MAP MAKING WITH A LASER ENGRAVER



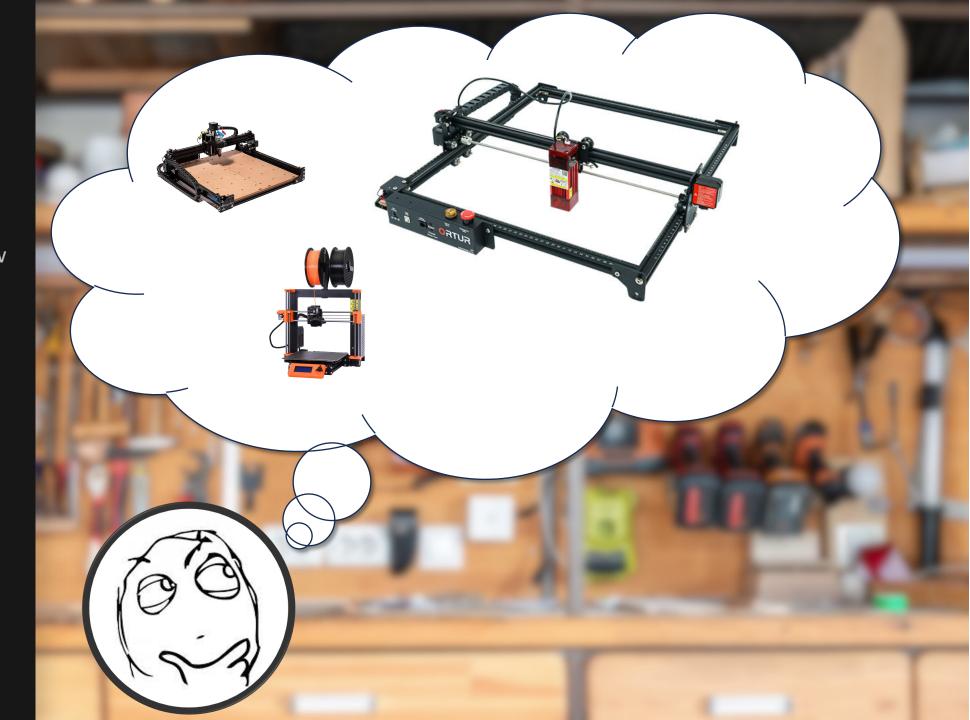
Luca Dall'Acqua
Twitter: @luca_northsouth

WHY A LASER?

Lasers are cool!

As an hobbyist I wanted a new tool...

3D Printer, Laser or CNC?



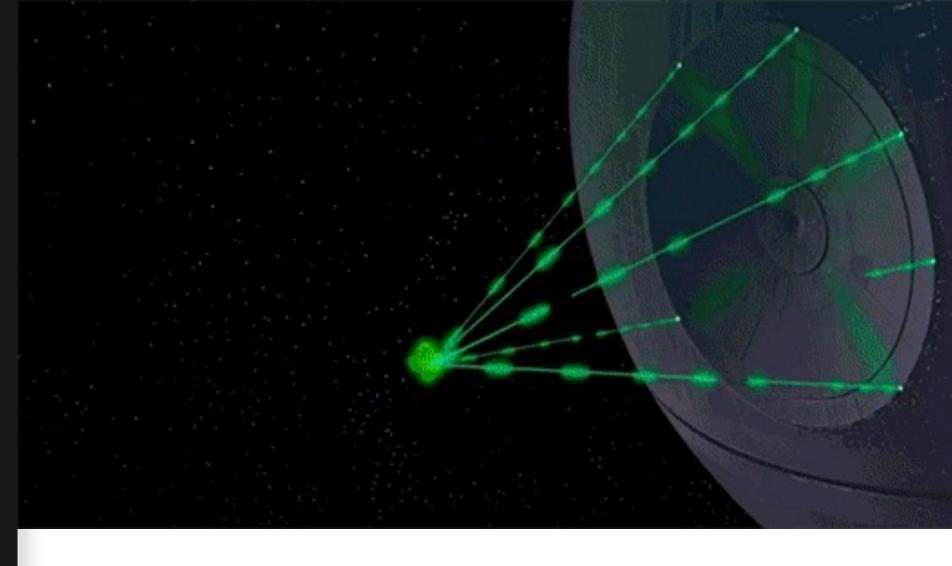
SAFETY FIRST

Lasers are dangerous:

- Fire
- Eye Exposure
- Fumes and smoke

Never leave the laser running unattended Wear protection glasses or use an enclosure

Use ventilation
Know the material you are burning











- Have a great idea (or steal take inspiration)
- Get the data
- Select the geometries
- Prepare layers for cutting or engraving
- Burn (and fail)
- Finish your artwork









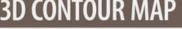






















NORMAN

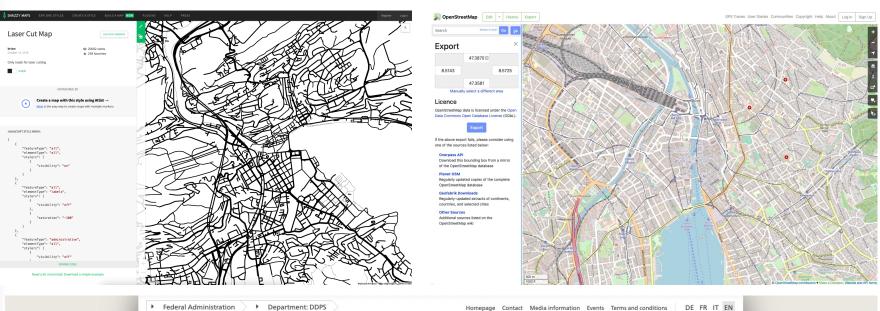








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Homepage > Geodata and applications

Geodata and applications





swisstopo's digital maps are available in raster and vector formats. They form an excellent basis for digital use and further processing.





Images

Images of swisstopo are an instant view of the landscape, taken from the ground, airplanes or from satellites. These survey photographs allow for geometrically high precision processing.





Landscape Models

Landscape models display objects of the landscape in a flexible vector format. They consist of thematic layers (e.g. traffic). Each layer comprises georeferential objects in a point, line or polygon...



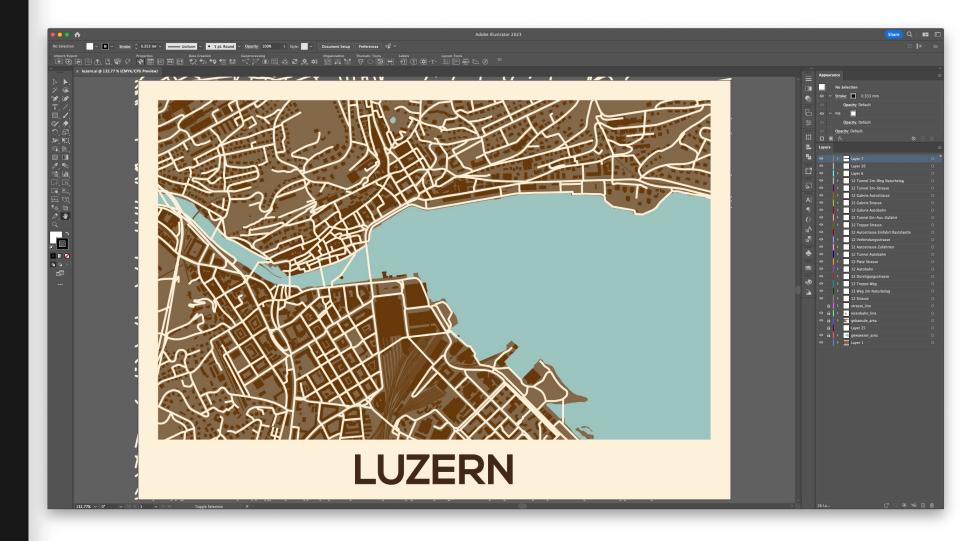


Height Models

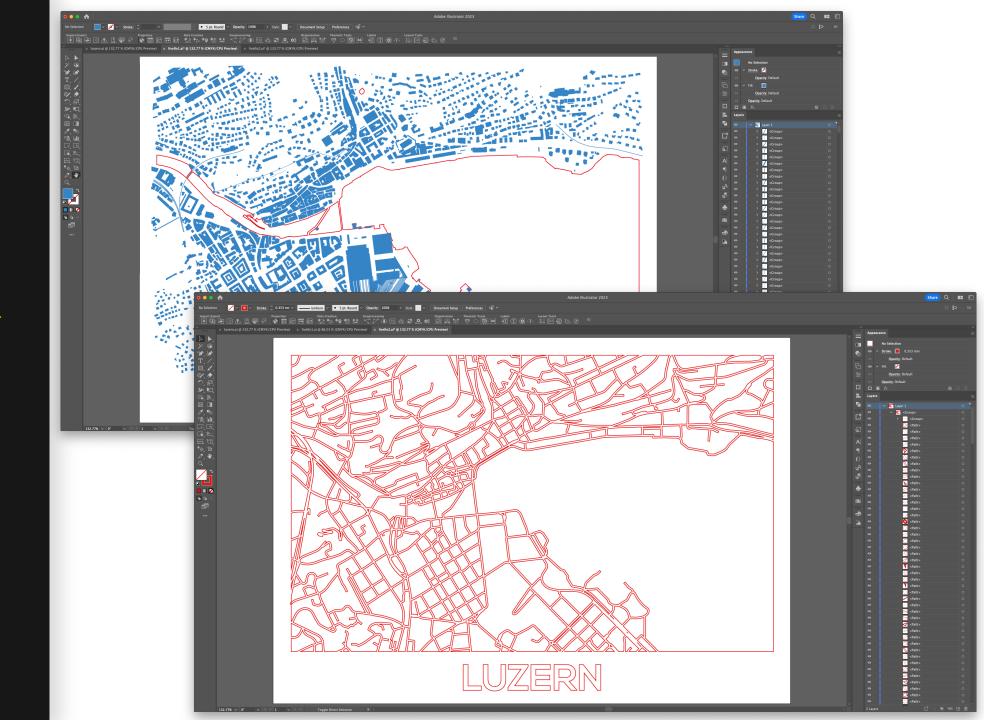
Height models are digital data sets which show the form of the earth's surface in 3D. Each set of X and Y coordinates shows a height (Z). Different products are available for various applications.



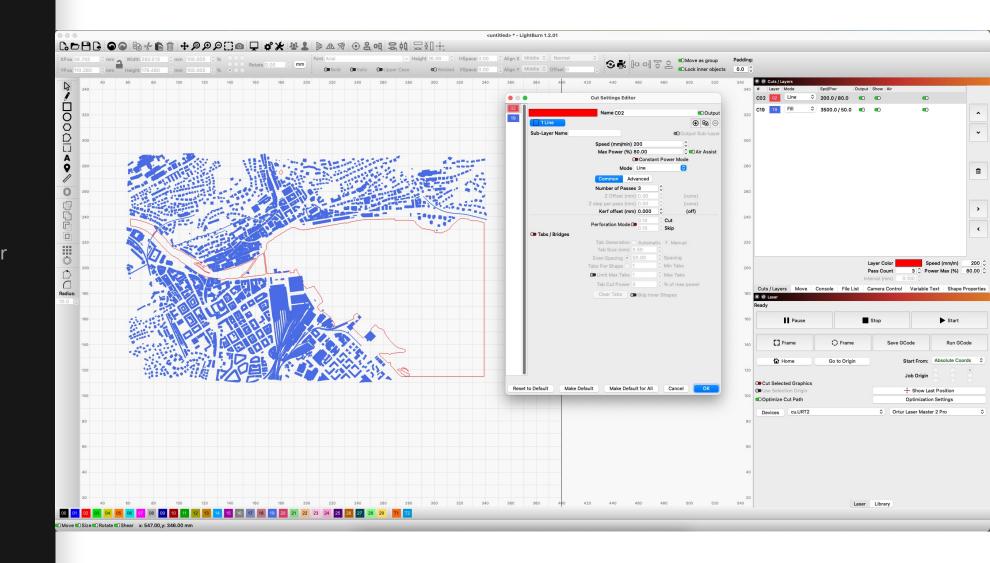
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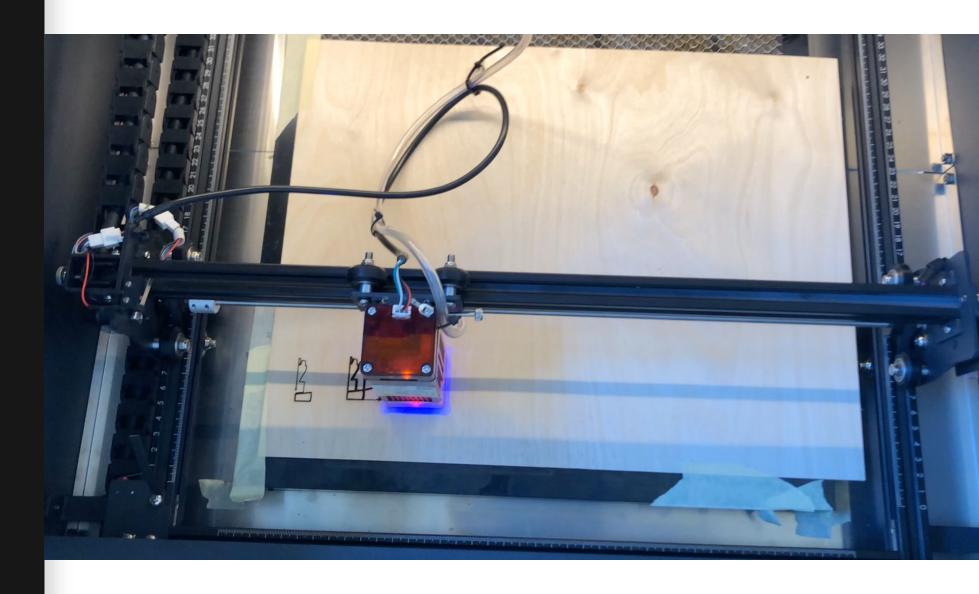
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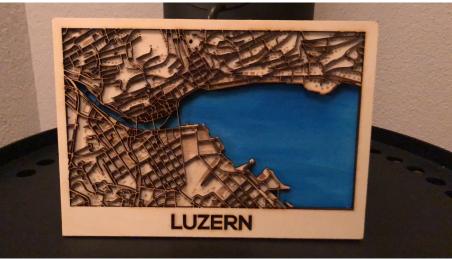


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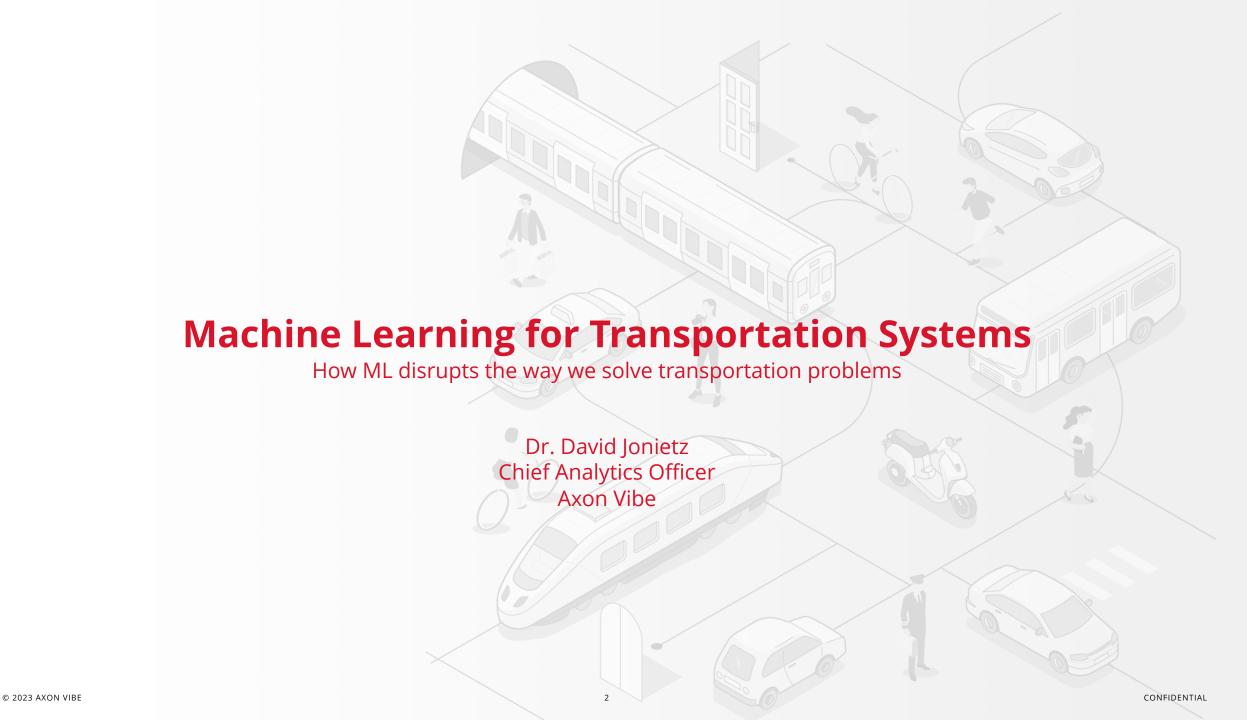






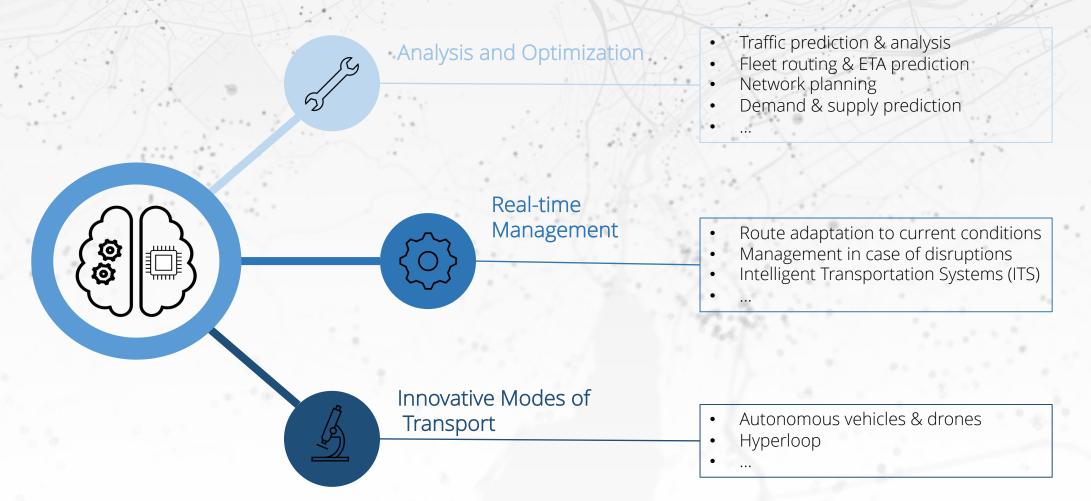






Potentials of ML for Transportation Systems

Where Al/ML can make a difference



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The Traffic4Cast Competition at NeurIPS

- Traffic prediction as a central prerequisite for traffic management
- Traditional methods based on a road network graph
- What if we got rid of the road graph and predicted traffic on pixels of a map/image?

Traffic prediction ≈ Video prediction

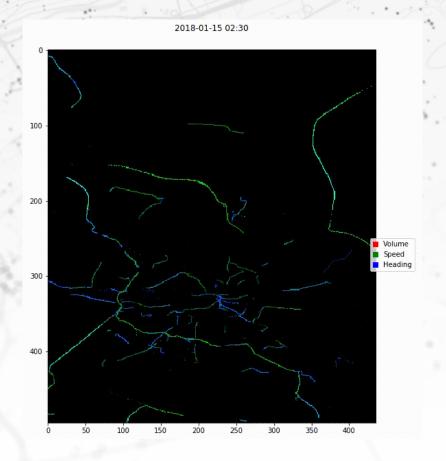


D. Jonietz, M. Kopp, M. Neun, B. Xu, A. Soleymani, "Method, apparatus, and system for end-to-end traffic estimation from minimally processed input data» United States Patent US20230067464A1

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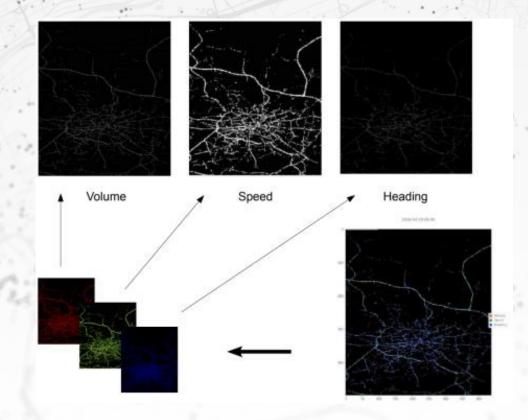


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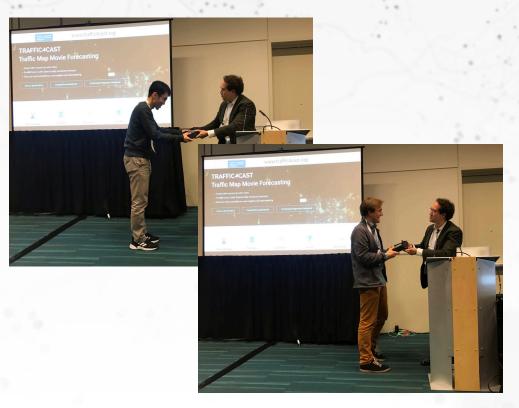
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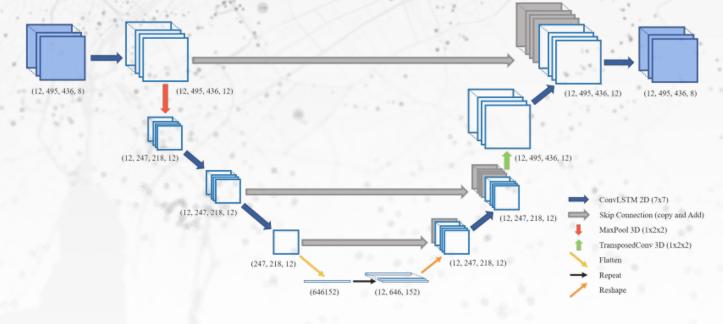


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TRAFFYC.4CAST





STAMP - ETA Prediction with Machine Learning

Spatio-Temporal Aggregated Machine-learning Prediction (STAMP)

- Estimated Times of Arrival as essential information for logistics, fleet management and navigation
- Traditionally based on pre-defined route (e.g., fastest path)
- What if we predicted robust ETAs without routes?

Historical Trips – Routes = ETA

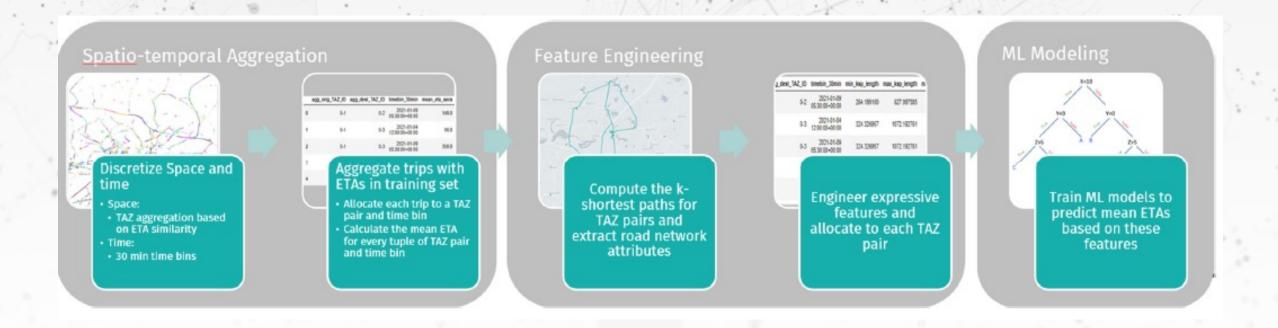


B. Xu, D. Jonietz, R. Gupta, A. Soleymani, K. Malm and R. Köhn, "STAMP: An Approach to ETA Prediction by Spatio-temporal Discretization and Machine Learning," *2022 IEEE 25th International Conference on Intelligent Transportation Systems (ITSC)*, Macau, China, 2022, pp. 893-900, doi: 10.1109/ITSC55140.2022.9922072.

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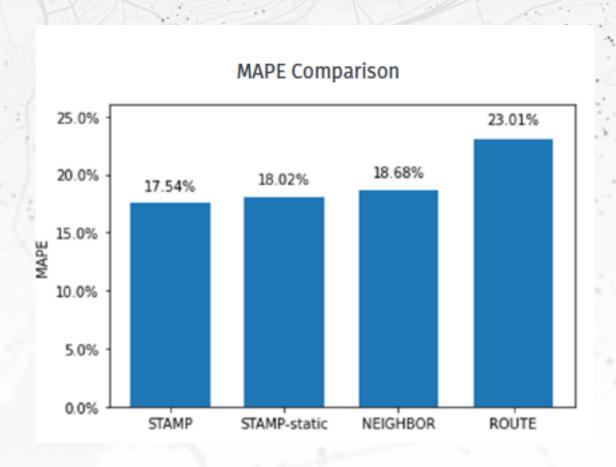
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Summary

- ML not only allows to address new tasks, but can also fundamentally disrupt the way we solve established problems
- ML is data hungry → effort required for compiling large-scale geospatial datasets
- What do we need for true geospatial ML?
- Challenge of root-cause analysis and reliable prediction of rare events