

# Towards a Weather-Based Prediction Model For Starlink Throughput

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Image: https://satellitemap.space/

### **General Motivation of Prediction Models**

Potential link prediction applications

- Schedulers for edge/cloud computing
- Proactive approaches (instead of reactive) for link reconfiguration

High interest in general performance (measurements, simulations)

Similar work exists for mobile network performance

# Starlink: What's that?

- Rapid expanding collection of Low-Earth-Orbit (LEO) Satellites
- High bandwidth, low latency
  Inter-Satellite Routing available
- Worldwide consumer level access to the internet

#### Starlink operates in low orbit

Low-Earth orbit satellites can link to Earth faster, but more are needed to provide coverage



Starlink dish

# Why weather-based prediction specifically on Starlink?

Satellite links are generally susceptible to environmental influences

- Space-Events
- Weather-Effects
- Local Environment

Similar work suggest severe impact on performance

- Rain
- Sun flares



# **Overview WetLinks Dataset**

About 140k datapoints covering network performance

Covers about six months

Supports Deutscher Wetterdienst (DWD) data integration

Represents largest and most complete Starlink dataset to date

Also includes...

- Round-Trip-Time (RTT)
- Packet Loss Rate (PLR)
- Traceroutes
- Weather Data (DWD, on-site)

#### root@vm952:~/collector# ls export

cloud_pictures	cws_sensors.csv	froggit.txt	net_iperf.csv	net_ping.txt	starlink.csv
cws_clouds.csv	cws_sensors.txt	metadata.csv	net_iperf.txt	net_traceroute.csv	starlink.txt
cws clouds.txt	froqqit.csv	metadata.txt	net ping.csv	net traceroute.txt	

Pv4 UDP throughput

of down- and uplink

using iperf3

IPv4 round trip time

and packet loss

using ping

Every six minutes:

IPv4 routing

using MTR



#### Uni Twente



Uni Osnabrück

# WetLinks in our context

Which sides were used?

- University of Osnabrück
- DWD
- Cloud-Pictures on-site

Time coverage? 4 months.

Utilized features from the dataset

- Environment data (temperature, wind, pressure)
- Cloud information and rain (extended by DWD)
- Image statistics (color channel details)



DWD Data



Custom Weather Station with Sky Camera

6

# Methodology: Pre-Processing and Models

Pre-processing of dataset

- Started with >500k datapoints in different locations
- Completed with about 1.1k datapoints in one dataset

Models (primarily decision trees)

- Random Forest
- Gradient Boosting
- AdaBoost
- K-nearest-Neighbor
- MTR-Regressor
- (Dummy Regressor)



# Methodology: Evaluation and Tuning

Metrics

- **R<sup>2</sup>** "How well does it fit?"
- Mean Absolute Error (MAE) "Mean average absolute error?"
- Root Mean Squared Error (RMSE) "Square root of errors squared?"

Groups and Training-Subsets

Fine-Tuning via Grid Search

- Max-Features
- Min-/Max- Samples per split
- Estimators

## Prediction Results: MAE, RSME and R<sup>2</sup>





Download





#### **Prediction Results: Feature Importance**



10

5

-15 -10

-5

0

SHAP value (impact on model output)

### Results in context of other works

Mobile Network Prediction: What did they do?

- Our R<sup>2</sup> is similar around 0.5<sup>1</sup> for ensemble methods
- They also utilized Support-Vector-Machines and Gaussian-Process-Regression, which yielded in better scores (around 0.65)

What can we tell about our result quality?

- Achieved similar performance with comparable models
- More data is needed (requiring a bigger dataset)
- Other models should be explored

<sup>1</sup>This is comparing their prediction, based on signal strength (db), with our throughput prediction (MB/s)

# Conclusion

Able to reproduce similar prediction performance on download throughput

- Rain is the most important feature (as expected)
- Cloudiness seems to be less important, but image statistics suggest otherwise

Upload throughput does not seem to be impacted by weather-conditions

More complete dataset will prove useful

• we saw 20% improvement by adding just one more month (from 3 months)

Neural Networks did not converge on our data

### Future Work

Reduced amount of models were used due to quick turnarounds needed

• Reference work suggest better predictions with highly increased training resources

Further parameter tuning is recommended (due to more accessible data)

More in depths Dataset-/Prediction-Model demo:

At IFIP Networking 2024 in Greece

...and here 😉

