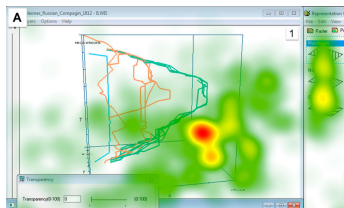
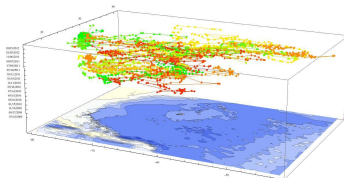


Parallel Detection of Movement Patterns in Large Spatio-temporal Datasets

February 1, 2019

Trajectory datasets

- ▶ Sensors, sensors everywhere!!!
- ▶ Anything that could move, will be tracked...
- ▶ Some applications:
 - ▶ Social behavior
 - ▶ Ecology (birds, sharks, ...)
 - ▶ Climate change (icebergs, cyclones, ...)
 - ▶ Software...



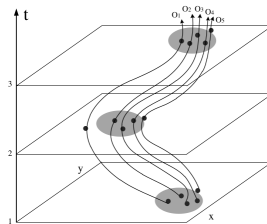
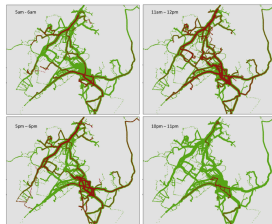
Complex movement patterns

Previous works focus on traditional queries:

- ▶ Range, Nearest Neighbors, Similarity, ...

Recent works look for the aggregate behavior:

- ▶ Moving clusters, Convoys, **Flocks**, Swarms, Gatherings, ...

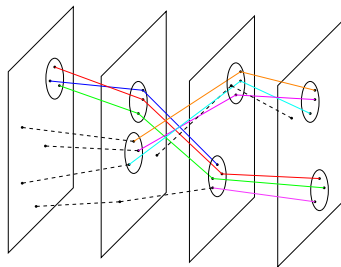


(b) Flock, convoy and swarm

What is a flock???

Definition $((\mu, \varepsilon, \delta) - \text{flock})$

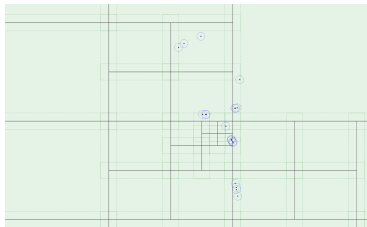
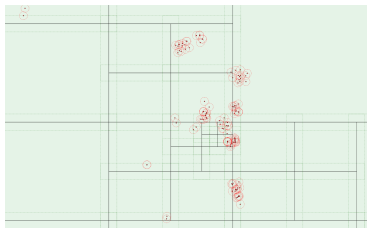
Sets of at least μ objects moving close enough (ε) for at least δ time intervals (Benkert et al, 2008).



- ▶ Vieira et al. (2009) proposed BFE algorithm (first polynomial solution).
- ▶ Drawbacks:
 1. Find disks is costly. They can be at any place.
 2. Huge amount of duplicate and redundant disks.
 3. Join between time intervals was a Cartesian product.

Contributions

1. Boost the detection of disks through a parallel approach (spatial partitioning + expansions).
2. Apply a frequent pattern mining approach to improve disk filtering (local + merge approach).
3. Use parallel distance joins to improve combination between consecutive time intervals (a Distance parameter).



Preliminar results

- Implementation using GeoSpark. Synthetic datasets using SUMO (Simulation of Urban Mobility).
- Berlin network (OSM). $\approx 20K$ points per timestamp, 10 timestamps. Varying ε ($\mu = 3$ and $\delta = 3$).

