ITUEvents

ITU-ML5G-PS-036: Radio Link Failure Prediction Challenge

Salih Ergüt, Turkcell 22 July 2020

ITU AI/ML in 5G Challenge

Applying machine learning in communication networks

ai5gchallenge@itu.int



Register here Join us on <u>Slack</u>

Turkcell Group Snapshot

Network Provider



Experience Provider

Turkey



- THE DIGITAL OPERATOR
- ~50 MILLION TURKCELL GROUP SUBSCRIBERS in 5 COUNTRIES
- ~170 MILLION DIGITAL SERVICES DOWNLOADS GLOBALLY
- ~86 MILLION DIGITAL SUBSCRIBERS GLOBALLY
- LISTED ON NYSE & BORSA ISTANBUL WITH A \$5 BILLION MARKET CAP



SDO Activities







The Challenge

Background



A Survey on Network Resiliency Methodologies against Weather-based Disruptions

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Fig. 1. Antennas mounted on a lattice tower (on the left), antennas mounted on a monopole (on the right).





A 2009 paper on the effects of weather and foliage

IEEE TRANSACTIONS ON VEHICULAR TECHNOLOGY, VOL. 58, NO. 8, OCTOBER 2009

4023

The Effects of Tropical Weather on Radio-Wave Propagation Over Foliage Channel

Yu Song Meng, Student Member, IEEE, Yee Hui Lee, Member, IEEE, and Boon Chong Ng, Senior Member, IEEE



ig. 1. Plantation under measurement.



Fig. 2. Schematic diagram of the measurement setup and generic views of the first Fresnel zones (240 MHz ____; 700 MHz _____).

 Given the region-wise, historical data sets on radio link (RL) performance and weather forecast predict the RL failures to assess risks

Problem





Dataset

- Training data includes pre-processed and anonymised RL KPIs from our networks and time-aligned weather data.
- RL data
 - KPI data includes date/time, frequency band, link length, error and failure statistics, availability ratio, stability score, capacity, modulation (128QAM, 256QAM, 512QAM, ...)
- Weather data
 - Forecast data includes status, temperatures, humidity, wind speed and direction for the following 5 days (Recorded twice a day)
 - Measurement data includes temperatures, humidity, wind speed and direction, precipitation and overcast (Recorded hourly)
- Distances
 - A matrix that gives distance for weather stations and RL sites

rl-kpis

type	RL equipment vendor
datetime	Date and timestamp
end-point	link end-point (NEAR/FAR)
mlid	Mini link ID
mw_connection_no	Unique internal connection ID
site_no	Site no
site_id	Site ID
polarization	RL antenna polarization (Vertical/Horizontal)
card_type	RL modem card type
adaptive_modulation	Whether adaptive modulation is available
freq_band	Frequency band
link_length	Distance between two sites (LOS)
severaly_error_second	Count of 1 sec periods with error that covers >=30% of the frame
error_second	Count of 1 sec periods with error
unavail_second	RL unavailable operation duration in seconds
avail_time	RL active time in seconds
bbe	Indicator of performance degredation. Background bit error count.
rxlevmax	RL received power level
scalibility_score	enabled
capacity	RL capacity
modulation	Modulation deployed

rl-sites

site_no	Site no	
site_id	Site ID	
clutter_class	E.g. average-dense-urban, open in urban, sparse tree, etc.	

met-stations

station-no	Weather station no
clutter_class	E.g. average-dense-urban, open in urban, sparse tree, etc.

distances

[RL_xyz	Radio link site
[WS_123	Weather station no

Submissions

- Predictions for RLF for the test data set (in CSV format)
- Trained ML model
- Design documentation and documented code
- Presentation on the approach, solution and results

Evaluation criteria

- Participants must use the provided data set to train a machine learning algorithm
- The output of the ML algorithm should be able to predict the performance obtained in a new network deployment
- The choice of the ML approach is decided by each participant
- A test data set will be provided to evaluate the performance of the proposed algorithms
- The evaluation of the proposed algorithms will be based on the average squared-root error obtained along with all the predictions compared to the actual result in each type of deployment
- The winners will be given prizes (and may be invited to publish the results in an academic publication or present in a conference, etc)



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https://www.itu.int/en/ITU-T/AI/challenge/2020/Pages/Turkcell.aspx



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