

# MAP MAKING WITH A **LASER** ENGRAVER



Luca Dall'Acqua  
Twitter: @luca\_northsouth

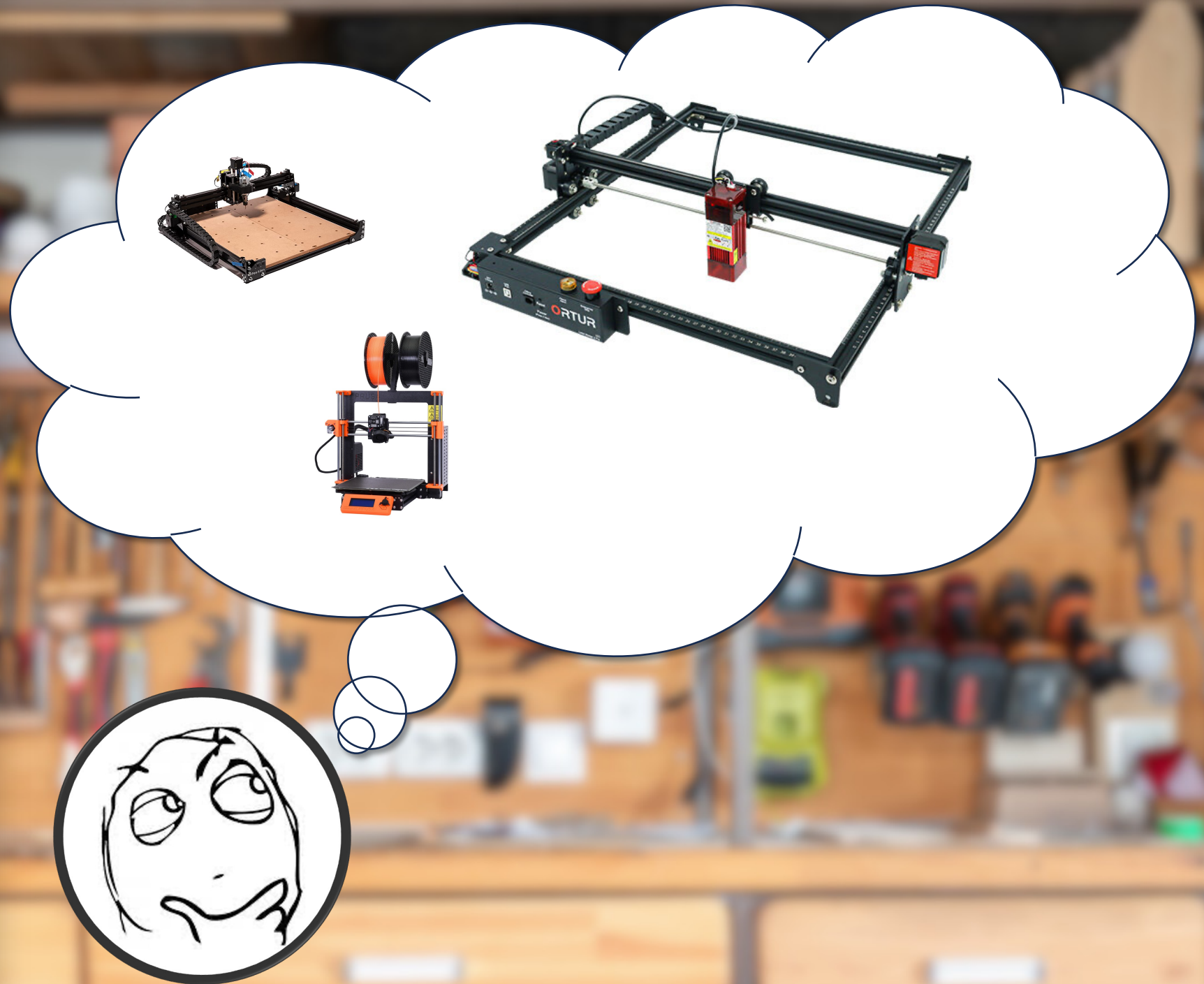


# WHY A LASER?

Lasers are cool!

As an hobbyist I wanted a new tool...

3D Printer, Laser or CNC?

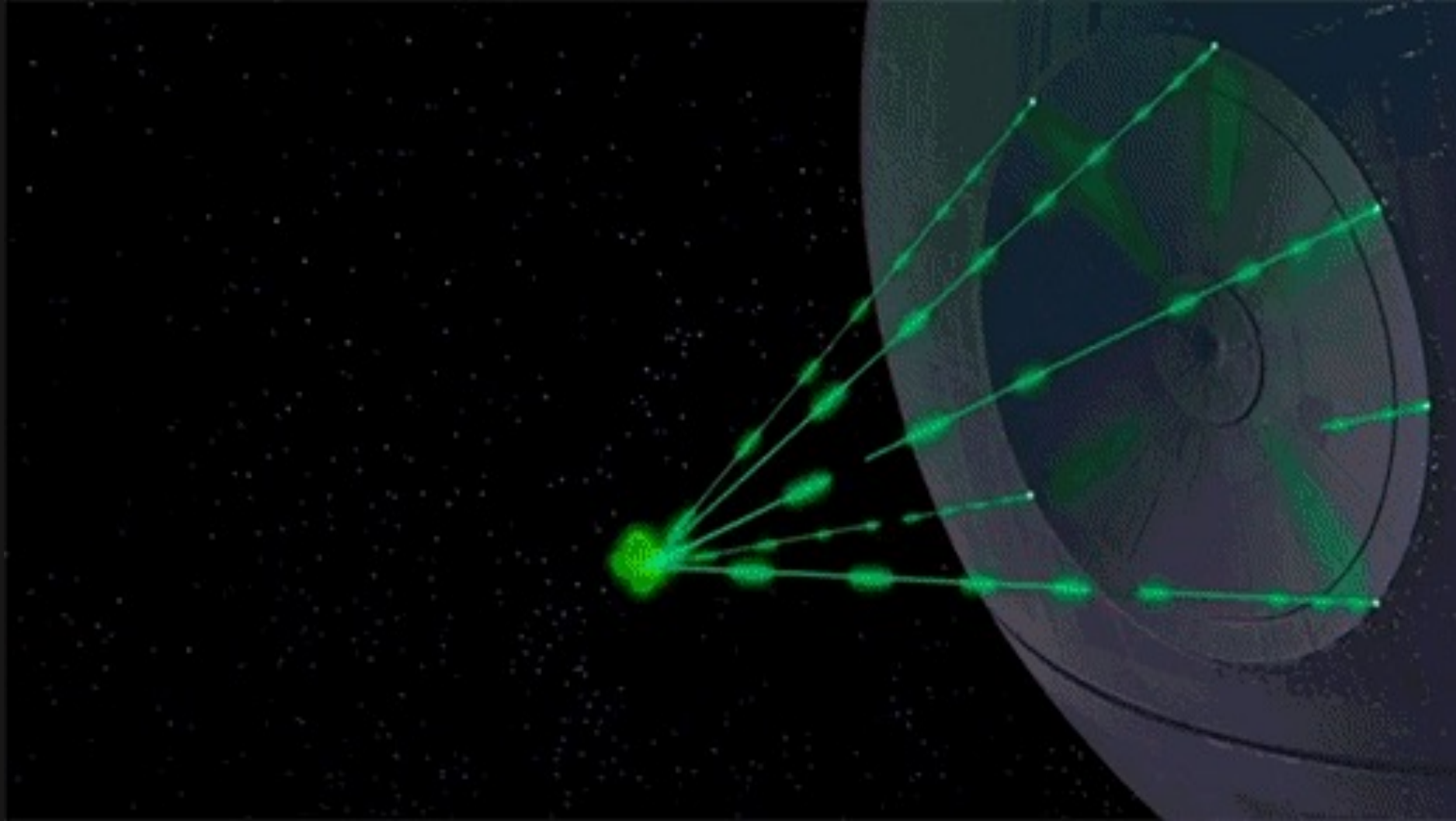


# SAFETY FIRST

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

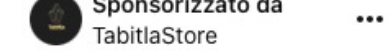
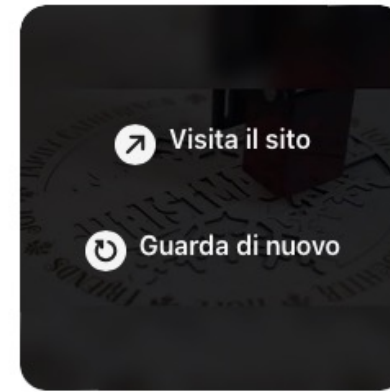
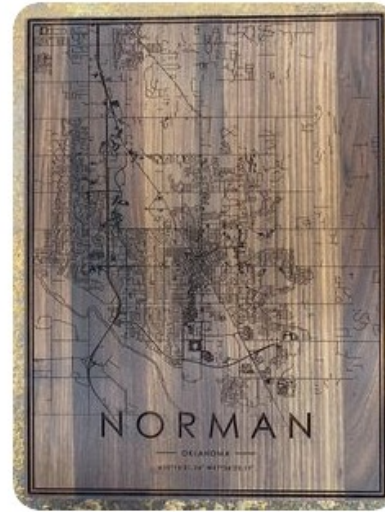
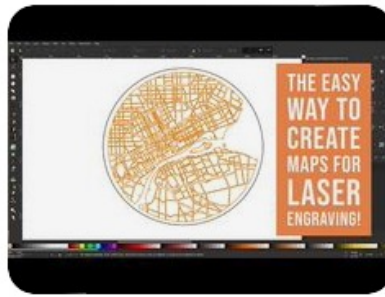


~~PXC~~



# HOW-TO

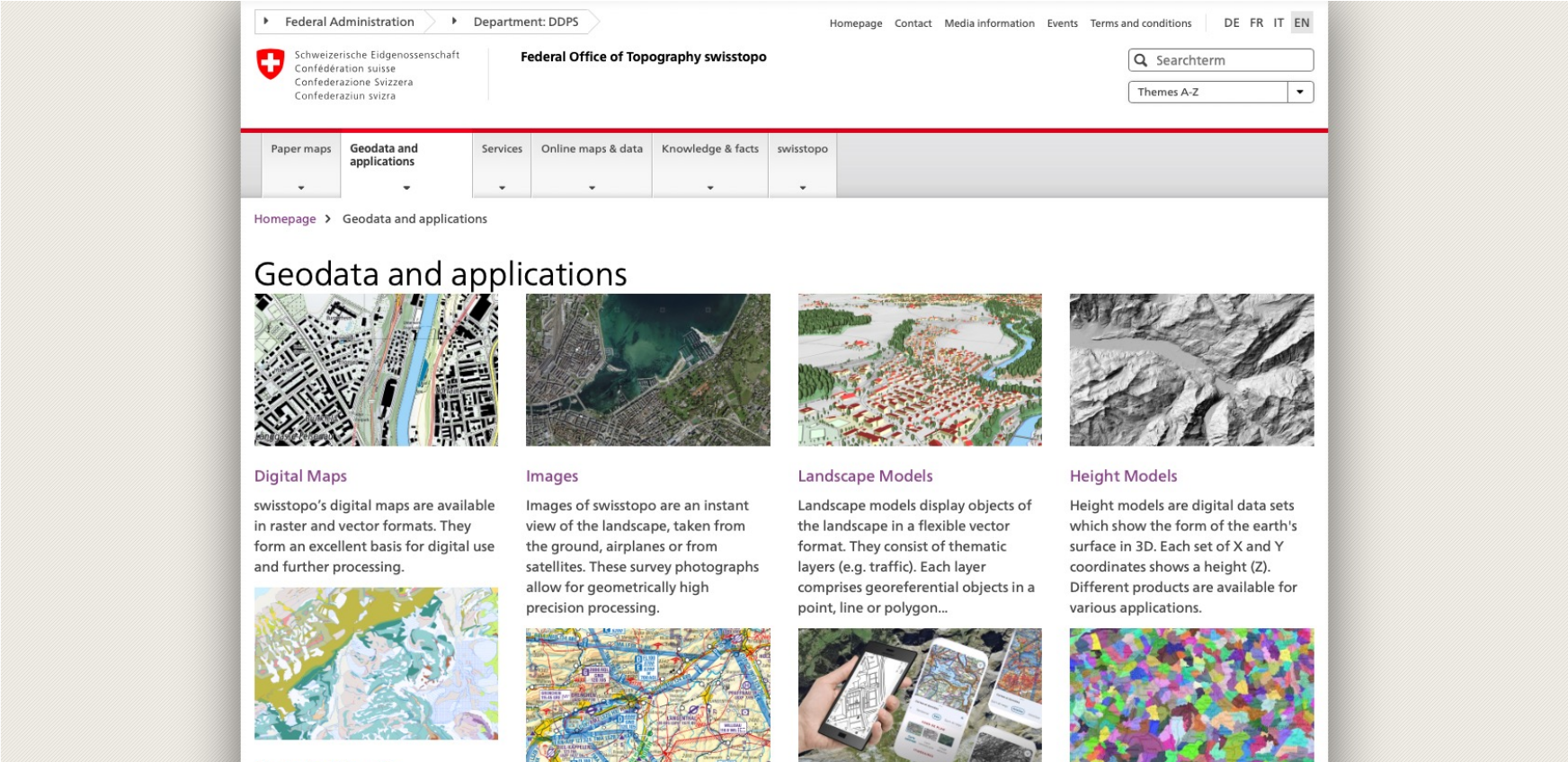
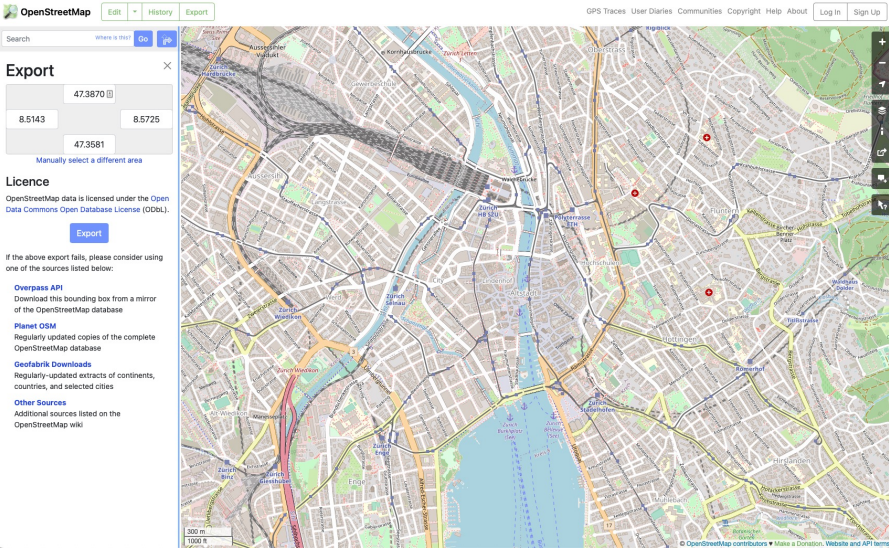
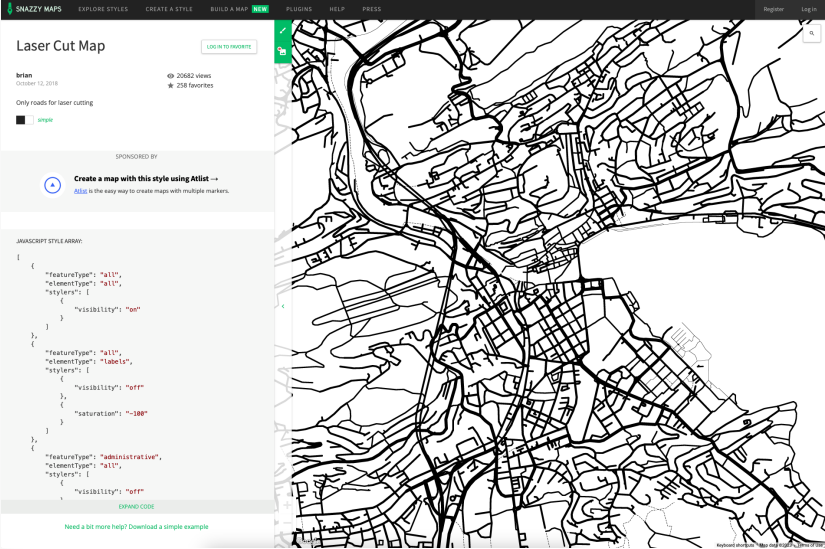
1. Have a great idea  
(or steal take inspiration)
2. Get the data
3. Select the geometries
4. Prepare layers for cutting or engraving
5. Burn (and fail)
6. Finish your artwork





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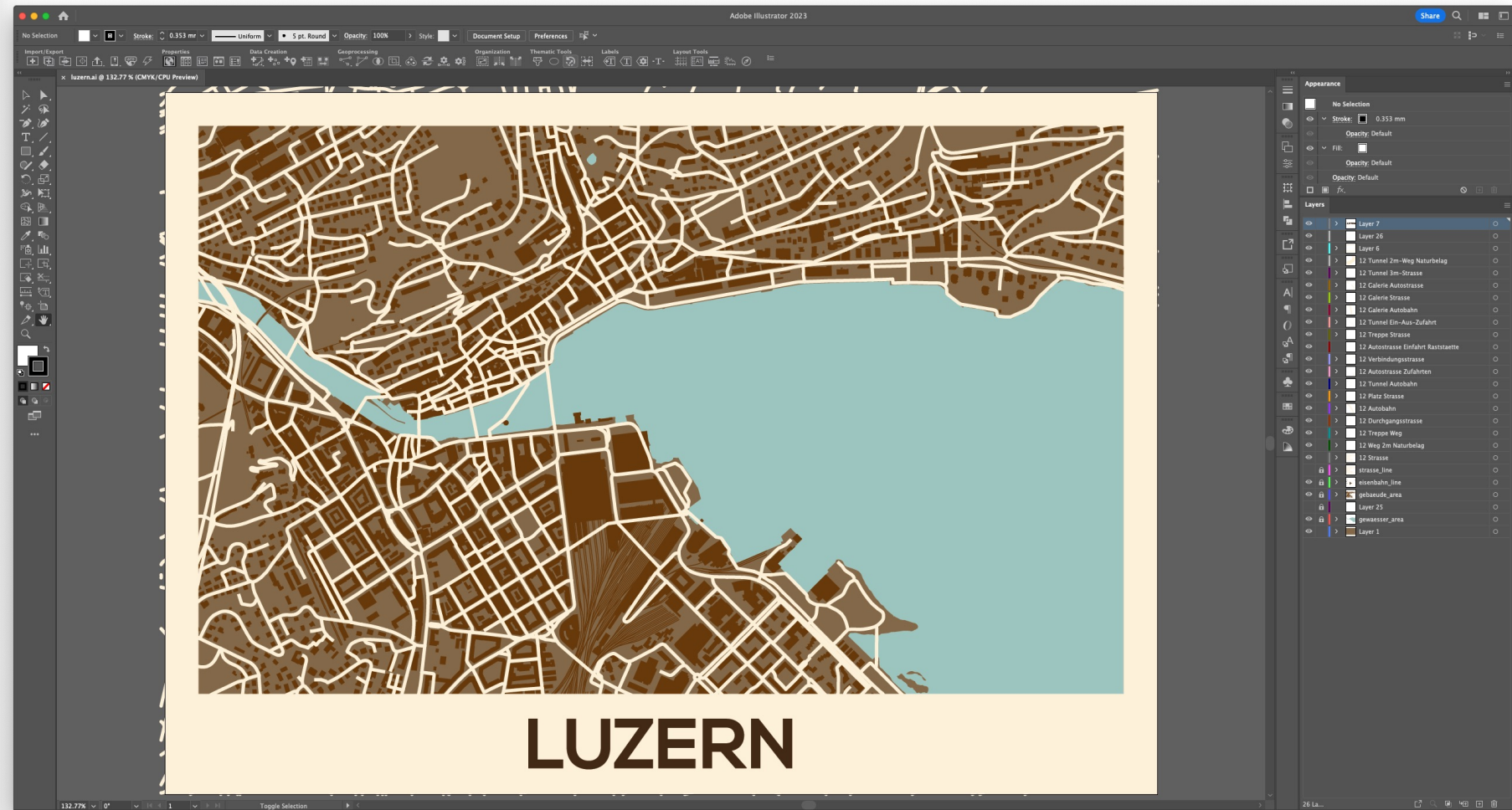
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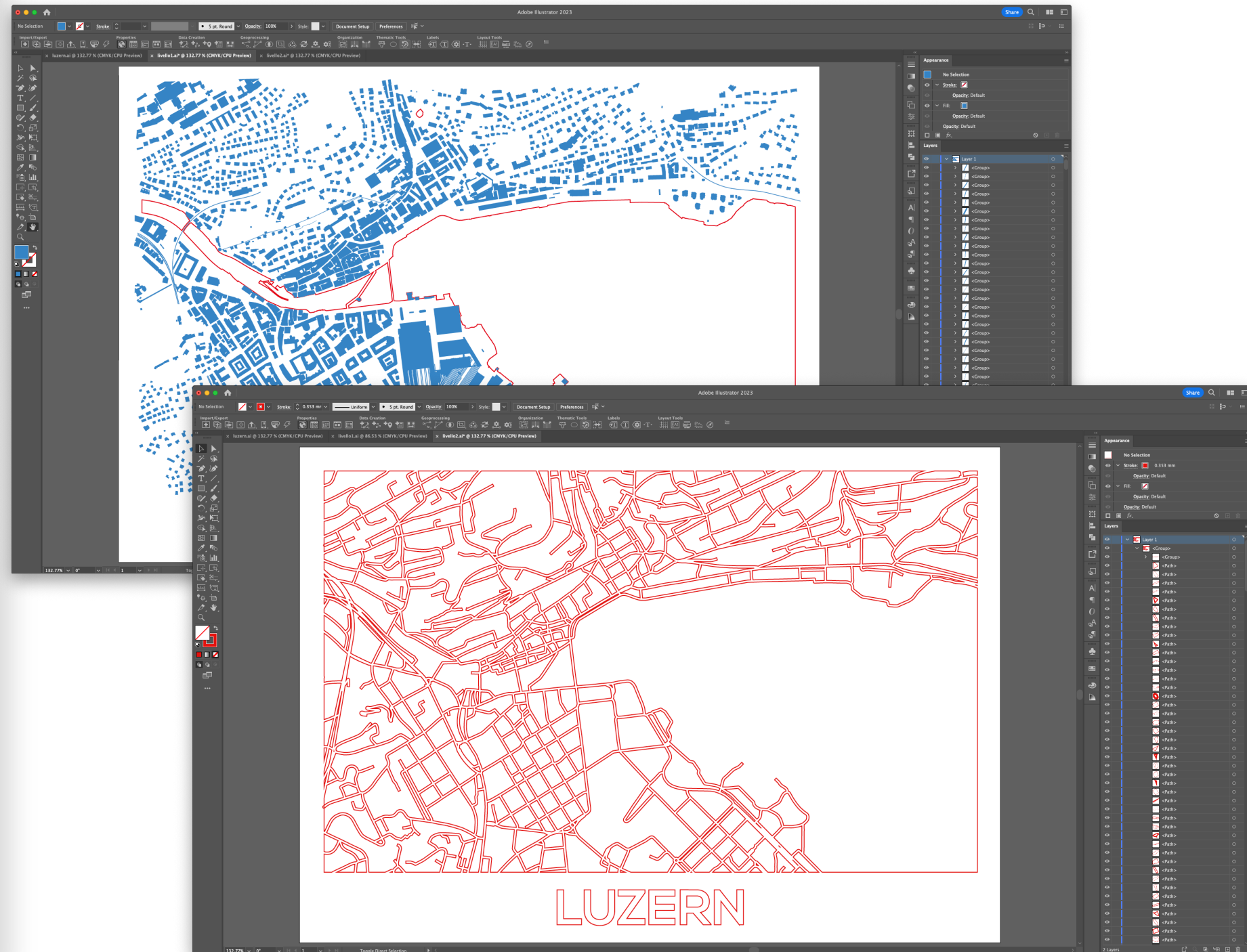
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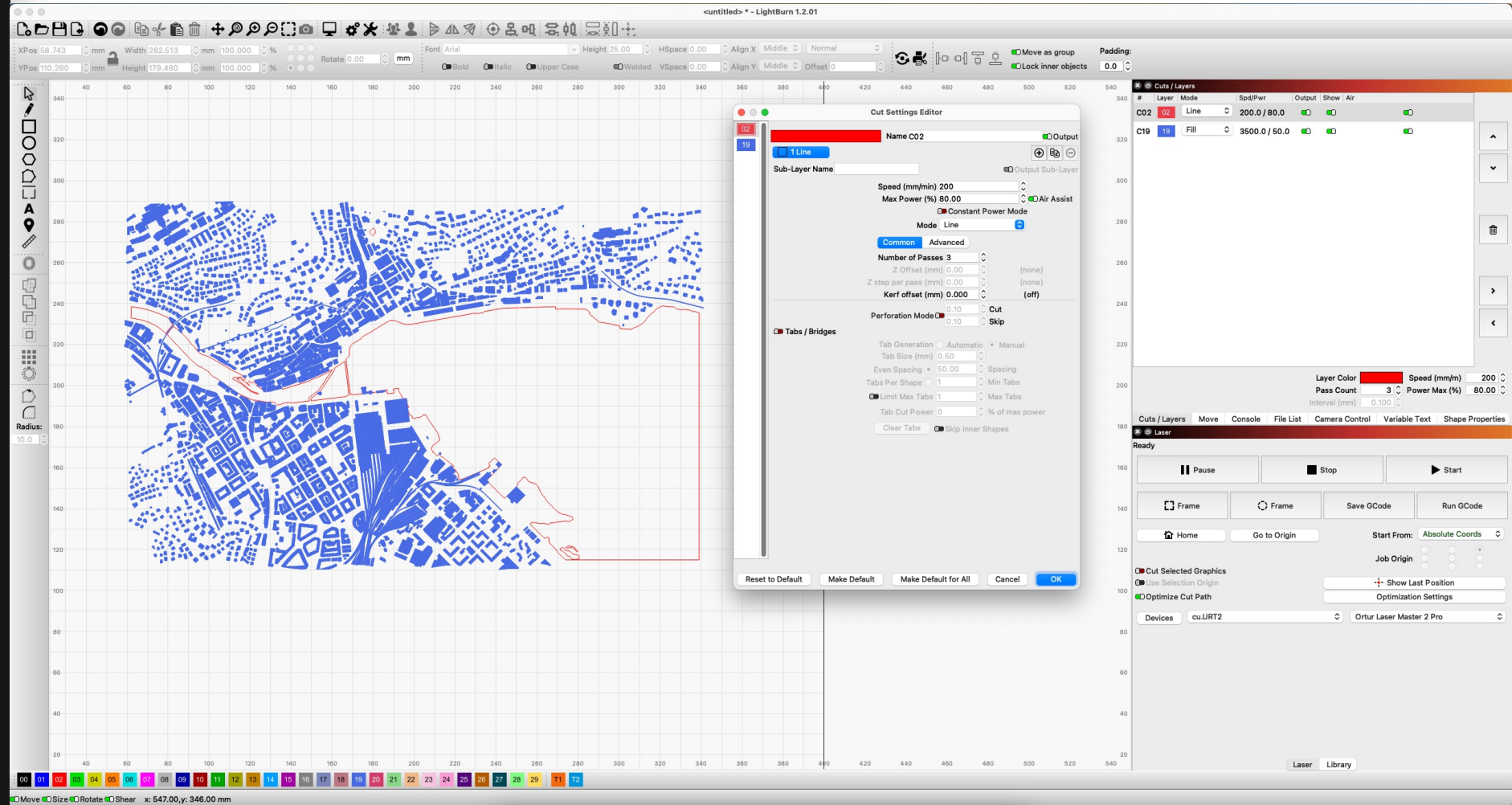
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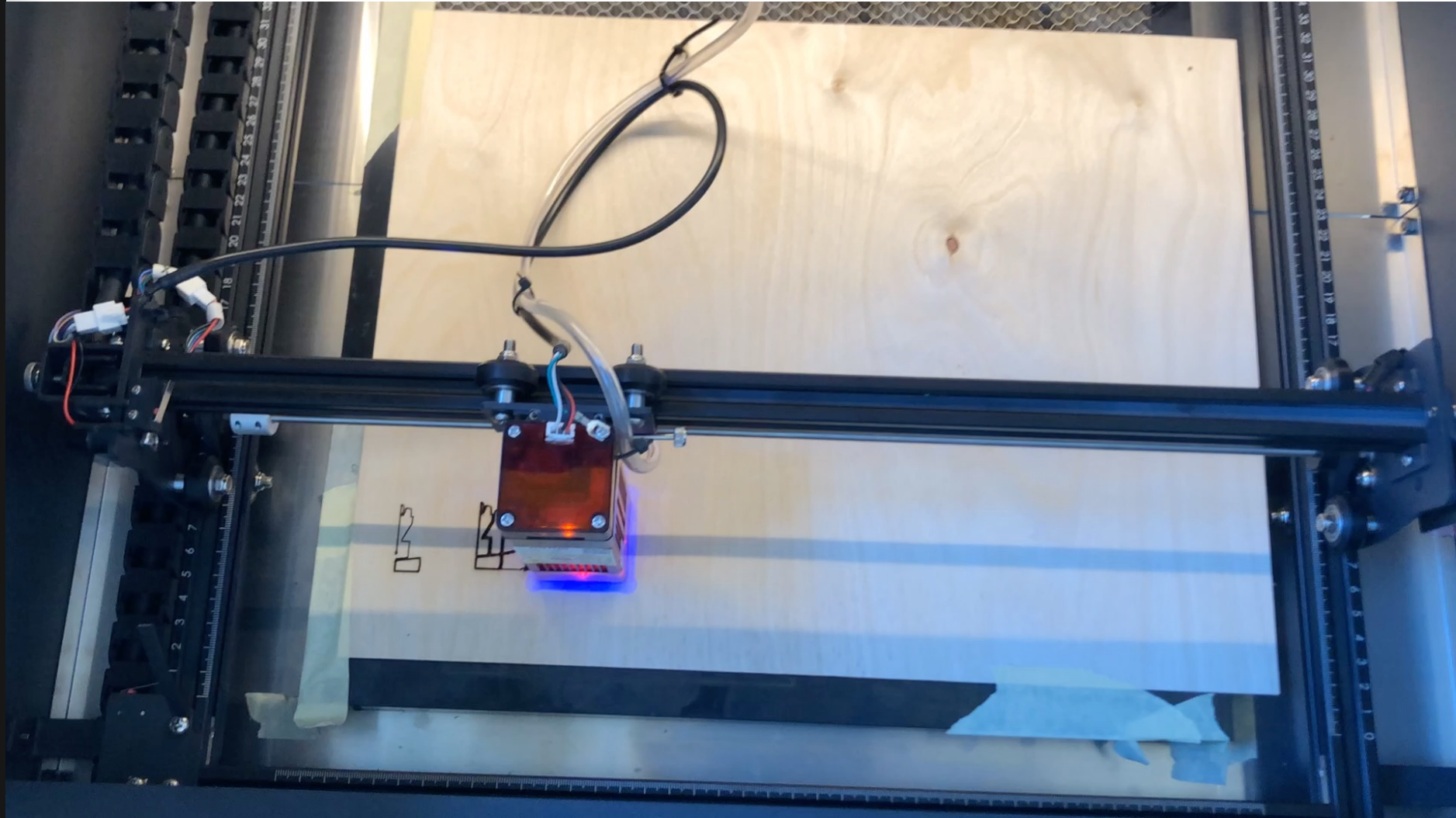
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AXON VIBE

three more smiles



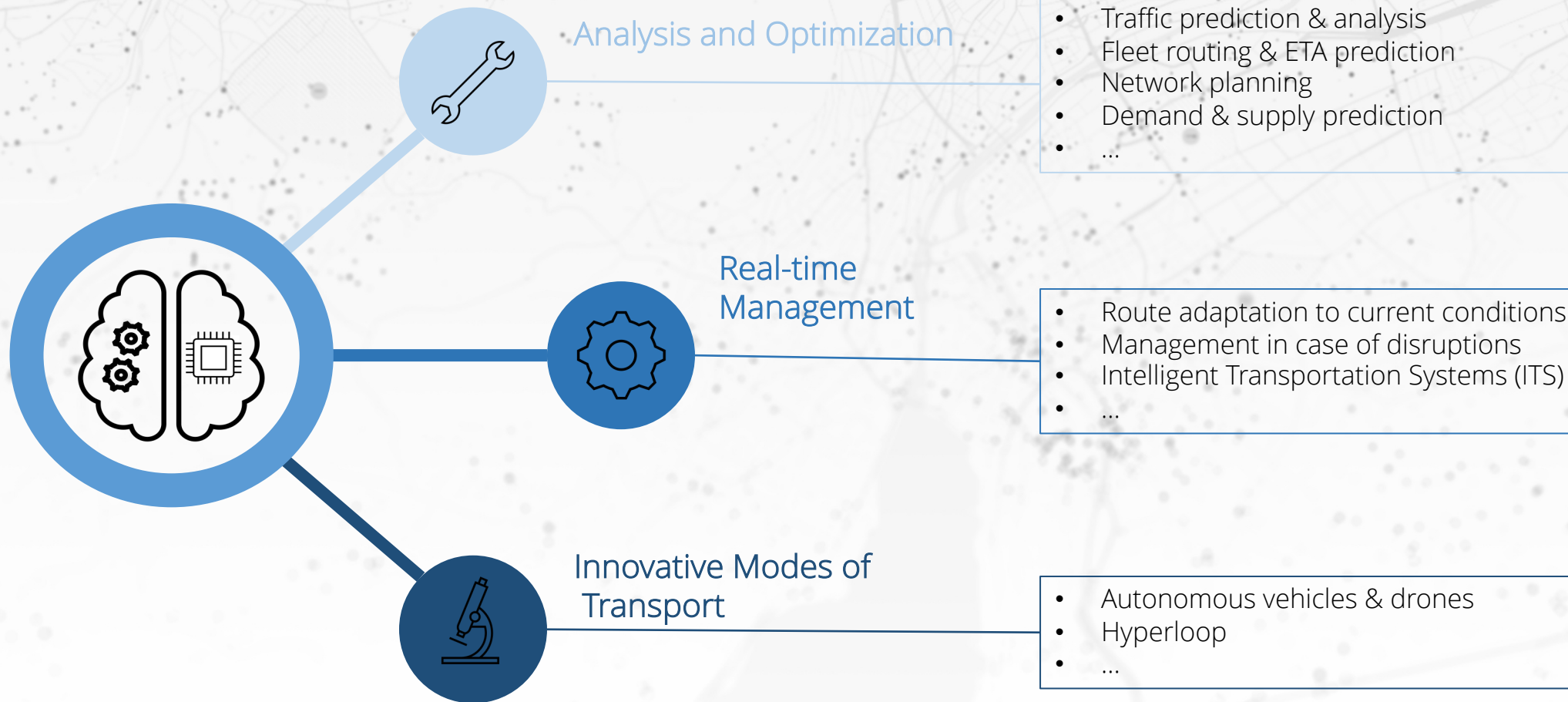
# Machine Learning for Transportation Systems

## How ML disrupts the way we solve transportation problems

Dr. David Jonietz  
Chief Analytics Officer  
Axon Vibe

# Potentials of ML for Transportation Systems

Where AI/ML can make a difference





# Computer Vision for Large Scale Traffic Prediction

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  $\approx$  Video prediction



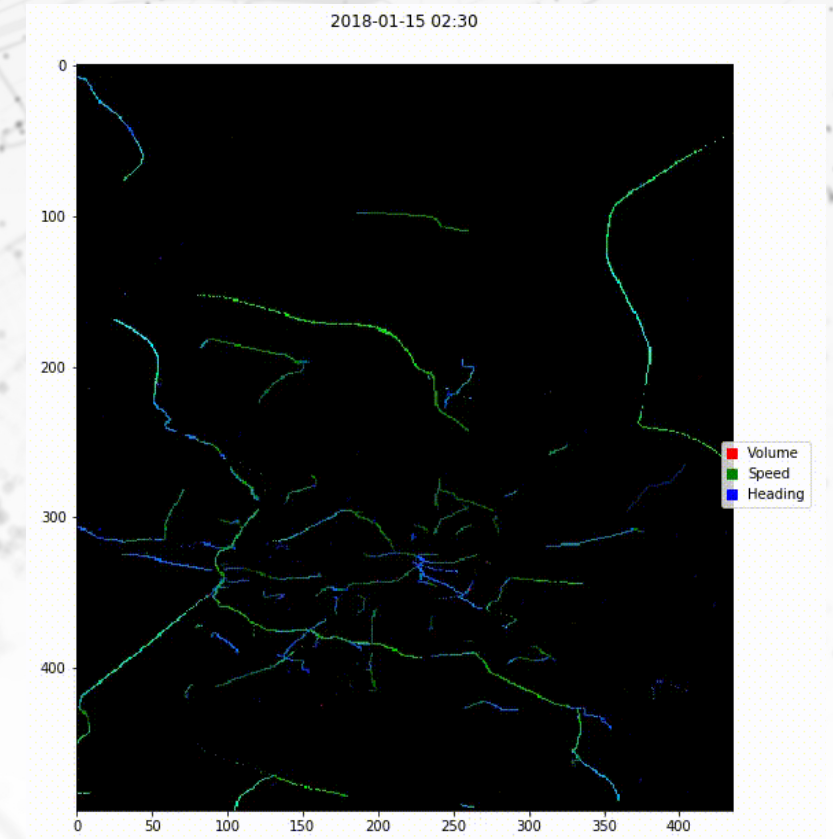
D. Jonietz, M. Kopp, M. Neun, B. Xu, A. Soleymani, "Method, apparatus, and system for end-to-end traffic estimation from minimally processed input data» United States Patent US20230067464A1

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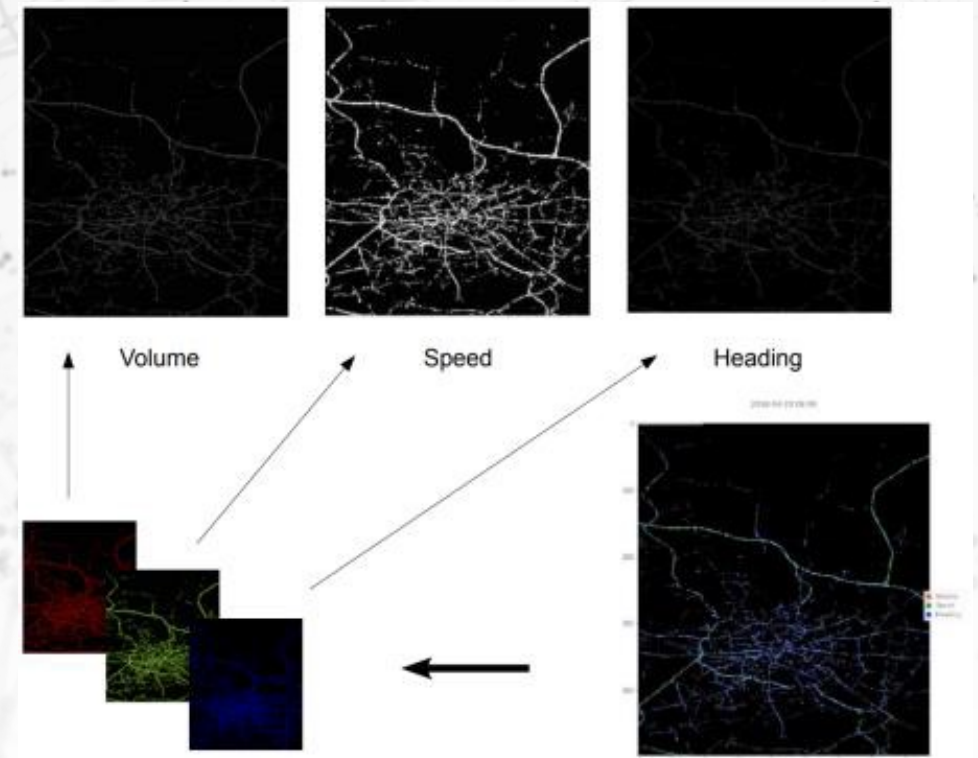


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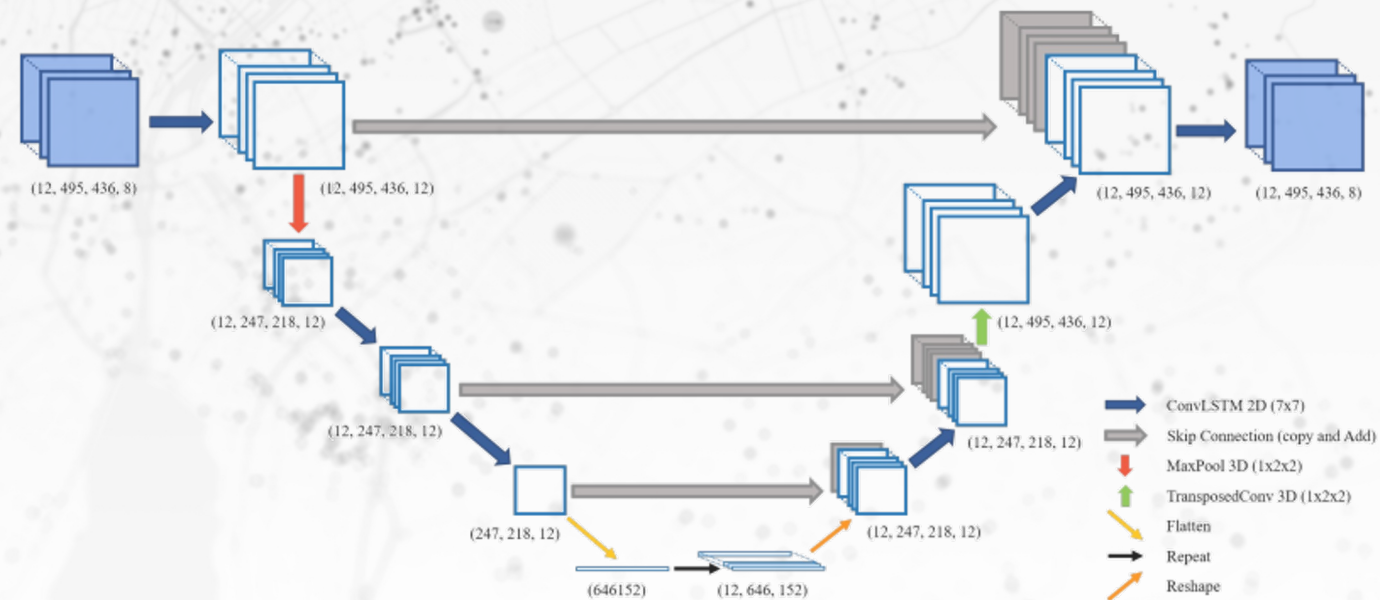
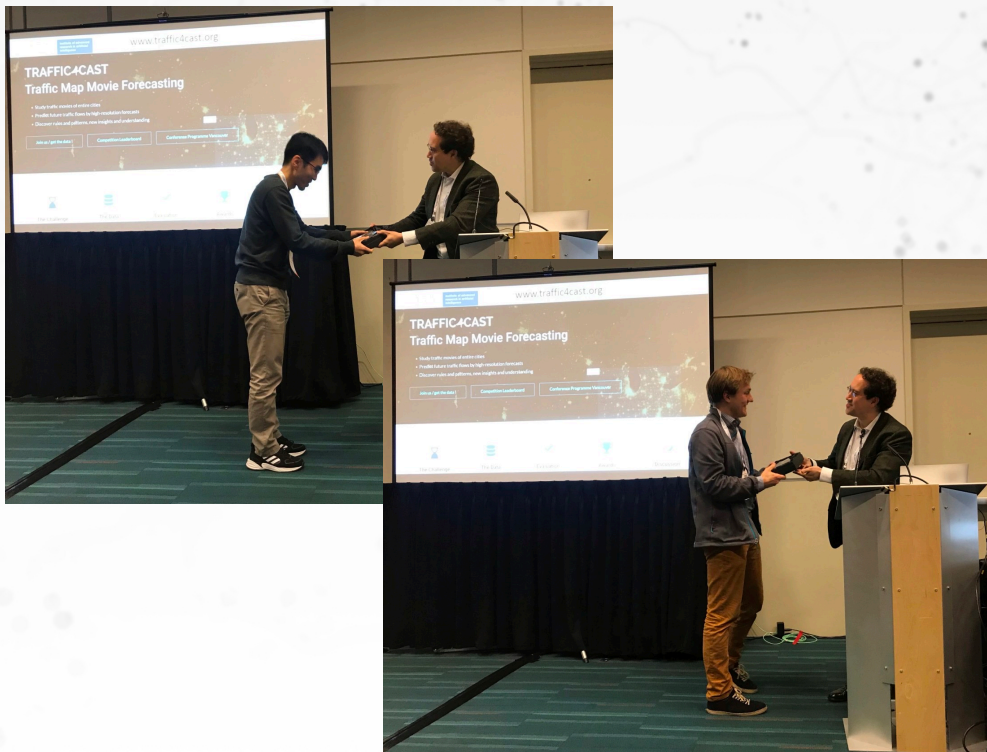


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# TRAFFIC4CAST



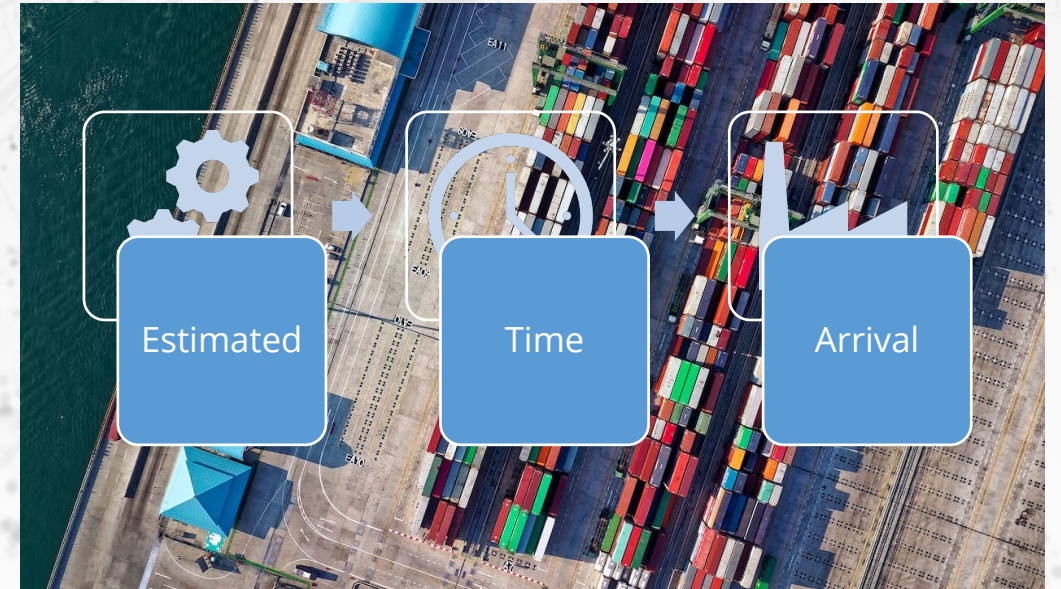


# STAMP – ETA Prediction with Machine Learning

Spatio-Temporal Aggregated Machine-learning Prediction (STAMP)

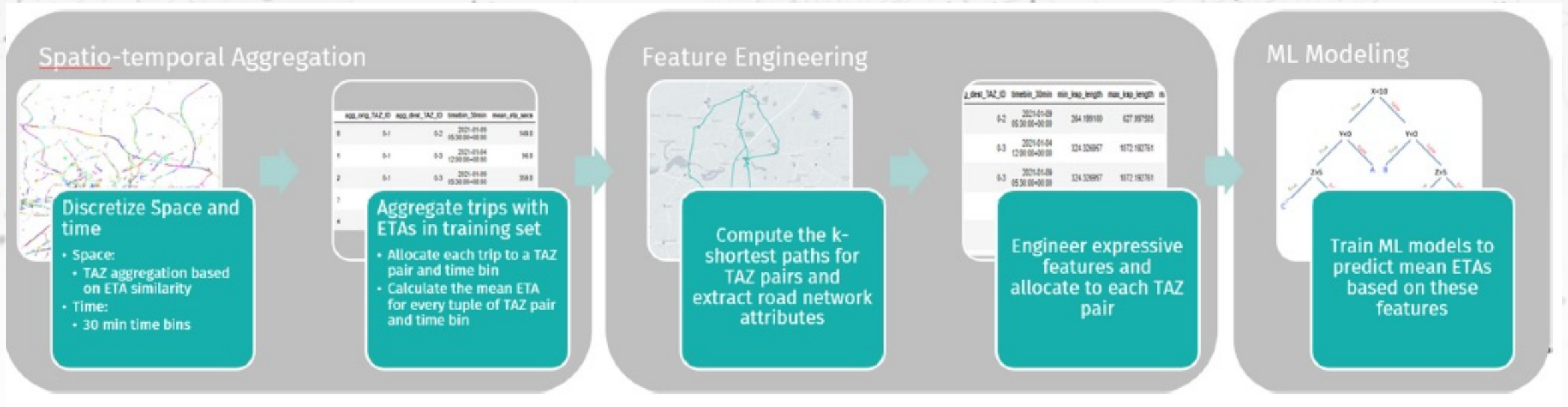
- Estimated Times of Arrival as essential information for logistics, fleet management and navigation
- Traditionally based on pre-defined route (e.g., fastest path)
- What if we predicted robust ETAs without routes?

Historical Trips – Routes = ETA



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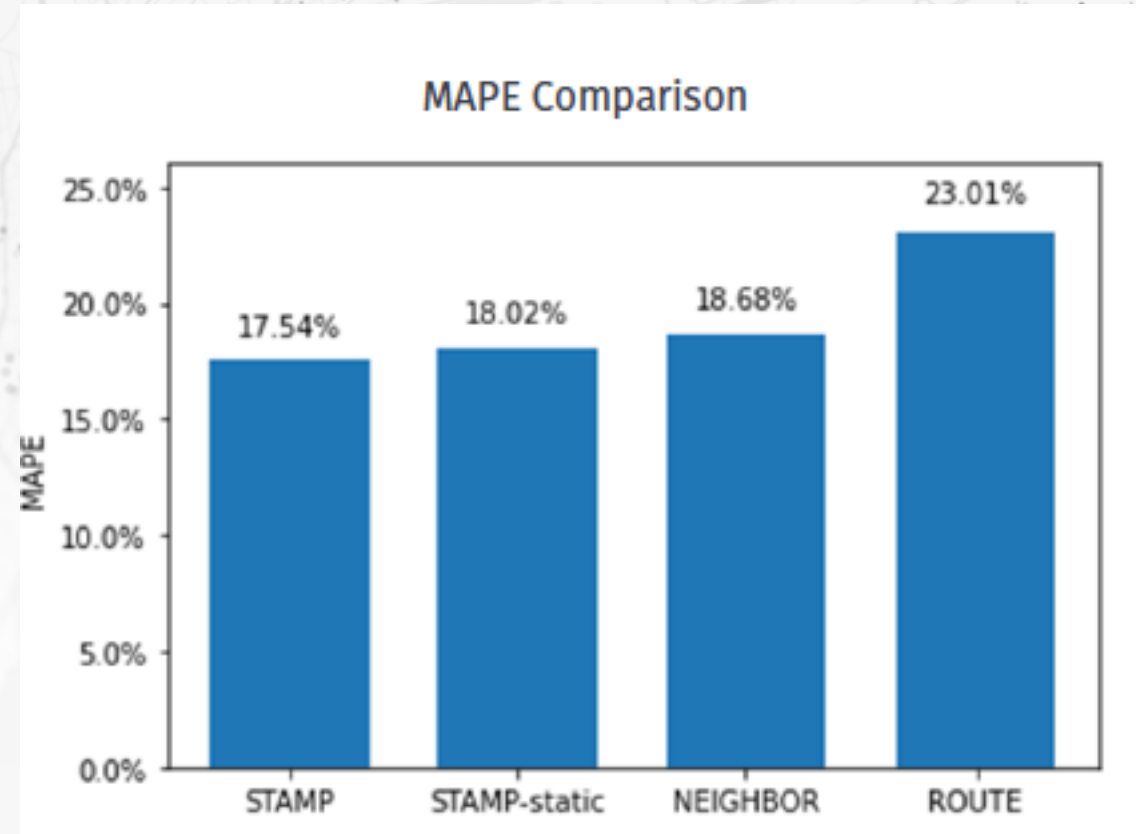


B. Xu, D. Jonietz, R. Gupta, A. Soleymani, K. Malm and R. Köhn, "STAMP: An Approach to ETA Prediction by Spatio-temporal Discretization and Machine Learning," *2022 IEEE 25th International Conference on Intelligent Transportation Systems (ITSC)*, Macau, China, 2022, pp. 893-900, doi: 10.1109/ITSC55140.2022.9922072.



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# 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