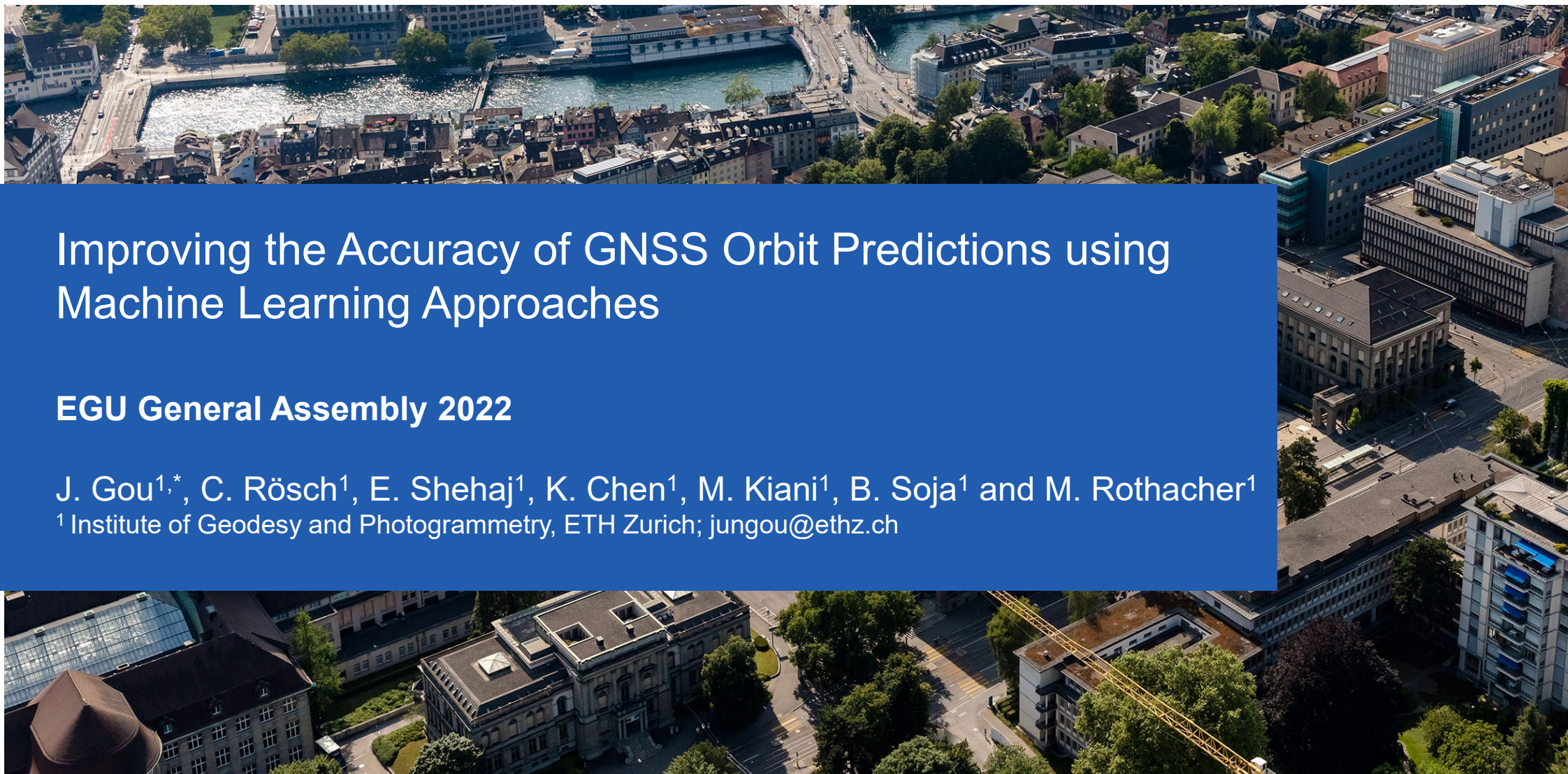


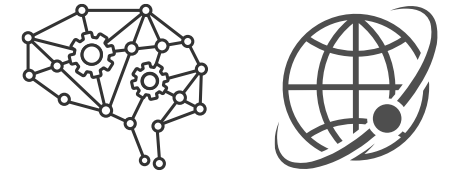
Improving the Accuracy of GNSS Orbit Predictions using Machine Learning Approaches

EGU General Assembly 2022

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- **Motivation**

- Number of space objects in earth's orbit is steadily increasing
- Accurate orbit prediction of space objects is of crucial importance
- Prediction errors of physics-based orbital propagator accumulate with time

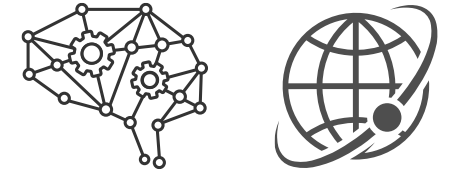
- **Method**

- Predict errors of ultra-rapid orbit predictions based on the difference between ultra-rapid and final orbits
- Machine Learning (ML) and Deep Learning (DL) algorithms are applied to the orbit errors
- Investigate additional features, such as solar activity

- **Application**

- Kinematic Precise Point Positioning (PPP) of station coordinates using improved satellite orbit predictions

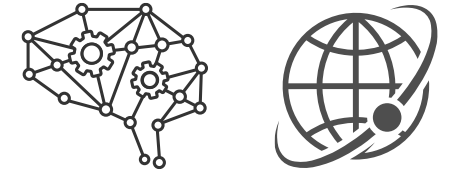
Data used in this study



- The **final** orbit products provided by GFZ^[1] → highest accuracy and latency of around two weeks
- The **ultra-rapid** orbit products provided by GFZ^[2] → published eight times per day
- GPS constellation (32 satellites)
- Data from **January 2019 to April 2021**
- **Train** samples 144'075 (70%), **validation** samples 30'873 (15%), **test** samples 30'873 (15%)



Schematic diagram of the ultra-rapid orbit products provided by GFZ

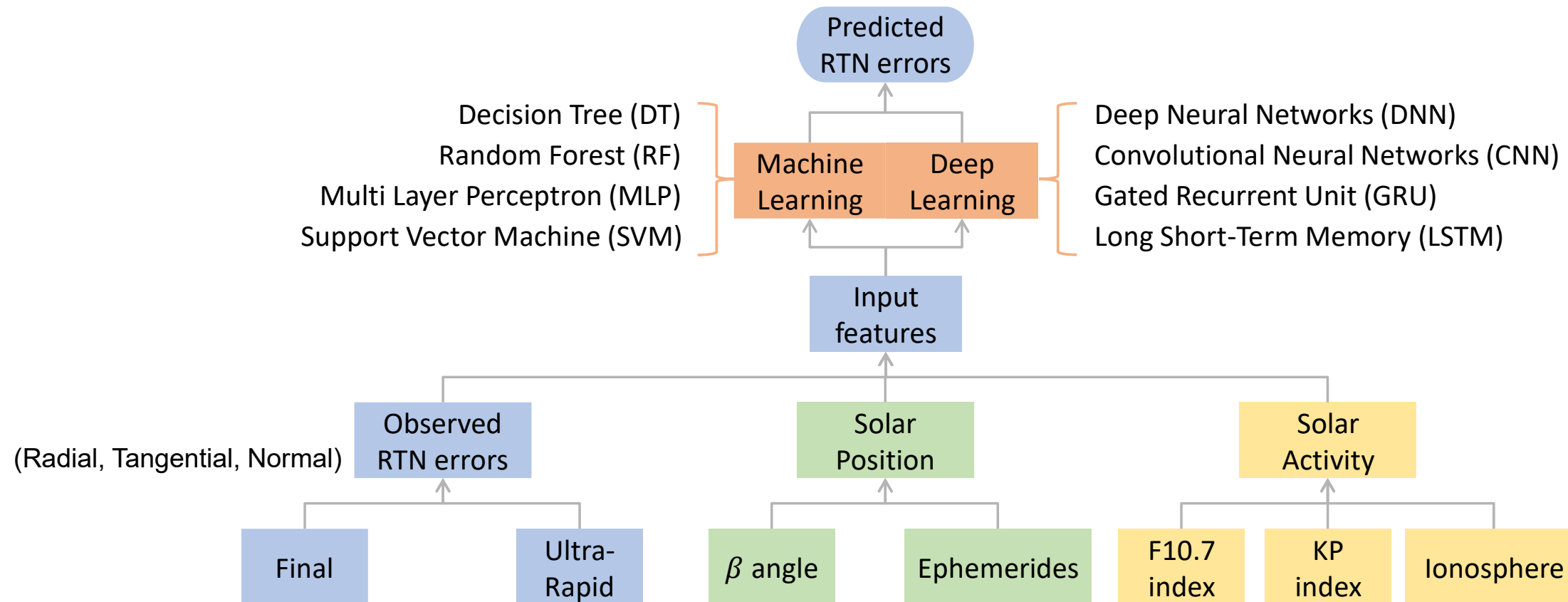
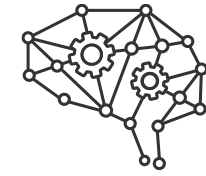


Solar information

- 10.7 cm Solar Radio Flux provided by Space Weather Canada^[1]
- Kp index provided by GFZ^[2]
- $C_{0,0}$ term of the global ionospheric maps provided by CODE^[3]
- Solar positions
 - Beta angle
 - JPL Ephemerides^[4]

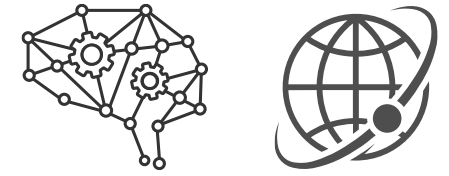
Method Overview

Method Overview

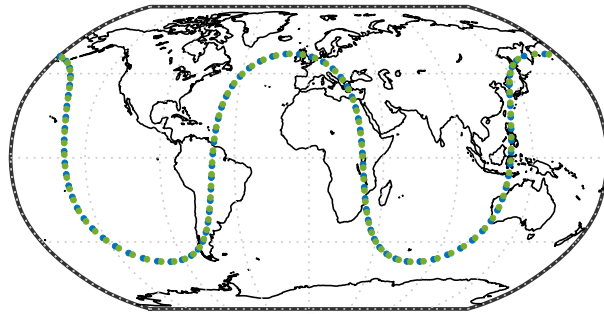


Graphical representation of the pipeline of the study

Obtain orbit errors



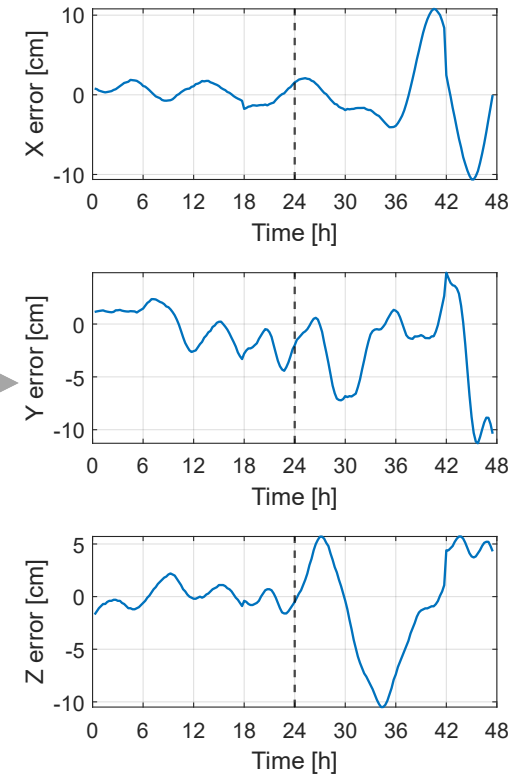
The Final orbit



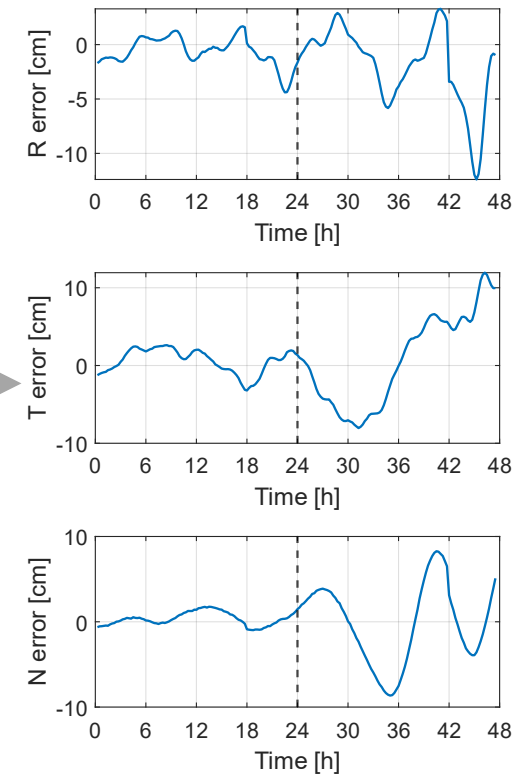
Earth Orientation Parameters^[1]



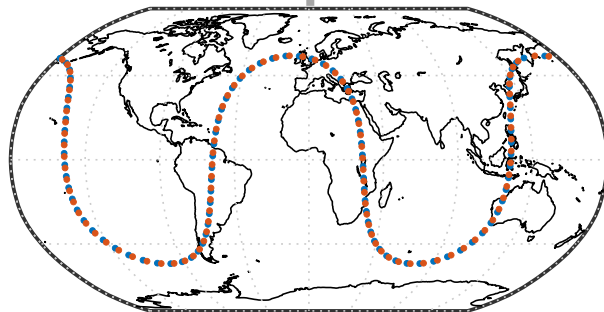
Errors in ECI



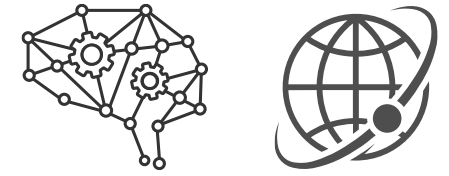
Errors in RTN



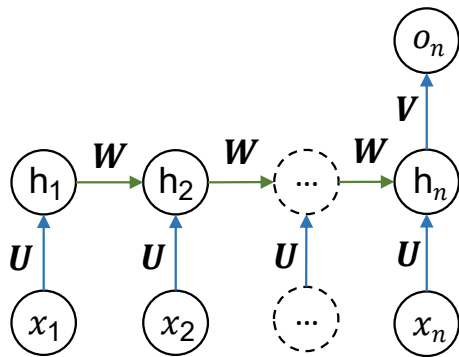
The Ultra-rapid orbit



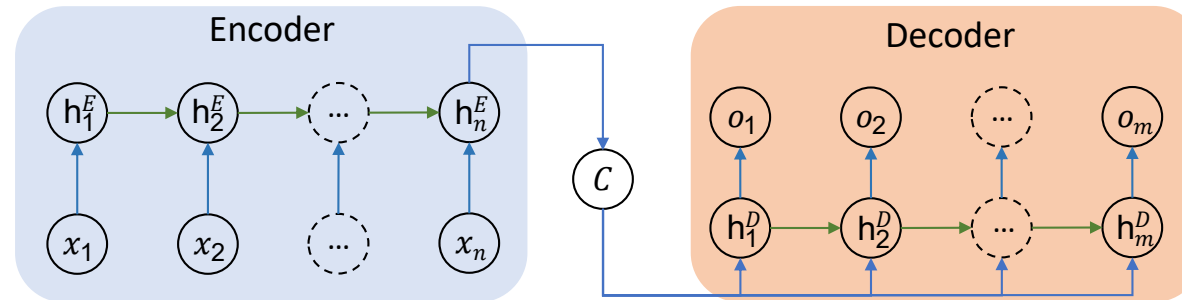
Selected method - LSTM



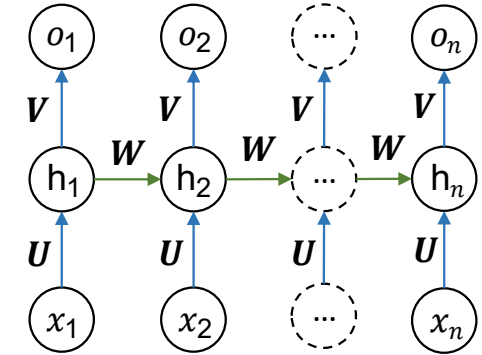
- Long Short-Term Memory (LSTM)^[1]
 - One of the most widely used Recurrent Neural Network (RNN)
 - Proven good performance for prediction of time series
 - Multiple variates
 - Best performance in this study → Show results based on this method



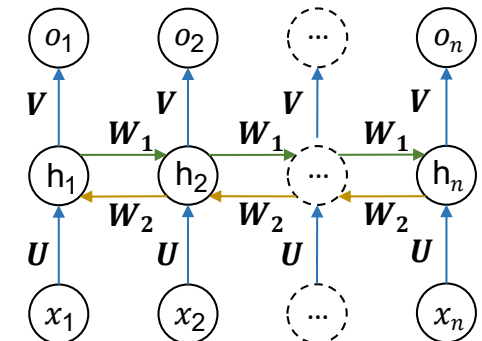
LSTM Sequence-to-One



Encoder-Decoder LSTM



LSTM Sequence-to-Sequence

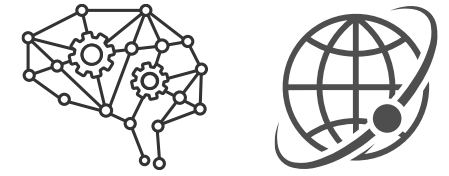


Bidirectional LSTM

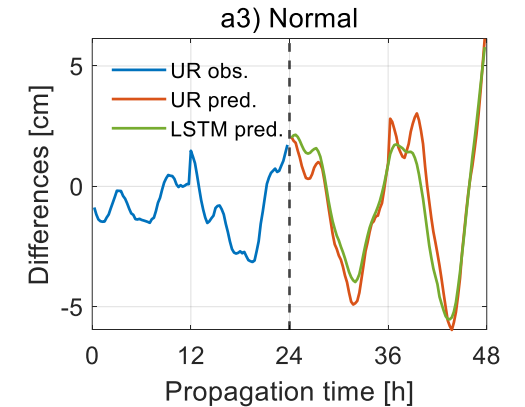
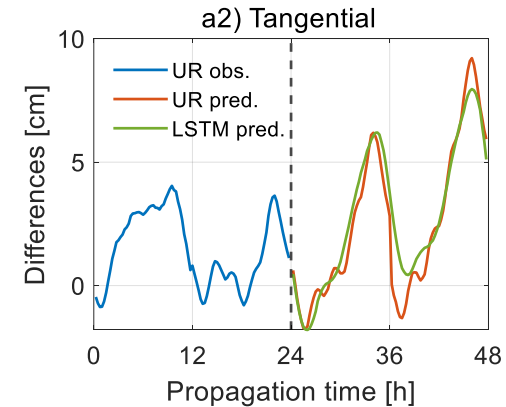
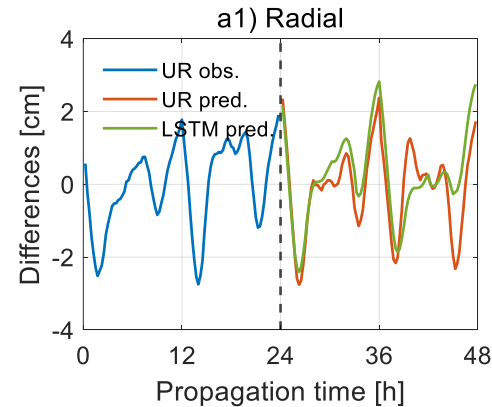
Complexity

Results and Discussion

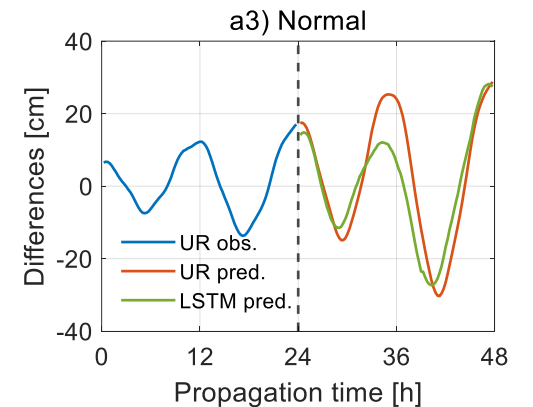
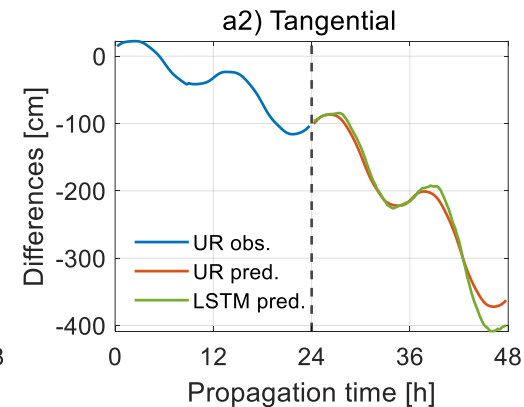
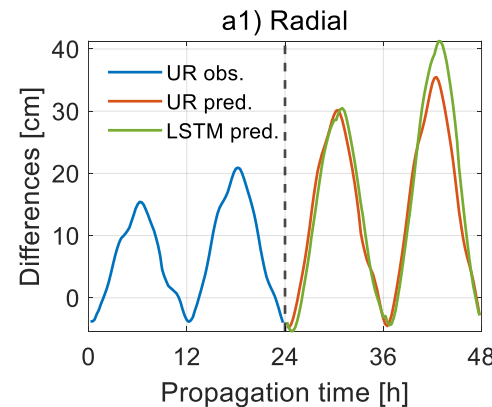
Results – Single samples



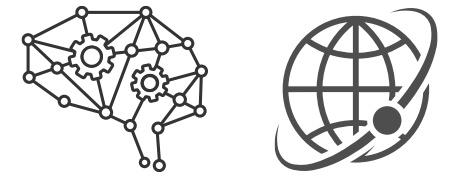
For orbits with low perturbation:



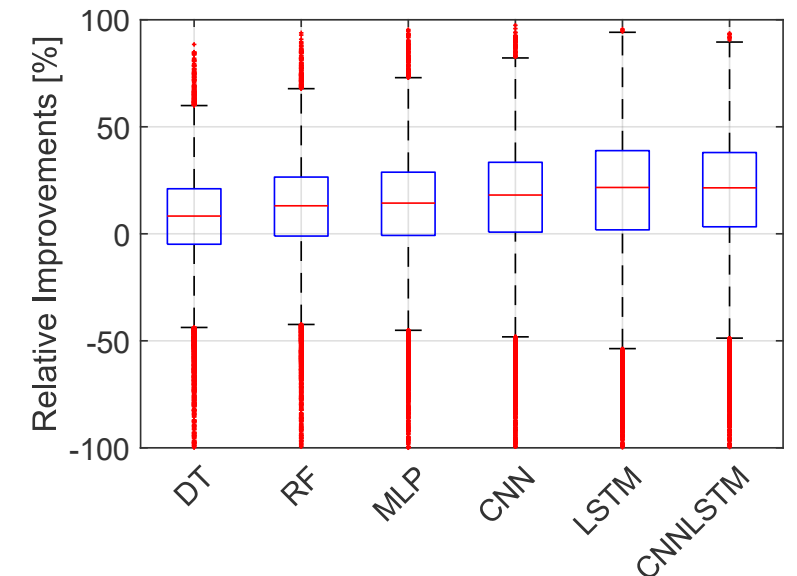
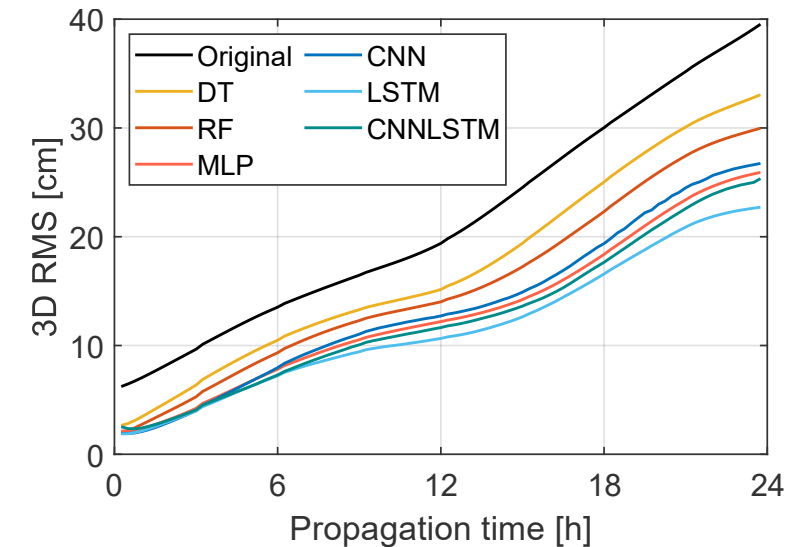
For orbits with high perturbation:

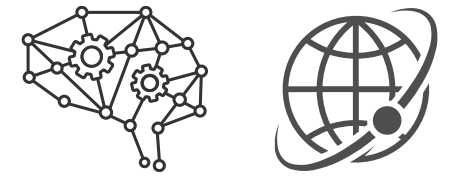


Results – Statistics



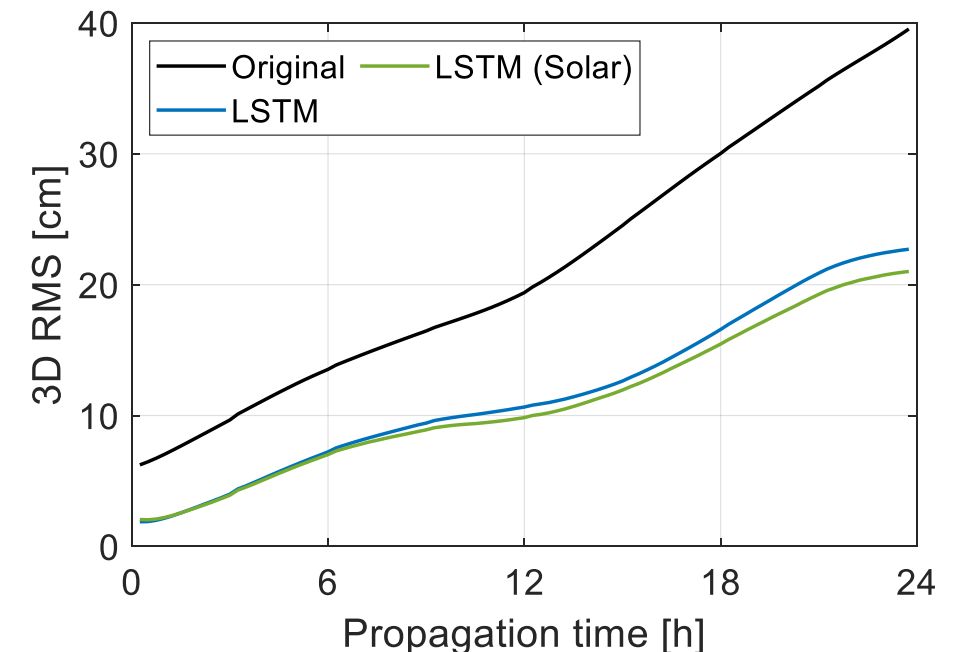
- Performance w.r.t. propagation time:
 - Most of the ML/DL algorithms show significant improvements
 - The improved orbit keeps accuracy better than 1 decimeter until **10h 15min (LSTM)**
 - **Sequential modelling is superior** because of the time-series characteristic of satellite orbits
- Performance on individual samples:
 - Positive Ratio (PR): Percentage of positive improvements
 - Most of the algorithms provide PR **around 75%**
 - LSTM: 77%



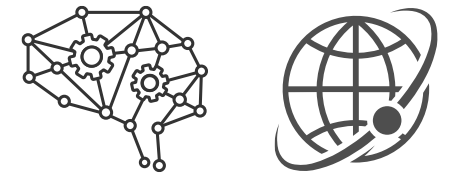


Results – Inclusion of solar information

- The 10.7 cm flux and K_p index do not have significant contribution
 - Low quality and low temporal resolution
- Combination of the solar position and $C_{0,0}$ terms of global ionosphere maps improve the results further
 - Projections of $C_{0,0}$ into the RTN frame using solar position serve as additional features
- Absolute improvements of 3D RMS:
 - Maximal error reduction from 16.7 cm to **18.4 cm**
 - Average error reduction from 9.7 cm to **10.5 cm**
 - More than 10% better by including solar activities
 - PR remains at 77%

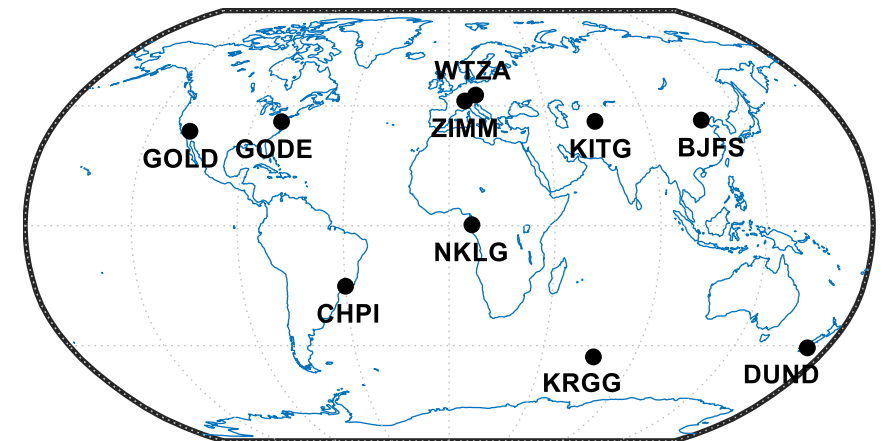
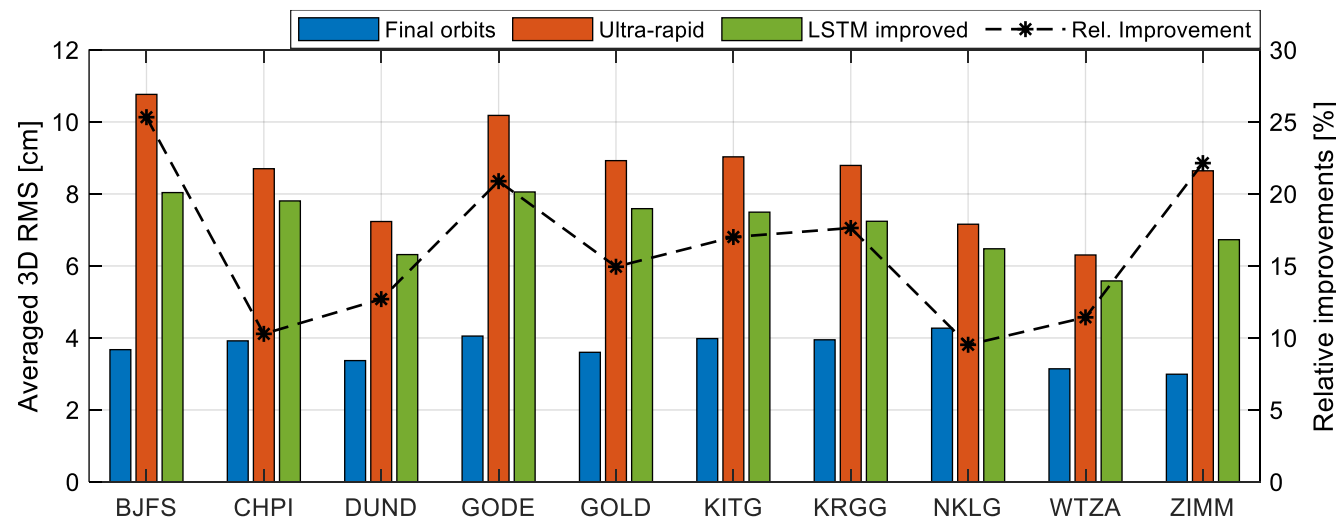


Potential Applications



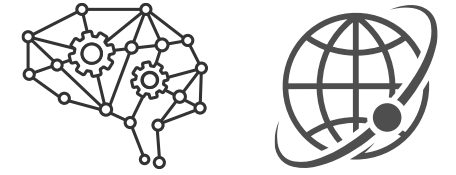
Preliminary results – Validation with Kinematic PPP

- **Experiment settings:**
 - Test on 10 IGS stations globally, 2021. (LSTM model trained on data from 2017 to 2020)
 - IGS combined weekly station positions^[1] serve as reference
 - Kinematic PPP processing using Bernese 5.2^[2]
- **Averaged improvements of 16%** compared to the solutions using ultra-rapid orbits



Conclusion and Outlook

Conclusion and Outlook



- **Conclusion**

- ML and DL algorithms can improve the accuracy of ultra-rapid orbit prediction **up to 73%**
- Improved GPS orbits keeps accuracy better than 1 decimeter up to **10 hours**
- Inclusion of ionospheric parameters and solar position enhances results
- Significant benefits for station coordinates estimation with **average improvement of 16%**

- **Outlook**

- Apply to other GNSS constellations
- Study the benefits for other geodetic applications

Thanks for your attention!

Looking forward to further discussion

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