

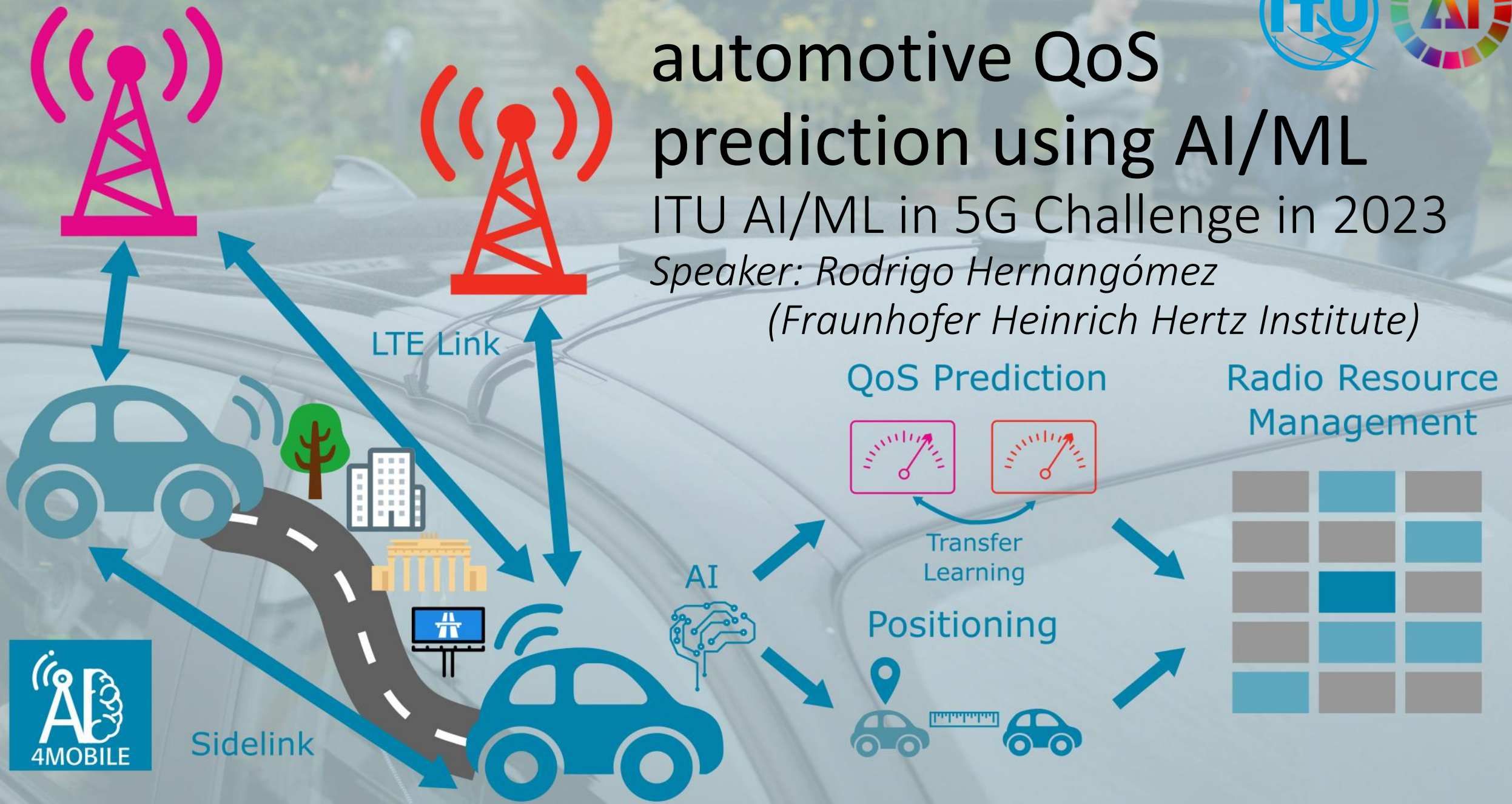
Multi-environment automotive QoS prediction using AI/ML



ITU AI/ML in 5G Challenge in 2023

Speaker: Rodrigo Hernangómez

(Fraunhofer Heinrich Hertz Institute)



AI4Mobile

AI-AIDED WIRELESS SYSTEMS FOR MOBILITY IN INDUSTRY AND TRAFFIC

- Research project funded by the German Ministry for Education and Research
- Consortium of 3 multinationals, 3 SMEs and 3 research institutions
 - Duration: 2020-2023
- **Research goal** – AI/ML for predictive Quality of Service (QoS) at high mobility
 - Twofold meaning of high mobility:
 - Traffic: Connected cars
 - Industry: Automated Guided Vehicles
- More info at ai4mobile.org

BMW
GROUP



ROLLS-ROYCE
MOTOR CARS LTD



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GÖTTING





















ENWAY

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AI4Mobile

DATASETS

- Goal: Test ML algorithms for predictive QoS
- Extensive measurement campaigns available at [IEEE Dataport](https://dataport.ieee.org/)
- Broad scope: Cellular/Sidelink/Campus network; Automotive/Industry

Datasets	Automotive 	Highway 	Sensor data 	Cellular 
	Industry 	Urban 	Campus-Net 	Sidelink 
<u>Berlin V2X</u>	Berlin Vehicle to Everything			    
<u>iV2I+</u>	industrial Vehicle to Infrastructure + Sensor			  
<u>iV2V</u>	industrial Vehicle to Vehicle			 

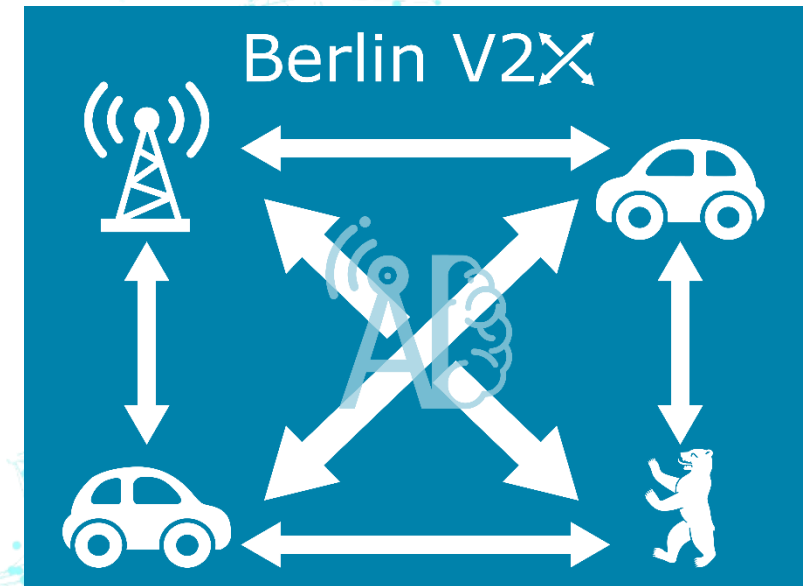
Multi-environment automotive QoS prediction with AI

OUTLINE

 Motivation & Challenges

 Dataset

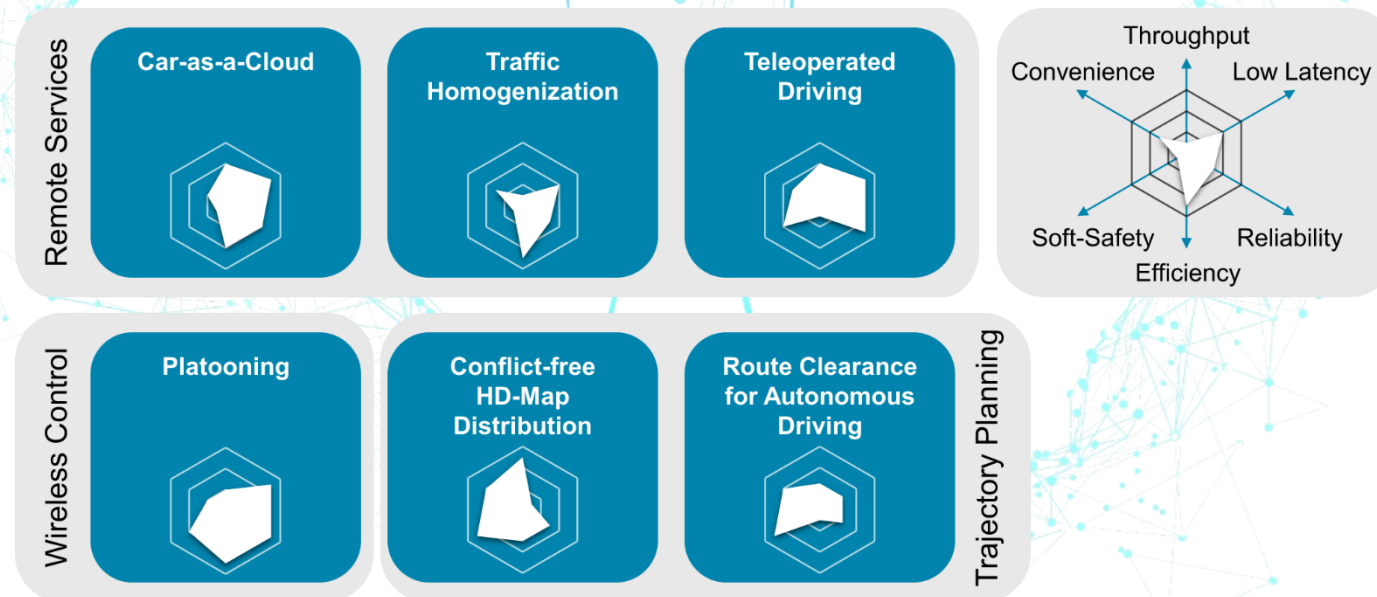
 Problem Statement



Multi-environment automotive QoS prediction with AI

MOTIVATION

- Vehicle-to-everything (V2X) communication at the core of new car services
- High demands on Quality of Service (QoS) and proactive resource allocation
 - Leverage boom of ML in communication networks for QoS prediction

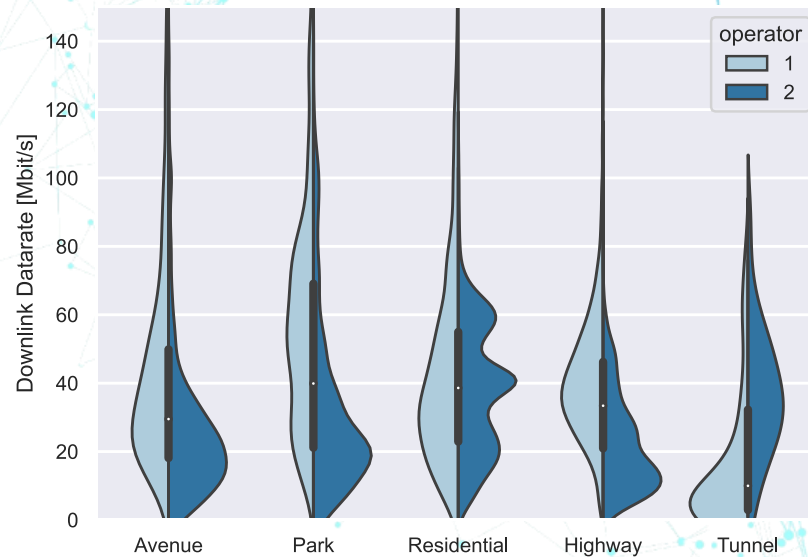


(Külzer *et al.*, 2021)

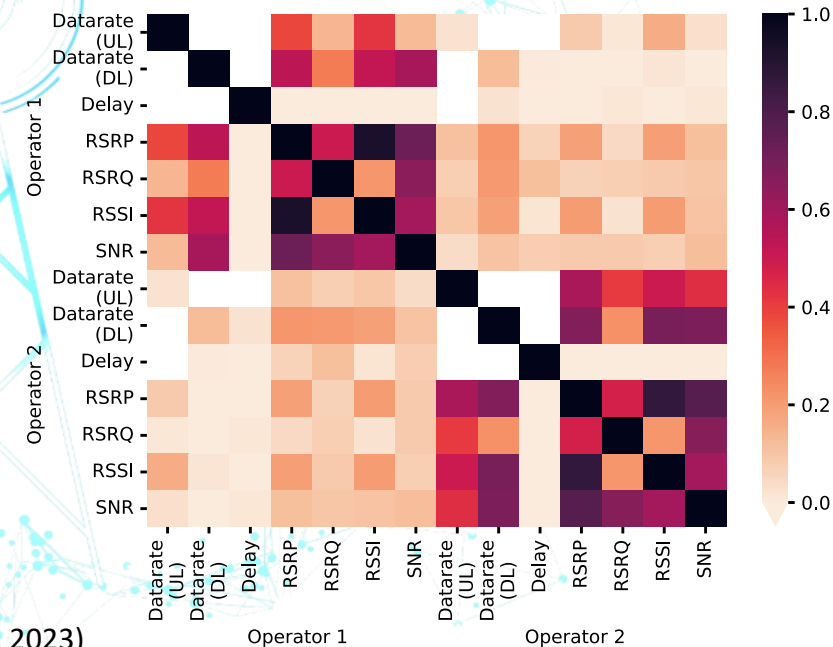
Multi-environment automotive QoS prediction with AI

CHALLENGES FOR AI/ML

- Drastic changes to the radio environment → Data distribution drifts (no i.i.d.)
- Generalization across different entities (vehicles/operators) or scenarios
 - Domain adaptation techniques might be helpful



(Hernangómez *et al.*, 2023)



BerlinV2X Dataset

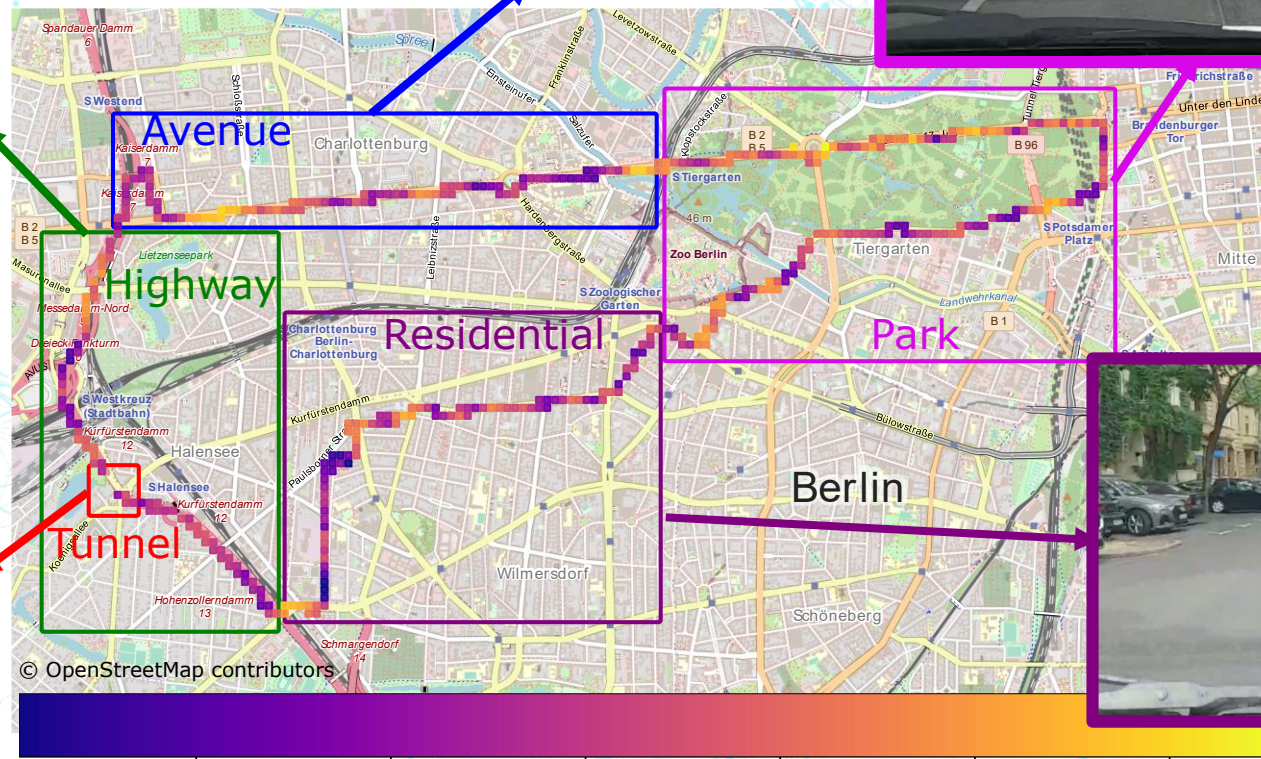
MEASUREMENT CAMPAIGN

- 2 commercial LTE mobile network operators
 - LTE frequency bands 700-2700 MHz
- V2V sidelink (3GPP Rel. 14, PC5 mode 4)
 - Out of the challenge's scope
- Simultaneous measurements from up to 4 vehicles
 - 2 vehicles per operator
 - Different drive modes: platoon vs. pair driving
 - 3 throughput profiles for diverse QoS measurements
- 17 rounds over 3 days
 - 1 round = 17 km across West Berlin
 - 45-60 minutes on a weekday
 - Diverse urban areas



BerlinV2X Dataset

URBAN AREAS



0 5 10 15 20 25
SNR [dB]

BerlinV2X Dataset

MEASUREMENT METHODOLOGY

- Full LTE stack captured with [MobileInsight](#) (MI)
 - (Li *et al.*, 2016)
- Traffic exchange with a server located at HHI
 - Datarate and jitter measurements with [iperf](#)
 - Ping-based delay measurements
 - Target datarate according to QoS measurement
- GPS localization
- Side information via APIs
 - Weather ([DarkSky](#))
 - Traffic conditions ([HERE](#))



Throughput profiles	QoS measurement	Target datarate
Low throughput	UL/DL delay	400 kbps
High throughput	UL datarate	75 Mbps
High throughput	DL datarate	350 Mbps

BerlinV2X Dataset

CAPTURED DATA

Data category	Source	Tool	Sampling interval	Features
LTE stack	In-vehicle device	MobileInsight	10 ms	PHY: SNR, RSRP, RSRQ, RSSI
			20 ms	PDSCH/PUSCH: RBs, TB Size, DL MCS, UL Tx Power
			Event-based	RRC: Cell Identity, DL/UL frequency, DL/UL bandwidth
Quality of Service	In-vehicle device	ping	1 s	Delay
		iperf	1 s	DL Datarate, Jitter
	Server	iperf	1 s	UL Datarate, Jitter
Position	GPS	NA	1 s	Latitude, Longitude, Altitude, Velocity, Heading
Side information	Internet database	HERE API	5 min	Traffic Jam Factor, Traffic Street Name, Traffic Distance
		DarkSky	1 hour	Cloud cover, Humidity, Precipitation Intensity & Probability, Temperature, Pressure, Wind Speed
Metadata	NA	NA	NA	Scenario, operator, drive type, target datarate, direction

BerlinV2X Dataset

PREPROCESSING

- Raw measurements as [parquet](#) files in [IEEE Dataport](#) (sources)
- Preprocessing code in [GitHub](#) under [preprocess](#)
 - Downsample to 1s
 - Extract relevant LTE params from MI (merge_mi_*.ipynb)
 - Merge on device, cell, and timestamp (merge_mi_all.ipynb)
 - Concatenate values for primary and secondary cells (carrier aggregation)
 - Merge all data on timestamp and device (merge_cellular.ipynb)
 - Select QoS data from server/device
 - Label with measurement metadata

DATASET FILES

cellular_dataframe.parquet	(24.43 MB)
sidelink_dataframe.parquet	(25.27 MB)
sources	
gps	
pc1.parquet	(1.65 MB)
pc2.parquet	(1.23 MB)
pc3.parquet	(1.63 MB)
pc4.parquet	(1.24 MB)
iperf	
mobile_insight	
README.md	(287 bytes)
pc1.zip	(8.99 GB)
pc2.zip	(3.17 GB)
pc3.zip	(5.01 GB)
pc4.zip	(4.01 GB)
pcap	
ping	

Search or jump to...

fraunhoferhhi / BerlinV2X Public

<> Code Issues Pull requests

Code

main

Go to file

- analyze
- data
- docs
- metadata
- mi
- pics
- plots
- preprocess
 - add_pos_in_ref_round.ipynb
 - merge_cellular.ipynb
 - merge_mi_all.ipynb
 - merge_mi_pdsch.ipynb
 - merge_mi_phy.ipynb
 - merge_mi_pusch.ipynb
 - merge_mi_rrc.ipynb
 - merge_sidelink.ipynb

BerlinV2X Dataset

PREPROCESSING

- Excerpt of `cellular_dataframe.parquet`

timestamp	device	PCell_RSSI_max	PCell_Cell_Identity	SCell_RSSI_max	SCell_Cell_Identity	...	ping_ms	datarate	Latitude	Longitude	temperature	Traffic Jam Factor	measured_qos	drive_mode	target_datarate	direction	operator	area
2021-06-22 14:34:10	pc4	-47.118750	29127680.0	-52.660625	29127683.0	...	2238.0	37700000.0	52.514862	13.322625	21.57	2.53887	datarate	platoon	350000000	downlink	1	Avenue
2021-06-22 14:34:10	pc3	-58.760000	51447562.0	-69.550000	51447567.0	...	47.7	403000.0	52.515300	13.323007	21.62	2.53887	delay	platoon	400000	downlink	2	Avenue
2021-06-22 14:34:11	pc3	-61.433125	51447562.0	-75.035625	51447567.0	...	41.9	403000.0	52.515213	13.322935	21.62	2.53887	delay	platoon	400000	downlink	2	Avenue
2021-06-22 14:34:11	pc2	-93.064375	51447562.0	-92.653750	NaN	...	35.9	403000.0	52.514997	13.322730	21.57	2.53887	delay	platoon	400000	downlink	2	Avenue
2021-06-22 14:34:12	pc2	-92.622500	51447562.0	-92.585625	NaN	...	44.9	403000.0	52.514923	13.322672	21.57	2.53887	delay	platoon	400000	downlink	2	Avenue
2021-06-22 14:34:12	pc3	-62.138125	51447562.0	-74.110000	51447567.0	...	40.7	394000.0	52.515130	13.322865	21.62	2.53887	delay	platoon	400000	downlink	2	Avenue
2021-06-22 14:34:13	pc3	-60.440625	51447562.0	-74.069375	51447567.0	...	38.8	413000.0	52.515048	13.322798	21.62	2.53887	delay	platoon	400000	downlink	2	Avenue
...

BerlinV2X Dataset

RESOURCES

- Data on **IEEE Dataport**: <https://ieee-dataport.org/open-access/berlin-v2x>
- Code on **GitHub**: <https://github.com/fraunhoferhhi/BerlinV2X>
- Documentation
 - Readme on **GitHub** and **IEEE Dataport**
 - Hernangómez, R. *et al.* (2023) 'Berlin V2X: A Machine Learning Dataset from Multiple Vehicles and Radio Access Technologies', in *2023 IEEE 97th Vehicular Technology Conference (VTC2023-Spring)*. 2023 IEEE 97th Vehicular Technology Conference (VTC2023-Spring), Florence, Italy. Preprint available at: <https://doi.org/10.48550/arXiv.2212.10343>.

IEEE DataPort™



GitHub



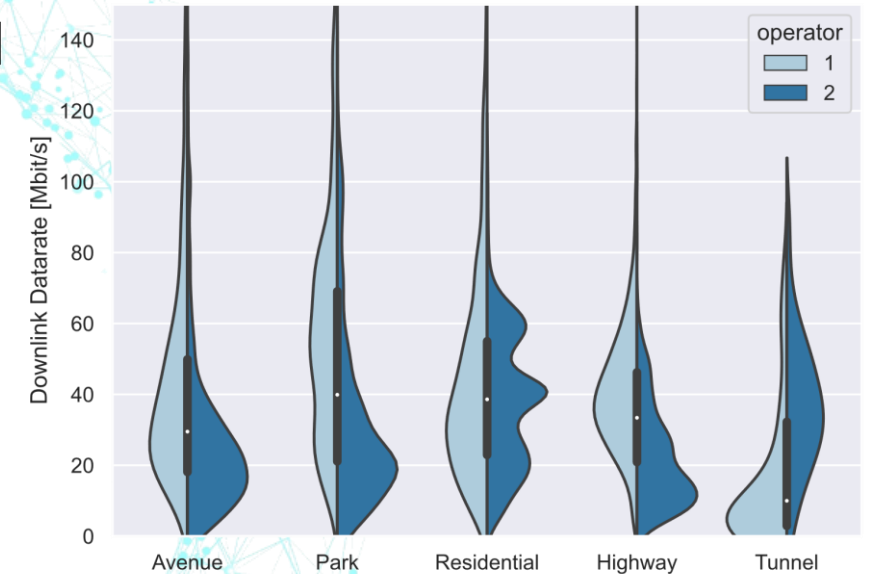
VTC2023-Spring
FLORENCE
Connecting the Mobile World

arXiv

ITU Problem Statement

QoS PREDICTION ACROSS ENVIRONMENTS

- Automotive communication is multi-environmental in essence
 - Underlying data distributions differ across vehicles, operators, areas
 - i.i.d. assumptions are systematically violated
- Train/test data split along such environments
 - Random split is often too indulgent (Palaios *et al.*, 2023)

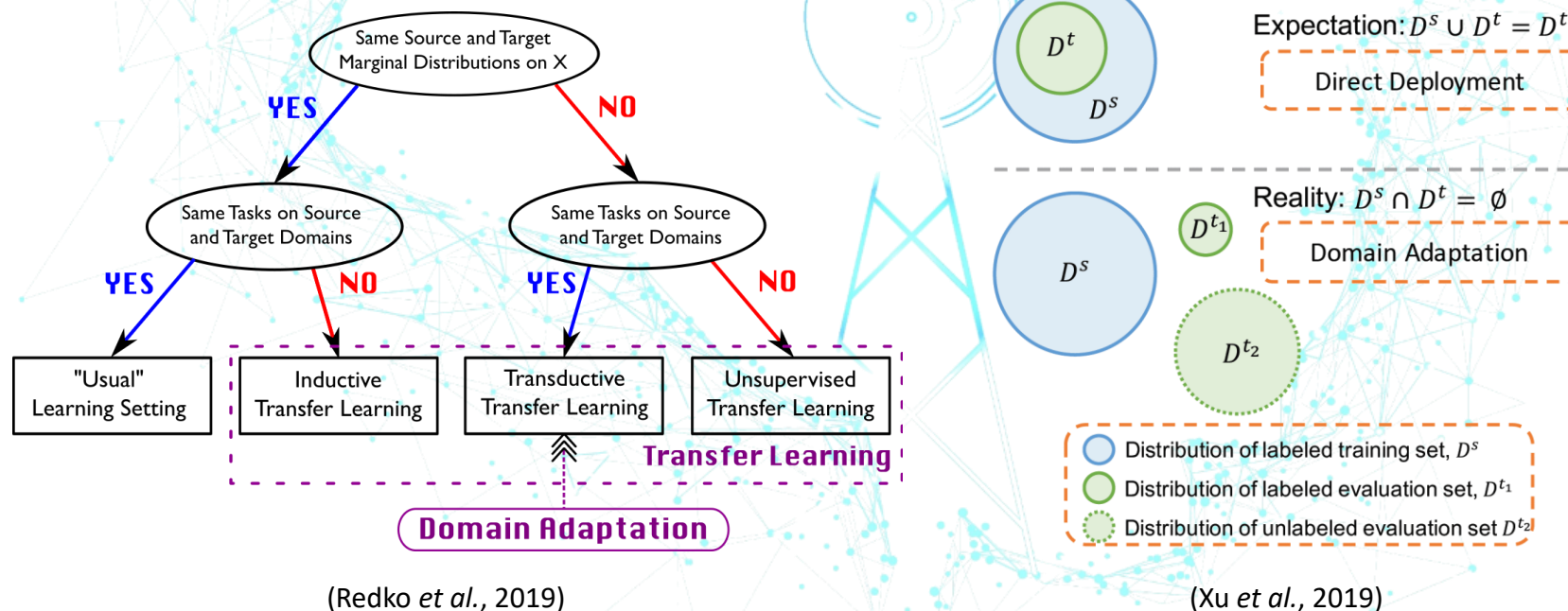


(Hernangómez *et al.*, 2023)

ITU Problem Statement

DOMAIN ADAPTATION

- Discussed as domain adaptation / transfer learning / concept drift
- Automotive QoS prediction hardly fits the „usual“ learning setting



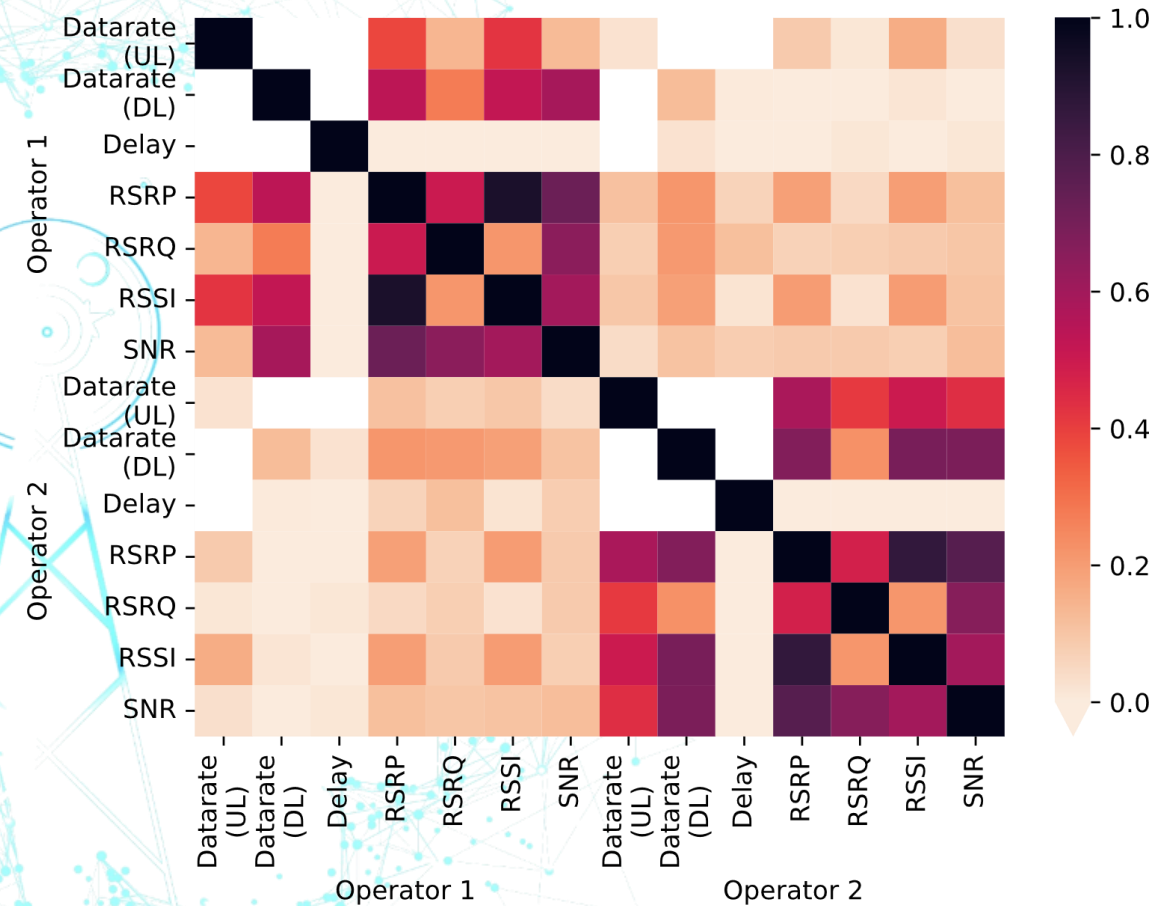
(Redko et al., 2019)

(Xu et al., 2019)

ITU Problem Statement

OPEN QUESTIONS

- Which QoS parameter?
 - Focus on datarate (uplink or downlink)
- Which input features?
- How to split train/test data?
 - Areas
 - Vehicles
 - Operators
 - Uplink/downlink
- Which algorithm?
 - Base ML regressor or Neural Network
 - Fine-tuning techniques: DA/TL



ITU Problem Statement

REFERENCE EXAMPLE

- Available on [GitHub](#)
 - Focused on downlink datarate as QoS with train/test split along operators
- Improvements
 - You can define a different pQoS problem
 - Feature selection is up to you
 - The less features, the better
 - You may choose a different train/test split (along discussed environments)
 - You are free to choose your ML/DL algorithms
 - You may apply (unsupervised) domain adaptation and transfer learning techniques
 - You can create your own preprocessing pipeline
 - Upsample GPS and datarate to ms-range instead of LTE downsampling
 - Include other features from the LTE stack

ITU Problem Statement

EVALUATION

- Weighted score
 1. Coefficient of determination R^2
 - Available for [Scikit-learn](#), [PyTorch](#), and [Tensorflow](#):

$$R^2(y, \hat{y}) = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2}$$

2. Number of used features:

$$R_{features} = (useful\ features - used\ features) = 84 - used\ features$$

3. Problem setup (qualitative)
 - The choices on predicted QoS, train and test datasets, etc.

- Submission on [ITU's challenge platform](#)

- Scores.csv
- Code deliverable
- Brief report (2-5 pages)
- (Optional) model and weights

team_id	predicted_qos	train_set	test_set	r2_score	used_features
REFERENCE	downlink datarate	operator 1	operator 2	0.806633	39

ITU Problem Statement

TIMELINE

- Competition Phase

- Registration from **23 May 2023** to **31 August 2023**
- Submission deadline **8 September 2023**
- Evaluation of solutions: **31 October 2023**

- Evaluation Phase

- **November 2023** – Judges Panel evaluates the best solutions from the Competition Phase
- **28 – 30 November 2023** – Best solutions pitch in a 3-day event to determine the finalists
- **13 December 2023** – Grand Challenge Finale

- Winning Prize: 1000 CHF + certificates

References

- Hernangómez, R. *et al.* (2023) 'Berlin V2X: A Machine Learning Dataset from Multiple Vehicles and Radio Access Technologies', in *2023 IEEE 97th Vehicular Technology Conference (VTC2023-Spring)*. *2023 IEEE 97th Vehicular Technology Conference (VTC2023-Spring)*, Florence, Italy. Available at: <https://doi.org/10.48550/arXiv.2212.10343>.
- Külzer, D.F. *et al.* (2021) 'AI4Mobile: Use Cases and Challenges of AI-based QoS Prediction for High-Mobility Scenarios', in *2021 IEEE 93rd Vehicular Technology Conference (VTC2021-Spring)*. *2021 IEEE 93rd Vehicular Technology Conference (VTC2021-Spring)*, Helsinki, Finland, pp. 1–7. Available at: <https://doi.org/10.1109/VTC2021-Spring51267.2021.9449059>.
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- Palaivos, A. *et al.* (2023) 'The Story of QoS Prediction in Vehicular Communication: From Radio Environment Statistics to Network-Access Throughput Prediction'. arXiv. Available at: <https://doi.org/10.48550/arXiv.2302.11966>.
- Redko, I. *et al.* (2019) *Advances in Domain Adaptation Theory*. Elsevier. Available at: <https://doi.org/10.1016/C2016-0-05108-2>.
- Xu, X. *et al.* (2019) 'd-SNE: Domain Adaptation Using Stochastic Neighborhood Embedding', in. *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, pp. 2497–2506. Available at: http://openaccess.thecvf.com/content_CVPR_2019/html/Xu_d-SNE_Domain_Adaptation_Using_Stochastic_Neighborhood_Embedding_CVPR_2019_paper.html (Accessed: 4 May 2020).

THANK YOU.

Questions?



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