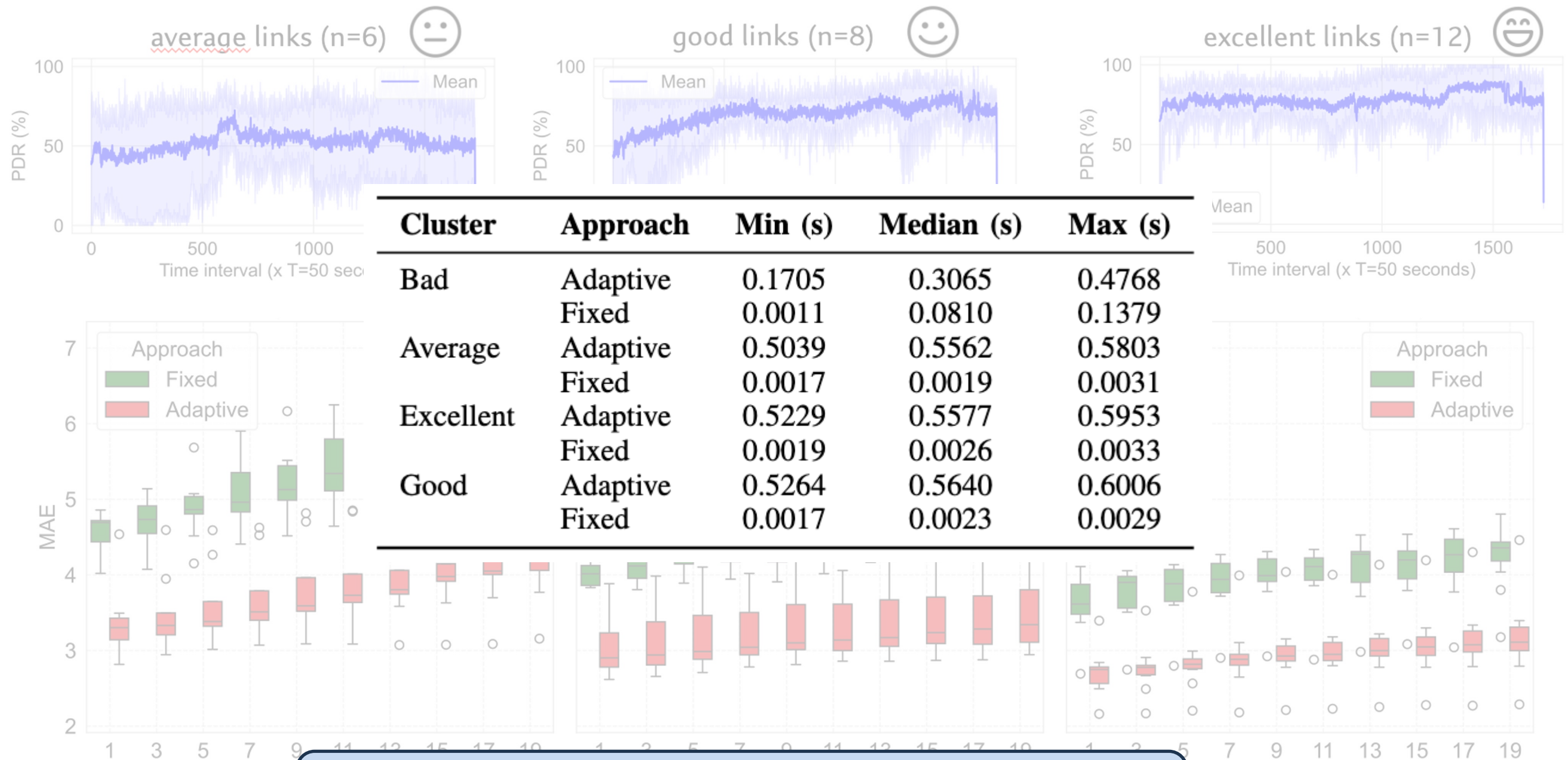
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❖ Mid-term predictions ($p > 1$):



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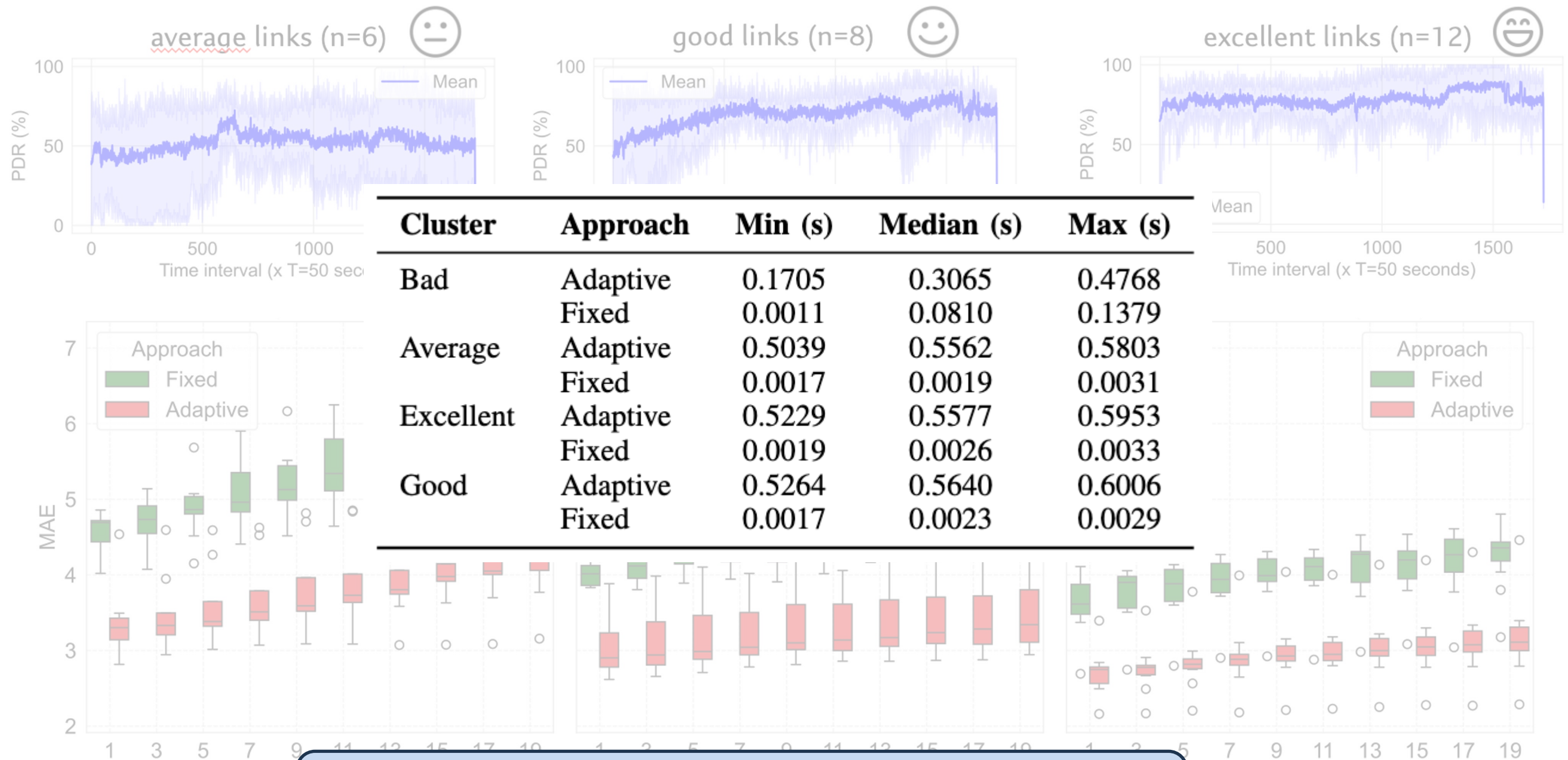
❖ Mid-term predictions ($p > 1$):



➔ Low prediction times for both approaches

Results

❖ Mid-term predictions ($p > 1$):



- ➔ Low prediction times for both approaches
- ➔ Accuracy comes at a cost...

Contributions

❖ Individual link modeling

- Captures radio links heterogeneity
- Uses lightweight models

❖ Adaptive algorithm for model selection

- Captures radio links dynamics
- Showcases tradeoff between accuracy and complexity

❖ Completely reproducible experiments and results:

<https://github.com/SamirSim/Wireless-Link-Quality-Prediction>

Takeaways & Perspectives

Takeaways:

- ❖ Importance of deployment conditions
 - **Need for reproducible experimental data**
- ❖ Network heterogeneity and dynamics
 - **Need for per-link modeling in network performance evaluation**
 - **Need for adaptive approaches**

Perspectives:

- ❖ Metrology with a dynamic traffic?
- ❖ Scalability?
- ❖ Generalization capabilities?

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Thank you for your attention!