# Spatio-temporal Data Analytics with KNIME

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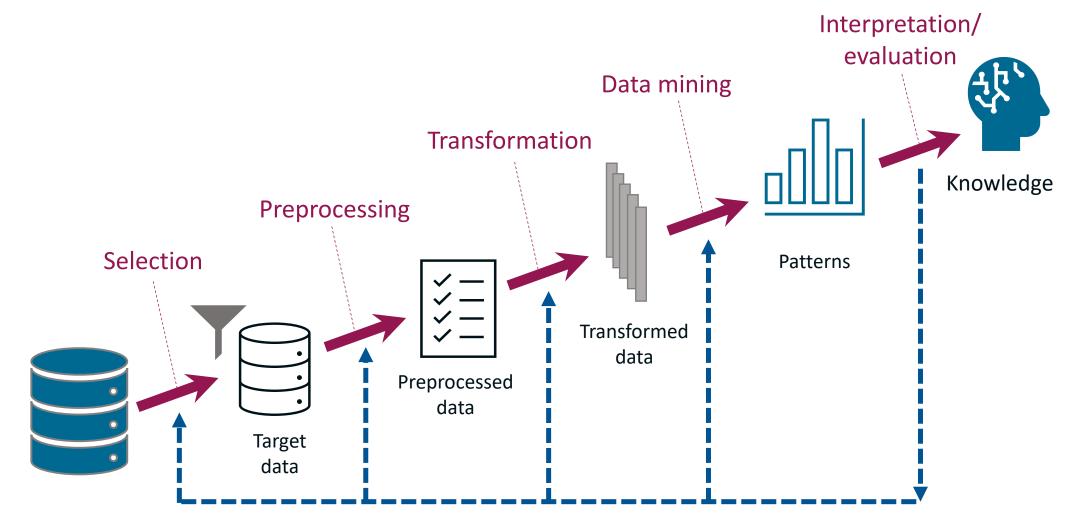
https://luistar.github.io

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#### Overview

- Knowledge Discovery from Data (KDD)
- Implementing KDD pipelines
  - Visual platforms (KNIME)
- KDD pipelines on Spatio-temporal (Mobility) Data
  - Peculiarities and Challenges
- A proposed solution: the KNIME Mobility Toolkit
- Examples from real-world Data Analytics scenarios

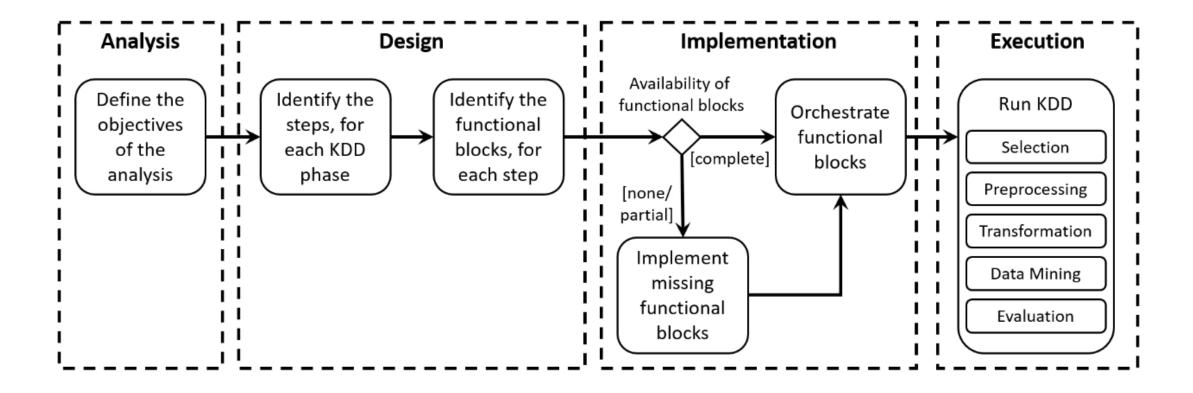
#### Knowledge Discovery from Data



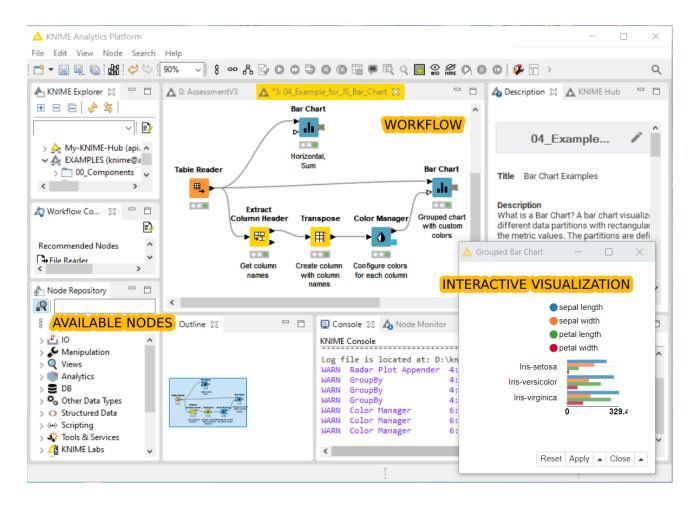
#### KDD pipelines

- Can be designed and implemented as modular software systems
- Each step is carried out by a specialized module (functional block)
- Modular design can help improve
  - **Understandability**: blocks are a first abstraction of the pipeline -> easier to focus on **what** we should do rather than on **how** should we implement it
  - Reusability: specialized functional blocks can be reused in other pipelines

#### Implementing KDD pipelines



#### Visual Data Analytics Platforms (Demo)

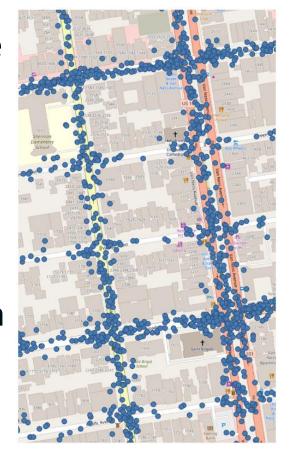


#### Spatio-temporal Mobility Data

- More and more sensors on vehicles + telematics = large amounts of spatio-temporal mobility data becoming available
- Typically consists of massive amounts of structured data
  - Timestamp
  - GPS position
  - Additional information (vehicle id, recorded temperature, speed, etc...)
- Some peculiar pre-processing steps are required to apply KDD

#### Trajectory Partitioning

- Many datasets consist of a single stream of spatio-temporal data
- A first step in most analyses is to split such data stream according to some criteria (e.g.: by vehicle, by trip, etc...)
- In most cases, the goal is to get a set of independent routes from an origin to a destination





### Map Matching

- Connecting subsequent GPS positions is typically not good enough to reconstruct accurate trajectories
- Map Matching aims at aligning raw (possibly inaccurate) positioning data with an underlying logical representation of the road network





#### Trajectory Restoration/Interpolation

- In presence of significant positioning errors and/or insufficient sampling rates, further processing of the trajectories might be required to reconstruct plausible trajectories
- In some cases, analysts might also be interested in manipulating the original trajectories to investigate what-if scenarios
  - What if all the taxis followed the shortest-path to destination?

#### Next Steps

- After trajectory partitioning, map matching and trajectory restoration, the next steps depend on the goals of the analysis
- In many cases, coverage analysis might be useful
- Given a set of trajectories:
  - How many times a certain road segment has been visited?
  - How frequently a certain area is visited by one of the monitored entities?

# Working with spatio-temporal data: Challenges

- Visual-based Tools do not support these peculiar preprocessing steps
- Practitioners typically re-implement the entire pipeline from scratch
  - Limited re-usability
  - Limited replicability
  - Hinders productivity

## KNOT: a KNime mObility Toolkit

KNOT is a KNIME plugin helping you visually compose intuitive, reproducible, and easy to distribute Knowledge Discovery pipelines for massive mobility datasets including — but not limited to — Floating Car Data (FCD) ones.



#### Get started in minutes

Getting started with KNOT takes only a few minutes. You install our nodes in KNIME and you're ready to go solve challenging issues!

#### **Customizable**

Most of the nodes provided by KNOT can be fine-tuned via the built-in dialogs, so that they can be effortlessly adapted to work with a number of different datasets and scenarios.

#### Focus on what really matters

KNOT offers out-of-the-box support for map matching raw trajectories to OpenStreetMap models, so that you can focus on the analyses that really matter.

#### **Easily Extendable**

KNOT is Open Source, and is designed to be easily extended. For example, if you are working on a novel map matching algorithm, you could easily extend our map matcher node to support it by implementing a simple Java interface.

#### We **Open Data**

KNOT supports OpenStreetMap data by default, and all the produced spatial objects are encoded in WKT format to maximize interoperability.

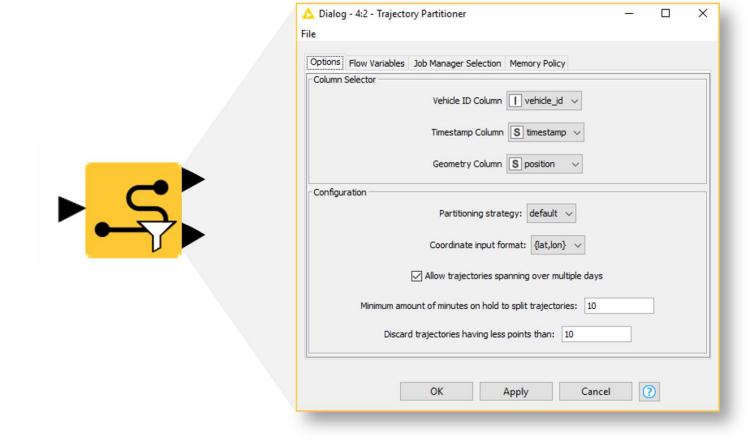
#### Leverage the power of KNIME

Along with the custom nodes KNOT provides, you can leverage the full power of the well-known KNIME Analytics Platform and streamline your knowledge discovery process with hundreds of additional nodes provided out of the box.

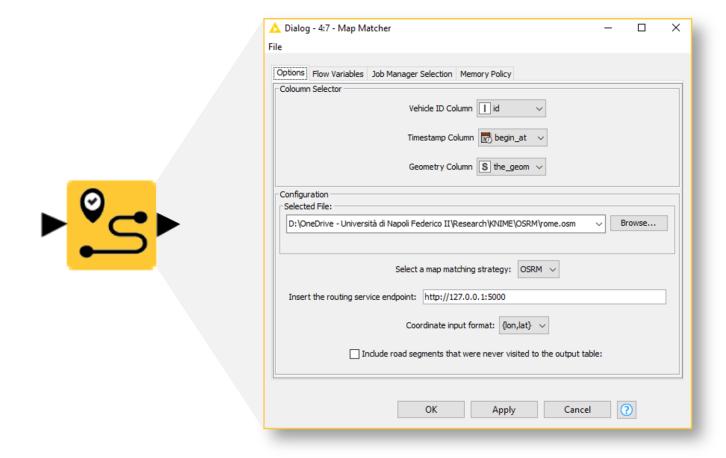
#### The KNIME Mobility Toolkit

- Collection of dedicated KNIME components to support typical spatiotemporal data processing steps
- Open-source, freely available (currently in the nightly builds)
- Each component can be customized and extended
- https://github.com/knot/

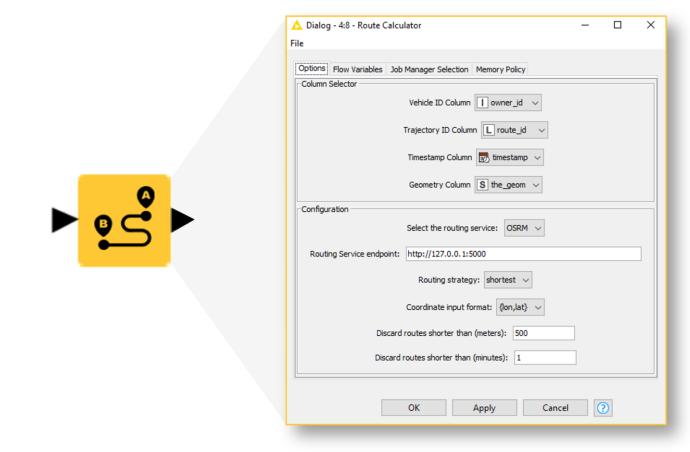
## Trajectory Partitioner



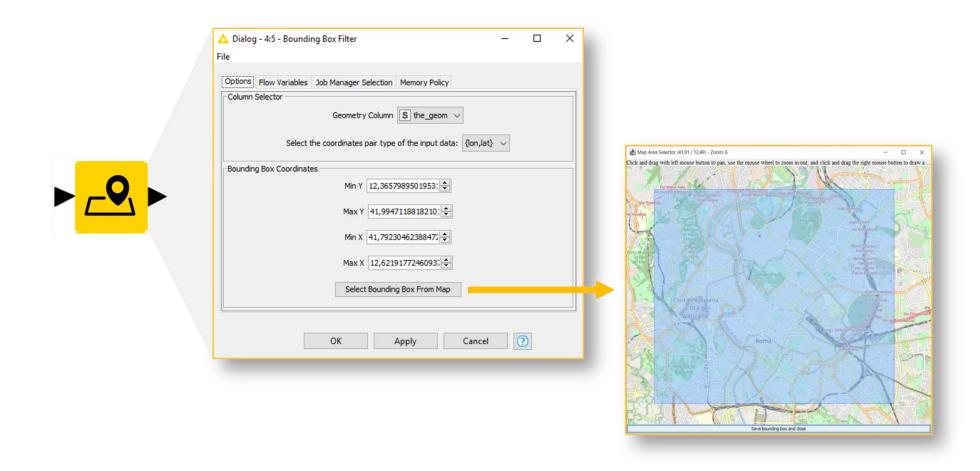
## Map Matcher



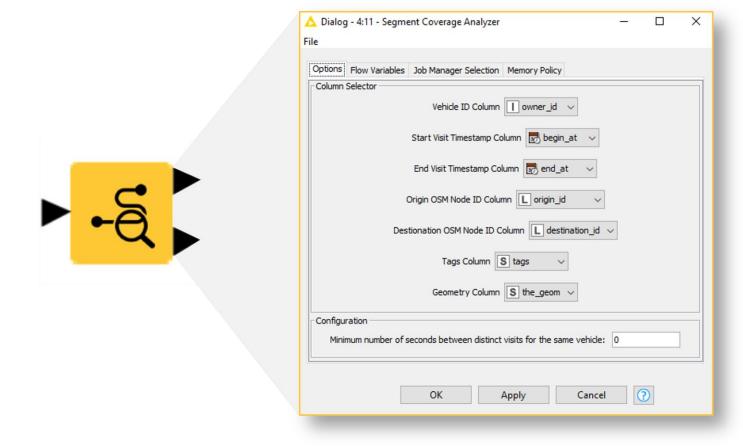
#### Route Calculator



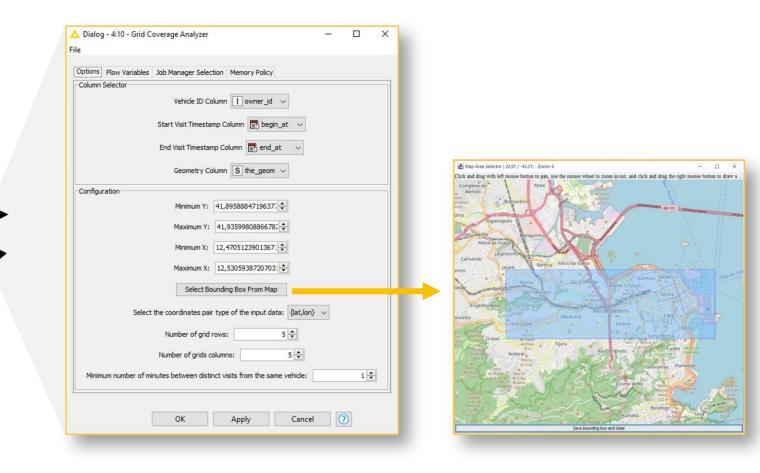
## Bounding Box Filter



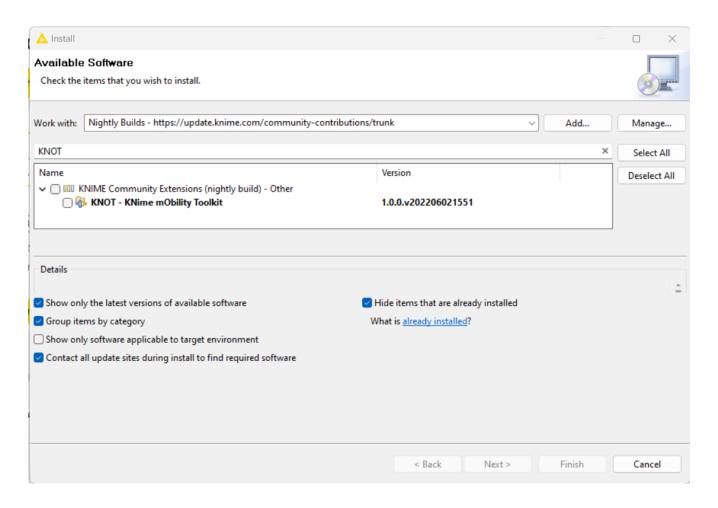
#### Segment Coverage Analyzer



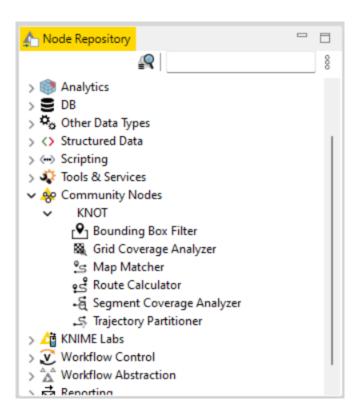
#### Grid Coverage Analyzer



## Installing KNOT



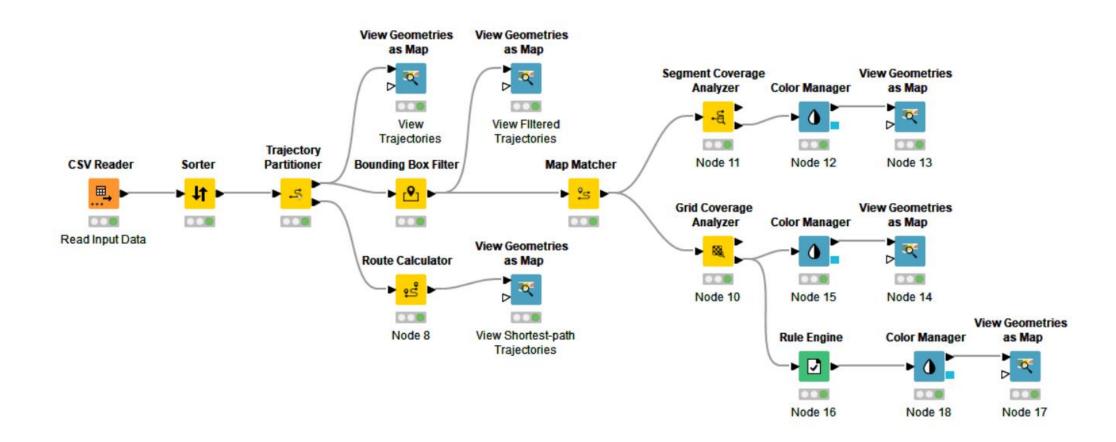
## Installing KNOT



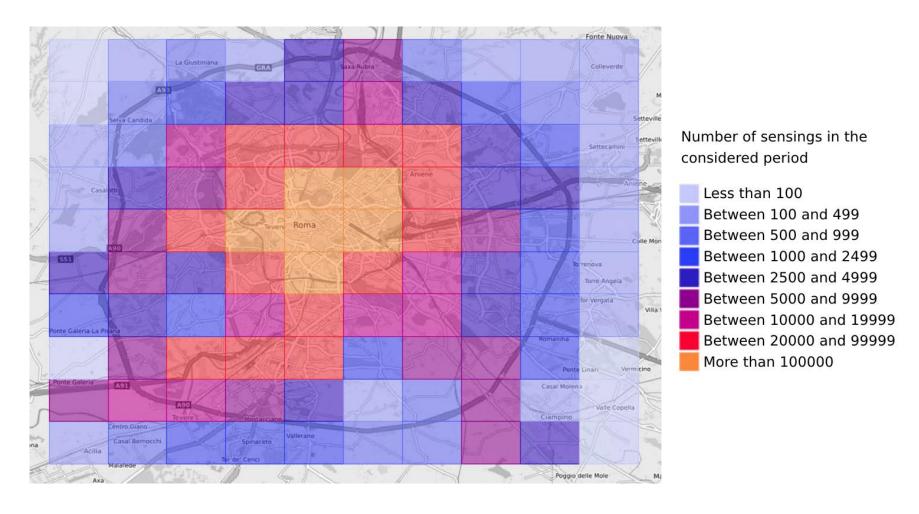
# Spatio-temporal Data Analytics: Real-world scenarios

- Can taxis be used for Vehicular Crowd-Sensing in Rome?
- How pervasive is public transportation in Rio de Janeiro?

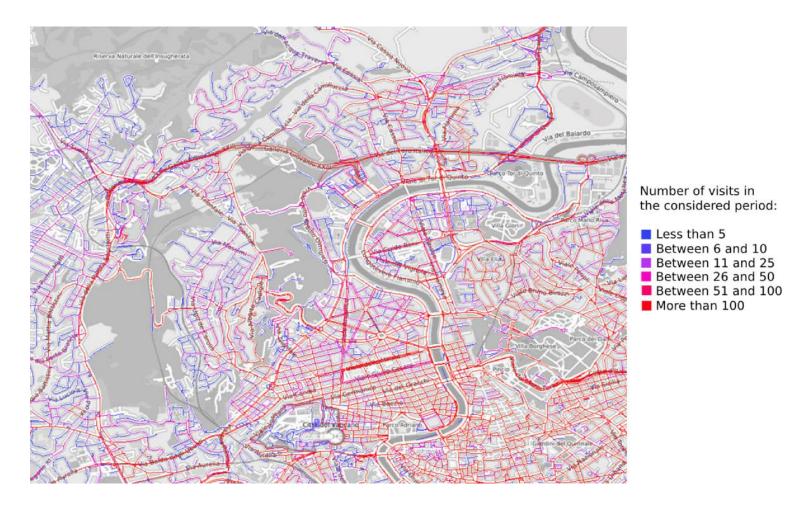
#### Road Network Coverage in Rome



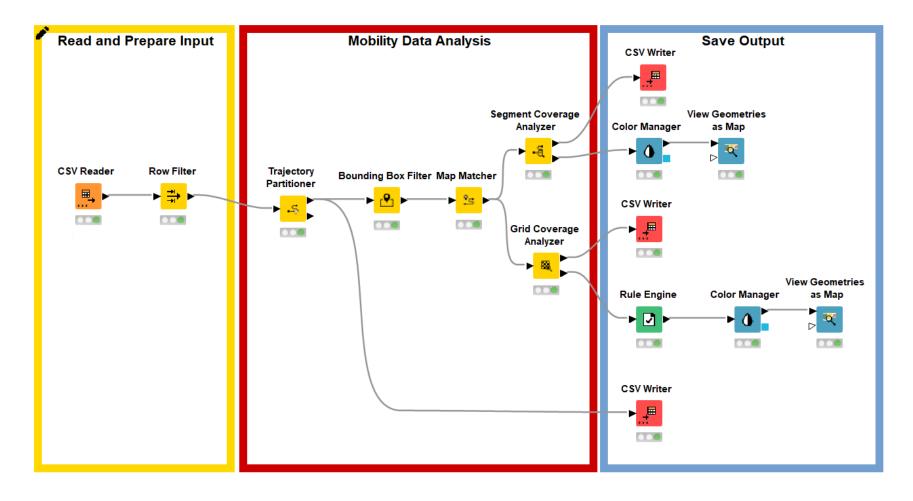
#### Area Coverage



## Segment Coverage



#### Public Transit Coverage in Rio De Janeiro



## Area Coverage



#### Take Home Messages

- Knowledge Discovery from Data
- Implementing KDD pipelines
- Visual Analytics Platform (KNIME)
- Challenges of KDD on Spatio-Temporal Mobility Data
- The KNIME Mobility Toolkit