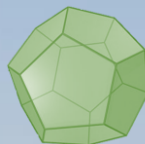




University
of Glasgow



School of Computing Science
Essence: Pervasive &
Distributed Computing

QUALITY-AWARE PREDICTIVE MODELLING & INFERENCE ANALYTICS AT THE NETWORK EDGE

PhD Thesis – Viva – 25.08.2021

Natascha Harth

Supervisors:

Dr. Christos Anagnostopoulos

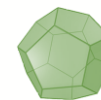
Prof. Dīmītrios Pezaros

**WORLD
CHANGING
GLASGOW**





Motivation



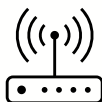
Adaptation



Bandwidth



Computation



Connectivity



Energy



Real-Time



Privacy



Storage



Cloud



Edge



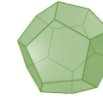
Feature



Limitation



Why Quality-Aware?



Use Cases



Autonomous
Car



Health



Smart Home



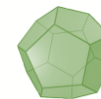
Speech
Recognition

Poor Quality Consequences

- Wrong Decision Making
- Costly Mistakes
- Harmful Outcome
- Decrease User Satisfaction
- Horrible Customer Experience
- Incorrect Conclusions



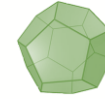
It is essential to have high
quality analytical results.



Communication-Efficient Data Forwarding	Latency-Efficient Local Analytics	Privacy-Efficient Local Learning
<ul style="list-style-type: none">• Reducing the frequency of data transmission by:<ul style="list-style-type: none">– Prediction-based forwarding– Time-Optimised delaying• Reconstruction of missing values (imputation)	<ul style="list-style-type: none">• Familiarity-based retraining of Local ML Model• Performance-based ML Model forwarding• Qualitative model selection for query analytics:<ul style="list-style-type: none">– Quantisation Input/Error-Space– Similarity-based Clustering	<ul style="list-style-type: none">• Enable <i>privacy by design</i> using Federated Learning• Combine generalisation and personalisation at the Edge• Qualitative local model selection:<ul style="list-style-type: none">– Adaptive Weighting– Optimal Time of Switching



Hypothesis 1

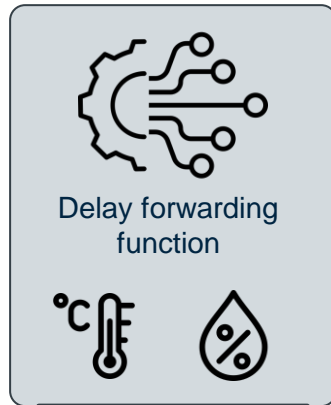


“Enabling computation at the Edge using efficient forwarding mechanism will reduce the communication, relieve the bandwidth and save energy at the device. Smart reconstruction functions at the central coordinator enable qualitative analytical results”

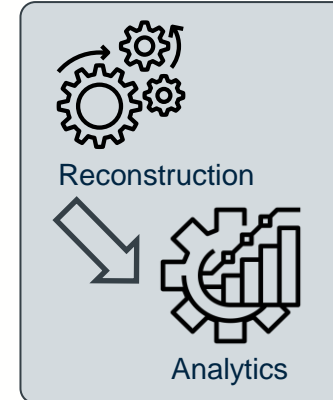
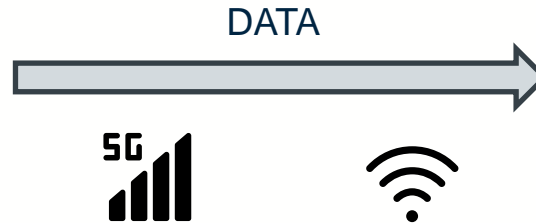
(1) Computational Intelligence

(2) Reduce Communication

(3) Quality-Aware Reconstruction



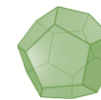
Sensing Device



Coordinator



Quality-Efficient Data Forwarding



Data Forwarding Mechanism

Prediction Based:

- SPD (Single Prediction Design)
- DPD (Dual Prediction Design)
- QEPD (Quality-Efficient Prediction Design)

Time-Optimised:

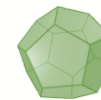
- TOFS (Time-Optimised Forwarding Strategy)
- HTOFS (Hybrid Time-Optimised Forwarding Strategy)

Reconstruction Policies

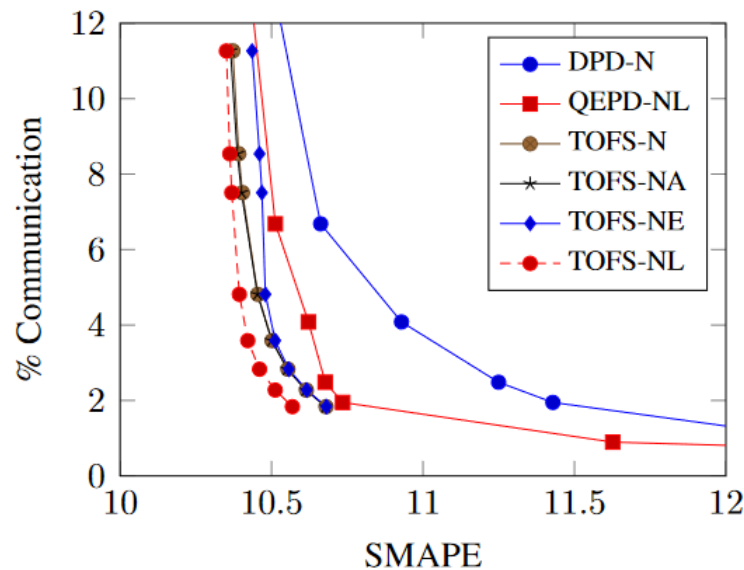
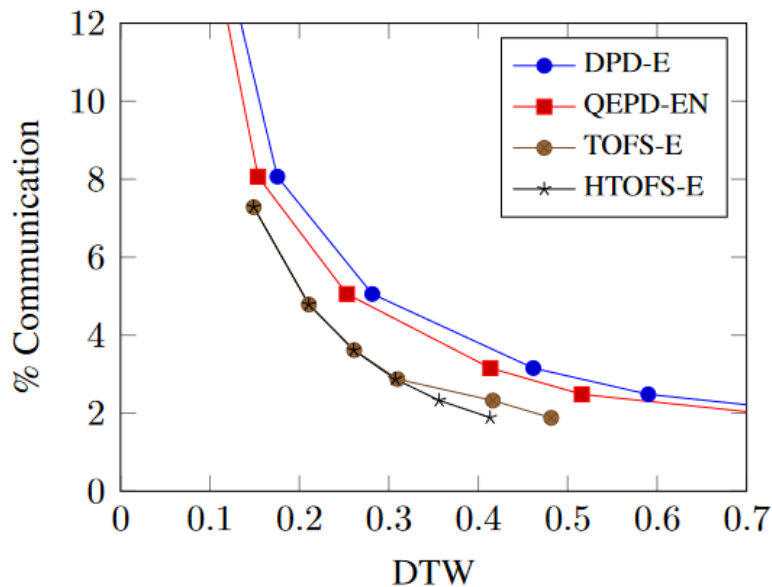
- NAIVE
- MEAN
- EWMA
- ARIMA
- LMS



Performance Evaluation

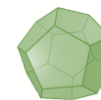


"Time-Optimised Data Forwarding Strategy reduces the communication while generating high quality results for analytical tasks"



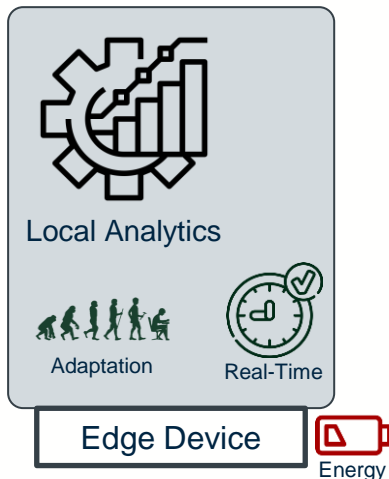


Hypothesis 2

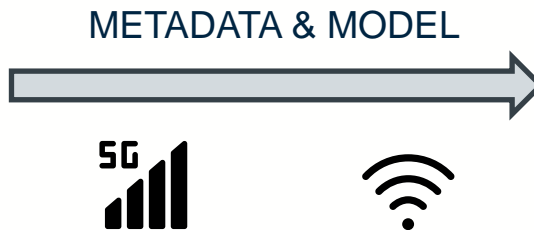


“Enabling analytics locally at edge devices will empower real-time applications that can adapt intelligently to concept drifts and changes of the continuous data arriving. These locally trained models can be selected through qualitative model selection methodologies at central coordinators”

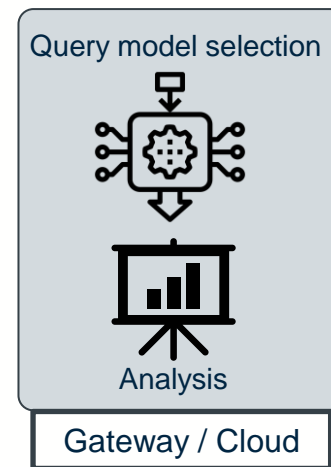
(1) Model Retraining



(2) Model Forwarding

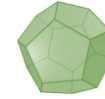


(3) Qualitative Model Selection





Latency-Efficient Edge-Centric Analytics



Retraining of Model based on novelty of input

Quantisation of the input
space and error as decision

Forwarding model based on

- Prediction Performance
- Model Fitting
- Parameter Divergence

Model Selection at the central coordinator

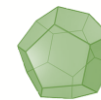
Based on the quantisation of
the Input/Error-Space
selecting Top-*K* models:

- IAM
- IEAM

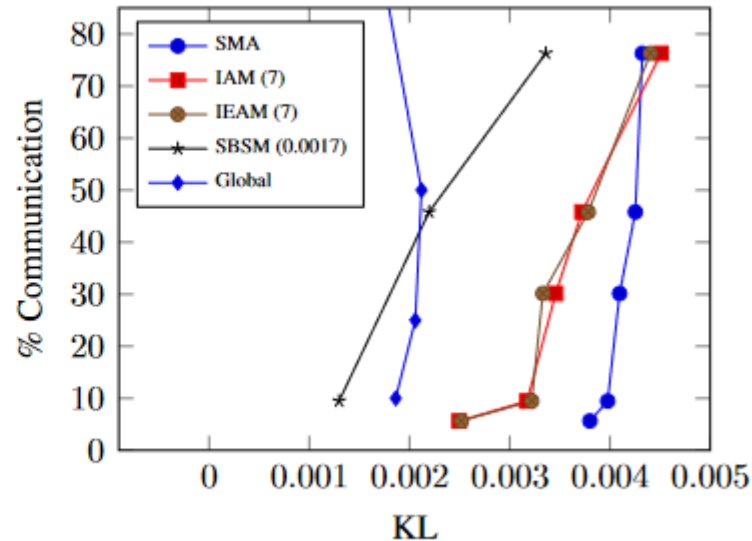
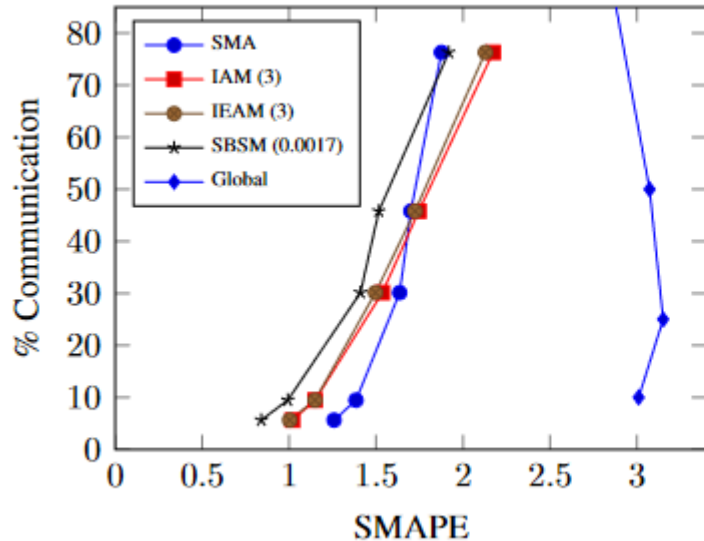
Based on similarity-based
clustering (SBSM)



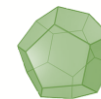
Performance Evaluation



“Similarity-based model selection for queries is generating high qualitative results for analytical task while reducing the communication using the predictive performance forwarding methodology”



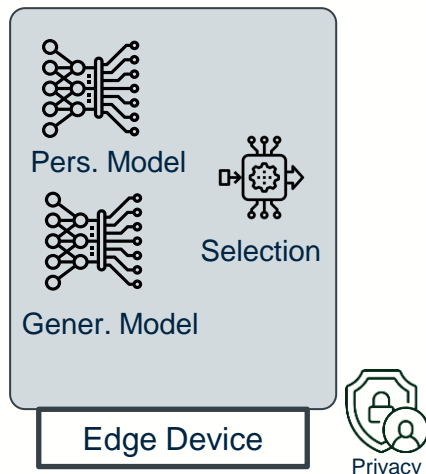
Hypothesis 3



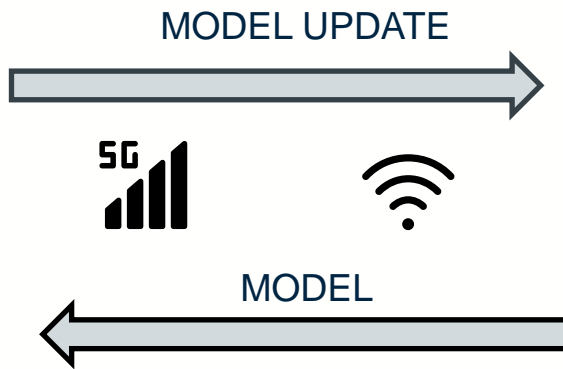
“Generalised models over privacy-preserved data will not provide qualitative results in constantly changing and heterogeneous environments. Local learned models and an intelligent model selection mechanism or weighting of personalisation and generalisation enables qualitative prediction results”

(1) Local personalised model

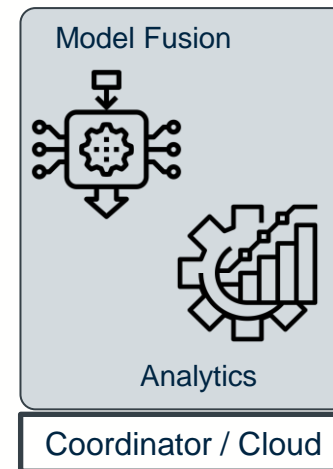
(4) Select best model for prediction



(2) Send generalised model

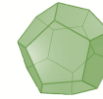


(3) Update generalised model





Privacy-Efficient Learning at the Edge



Federated Learning to enable privacy-preserving of data by design

Continuously evolving and changing environments do not converge,
therefore personalisation and generalization is needed

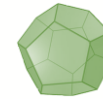
What model is the best to use for a prediction locally:

Adaptive Weighting of personal
and generalised model

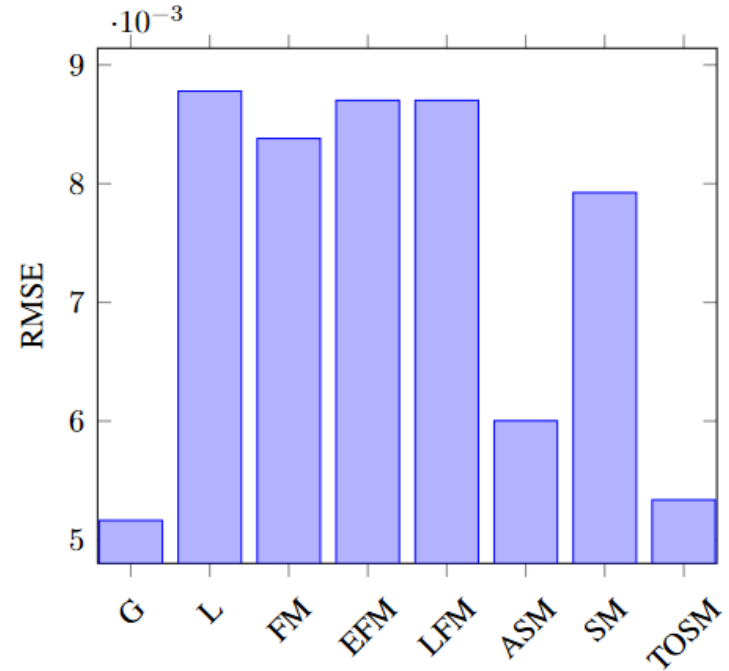
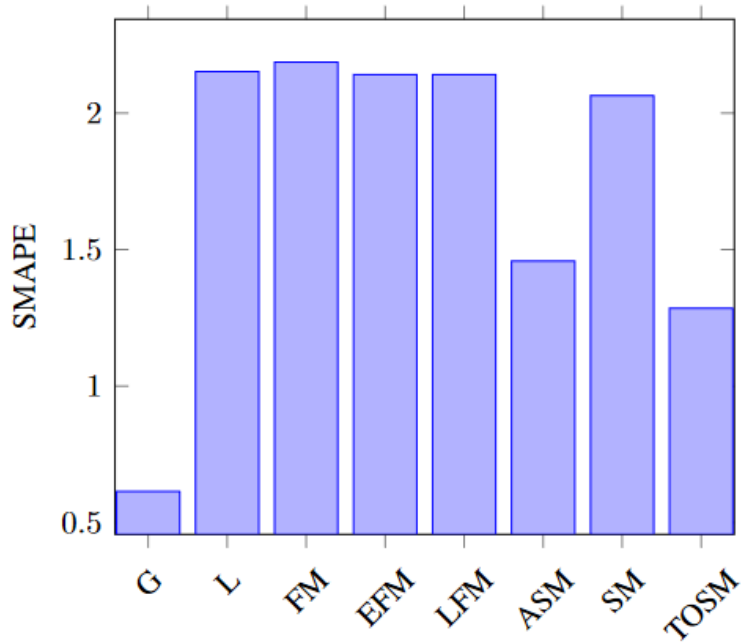
Identifying the optimal time
to switch the models



Performance Evaluation

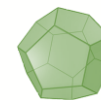


“Time-Optimised model selection and adaptive weighting of models is generating qualitative better results than the federated or local model”





Summary



Applications able to handle raw-data while trying to increase lifetime of devices and reducing bandwidth.

TOFS



Applications important to perform local real-time decision making but the need of central query analysis.

SBSM

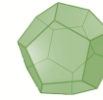


Applications important to enable privacy and decentralized learning over continuous evolving data.

ASM & TOSM



Next Steps



PhD Thesis



... over multiple hierarchies



... with other security mechanism



... on unsupervised learning tasks

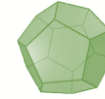


... using active learning

Federated Learning:



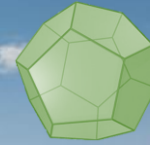
Publications



- Natascha Harth, and Christos Anagnostopoulos. "Edge-centric efficient regression analytics." *IEEE International Conference on Edge Computing (EDGE 2018)*. San Francisco, CA, USA, July 2018, pp. 93-100
- Natascha Harth, and Christos Anagnostopoulos. "Quality-aware aggregation & predictive analytics at the edge." In: *IEEE International Conference on Big Data (IEEE Big Data 2017)*. Boston, MA, USA, Dec.2017, pp. 17-26
- Natascha Harth, Christos Anagnostopoulos, and Dimitrios Pezaros. "Predictive intelligence to the edge: impact on edge analytics". *Evolving Systems*, 9(2), Aug. 2017, pp. 95-118. DOI:10.1007/s12530-017-9190-z
- Natascha Harth, Kostas Delakouridis, and Christos Anagnostopoulos. "Convey intelligence to edge aggregation analytics." *New Advances in the Internet of Things*. Springer, Cham, June 2017. pp. 25-44. DOI: 10.1007/978-3-319-58190-3 2



University
of Glasgow



School of Computing Science
Essence: Pervasive &
Distributed Computing

**Thank you and looking
forward to the Questions!**

<http://www.dcs.gla.ac.uk/essence/>

*Pictures have been used from: freepik.com & flaticon.com