

TRAJECTORY PATTERN MINING

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Introduction

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- Nowadays data on the spatial and temporal location of objects is available.
- Gps, GSM towers, etc
- What can we do with this data?

Introduction

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- Nowadays data on the spatial and temporal location of objects is available.
- Gps, GSM towers, etc
- What can we do with this data?
- Data mine it for patterns!! (of course)

Trajectory Pattern Mining

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- A way to represent Spatial patterns.

Bistro Square $\xrightarrow{10min}$ *Appleton Tower* $\xrightarrow{2hr30min}$ *Potterrow*

Chambers Str $\xrightarrow{15min}$ *The Mound* $\xrightarrow{5min}$ *Waverly Station*

Possible uses

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- Prediction of movement
- Aggregate movement behaviour
- Region of interests discovery
- Discovery of traffic flow & blockages

Structure of presentation

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- Introduction
- Mechanics
 - ▣ Data structure
 - ▣ Regions of Interest (ROI)
 - ▣ T-Patterns using these ROI
- Results

Data structure

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□ Raw Input

- List of $(Lat, Long, timestamp)$
- Multiple Moving Objects

□ Convert

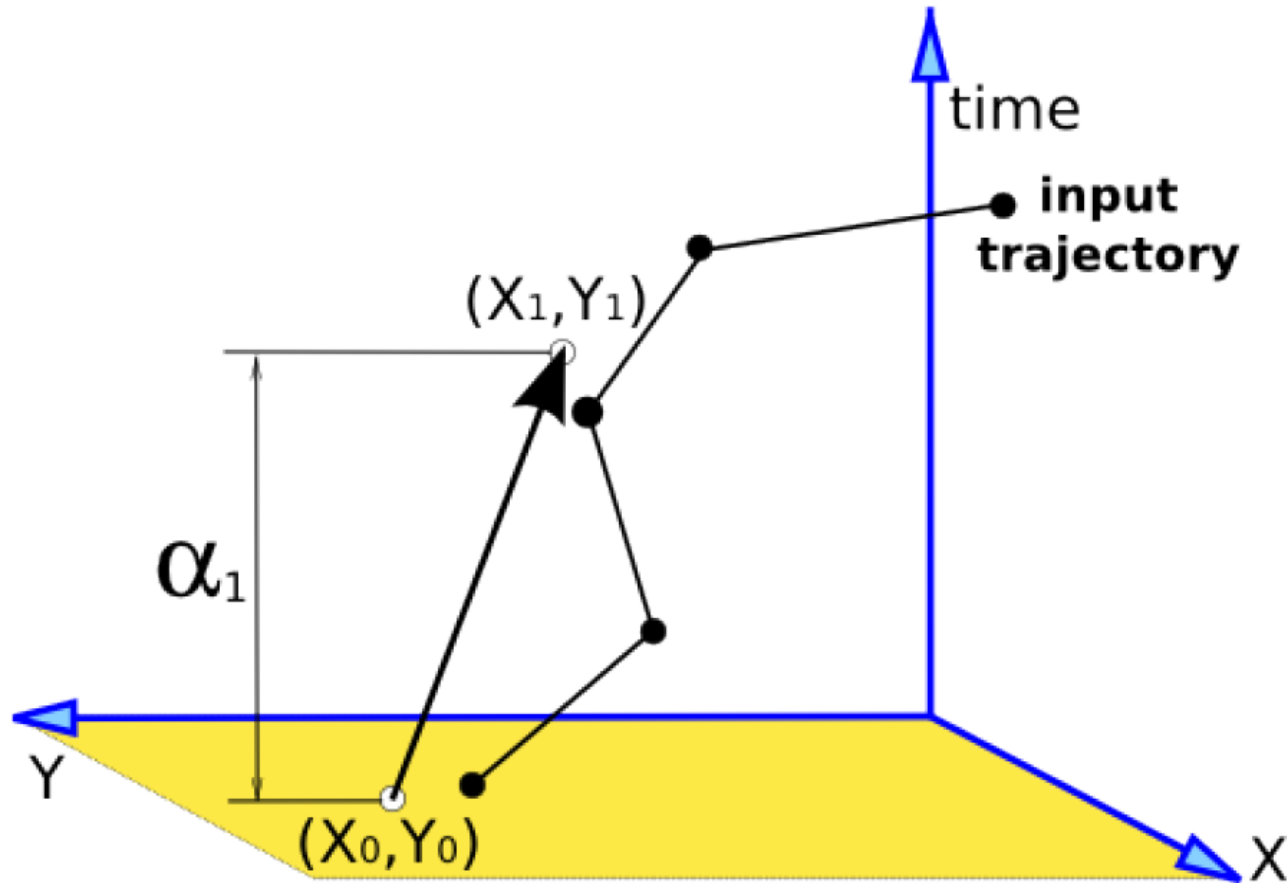
- $(Lat, Long, t) \xrightarrow{convert} (x, y, t) \quad (x, y) \in R^2$

□ T-Patterns

- $T = s_0 \xrightarrow{\alpha_1} s_1 \xrightarrow{\alpha_2} \dots \xrightarrow{\alpha_N} s_N$

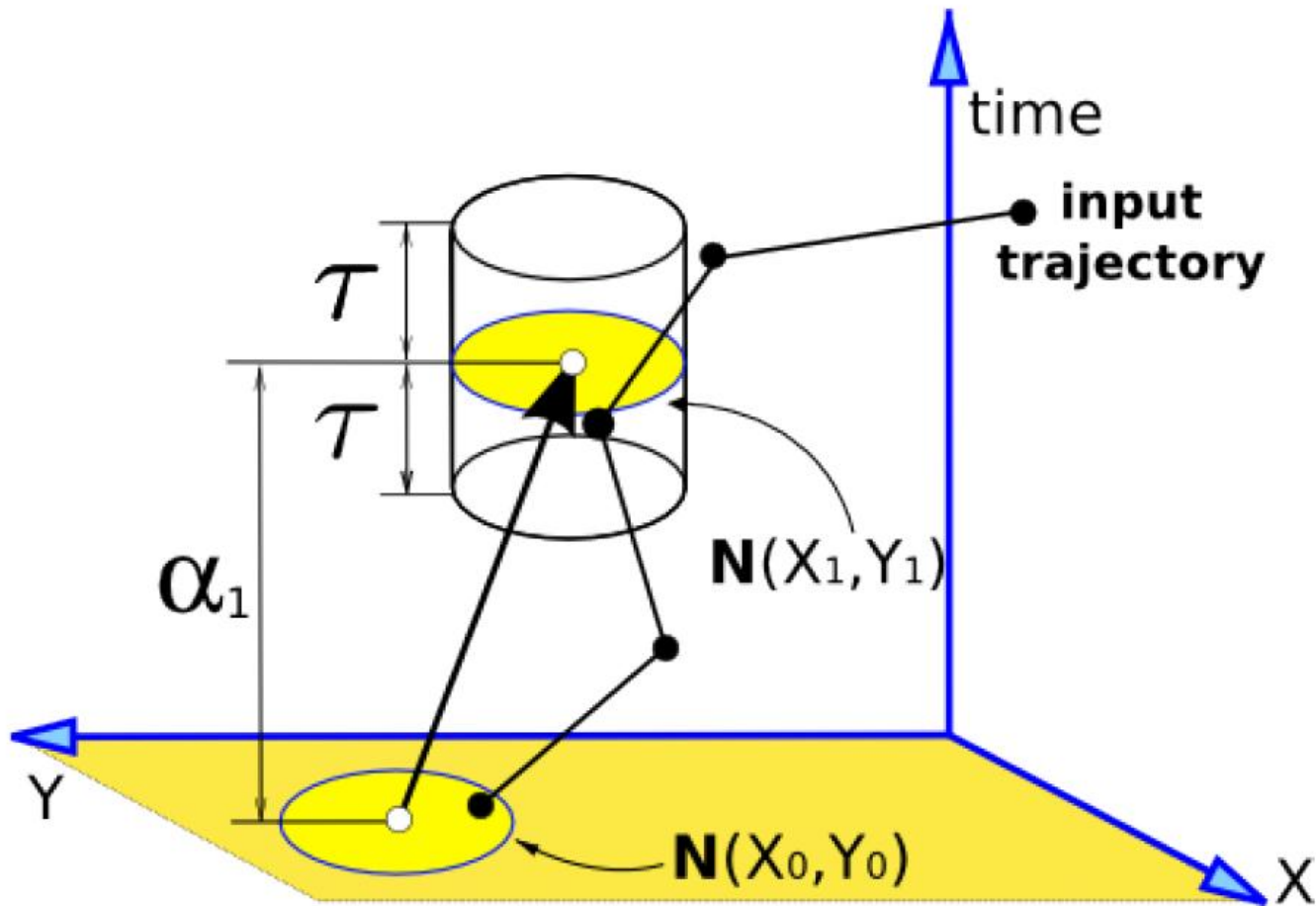
Example: Raw input

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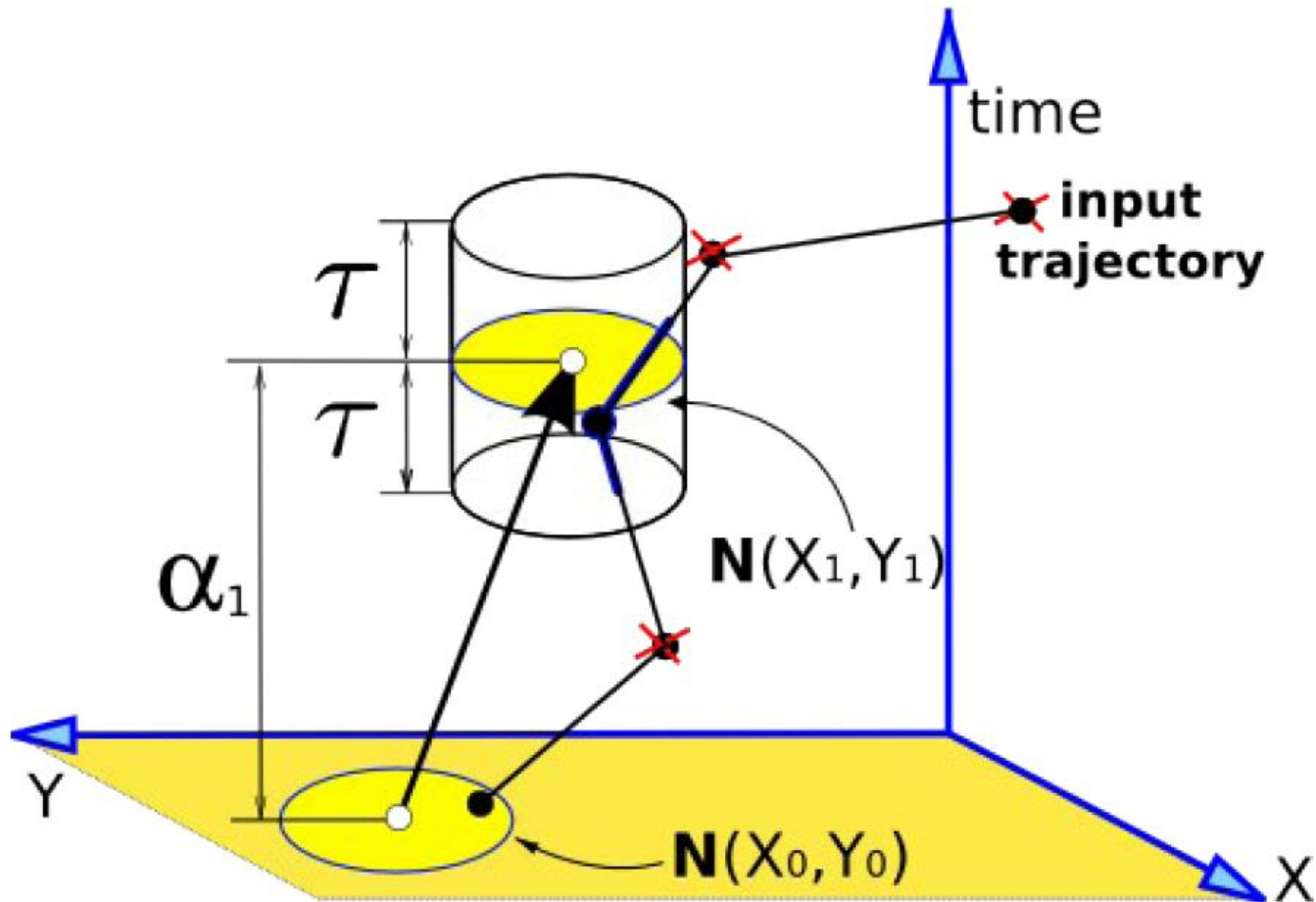
Example: T-Patterns

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Example: T-Patterns

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Support Threshold

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- Support: Percentage of trajectories which are contained within a pattern.

Example: Support threshold of 0.2.

- A T-Pattern is kept only if 20% of the trajectories support it.

Curse of Dimensionality

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- Data in 3 dimensional space (x, y, time).
- With Pattern length 3:
 $R^{(3 \text{ points} * 2D) + (2 \text{ transition} * 1D)}$ combinations
- As pattern length increases to n, search space
 $\text{Search Space} = R^{(n * 2) + ((n-1) * 1)} = R^{3n-1}$
- Therefore intractable.
- Reduce dimensions through Regions of Interest.

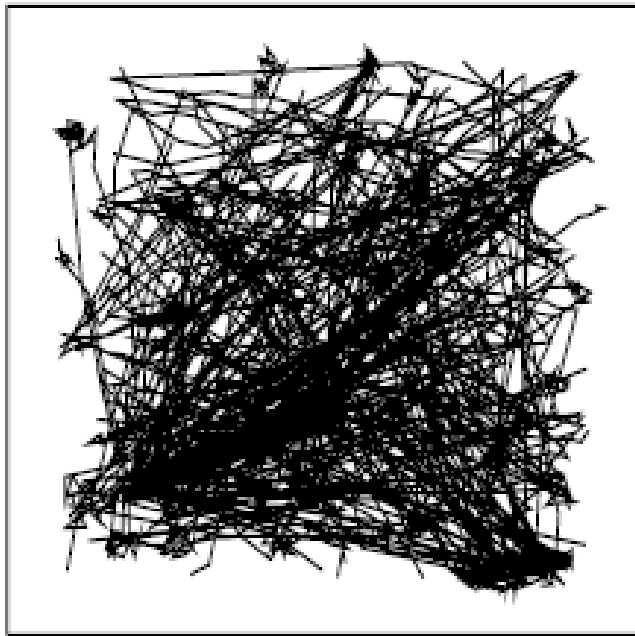
Regions of Interest

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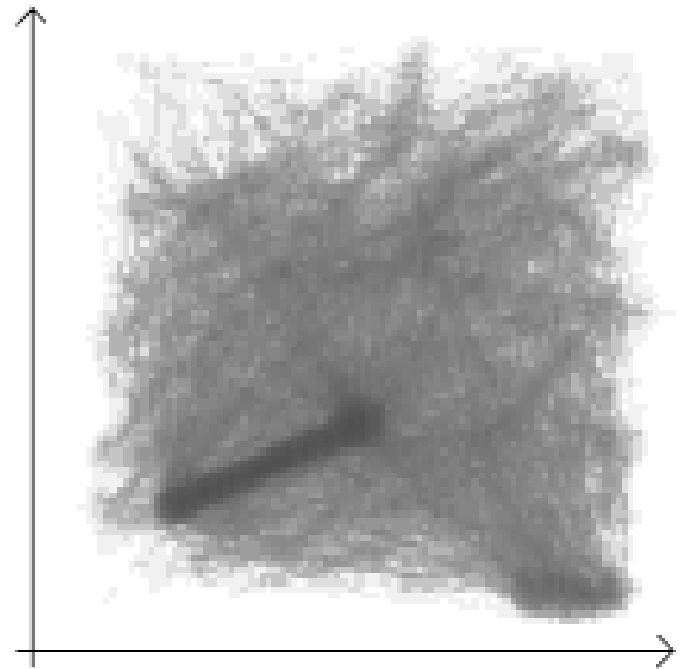
- Static pre-processed ROI
- ROI discovery
 - ▣ List of candidate places
 - ▣ Dynamic ROI
 - Popular points detection
 - ROI construction

Popular Points detection

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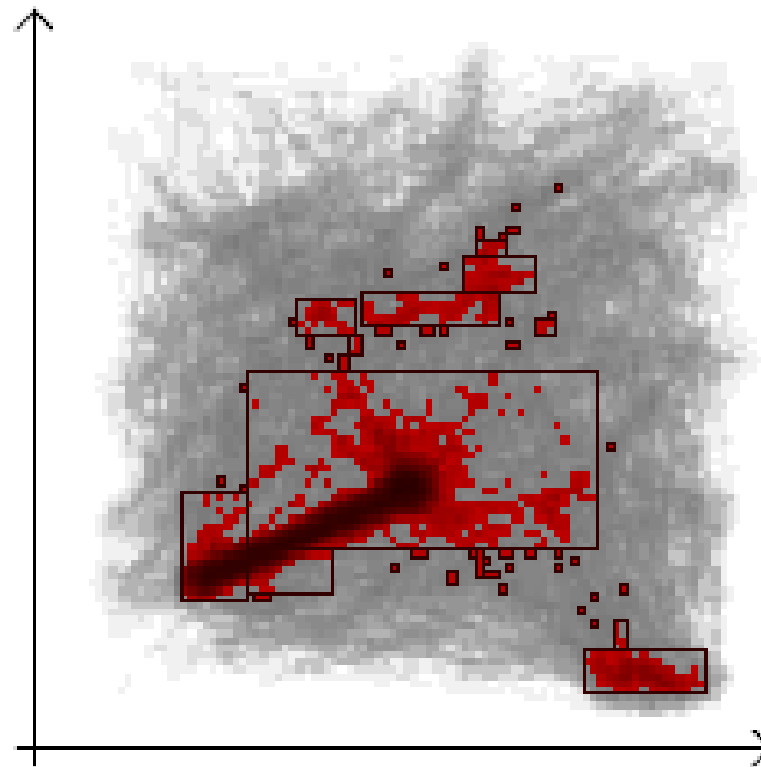
(a) input trajectories



(b) density distribution

ROI construction

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(c) dense cells and extracted ROI

T-Pattern Mining (1)

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- Two Approaches
 - Static
 - Dynamic
- Input: Trajectories and threshold parameters
 - Neighbourhood and time threshold
 - Support threshold
- Output: T-Patterns

T-Pattern Mining (2)

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- Step-wise Heuristic
- Any frequent T-Pattern of length $n+1$ is the extension of some frequent T-Pattern of length n .

$$\begin{aligned} (x_0, y_0) \xrightarrow{\Delta t_1} \dots \xrightarrow{\Delta t_{n+1}} (x_{n+1}, y_{n+1}) &\preceq_{N, \tau} T \\ (x_0, y_0) \xrightarrow{\Delta t_1} \dots \xrightarrow{\Delta t_n} (x_n, y_n) &\preceq_{N, \tau} T \end{aligned}$$

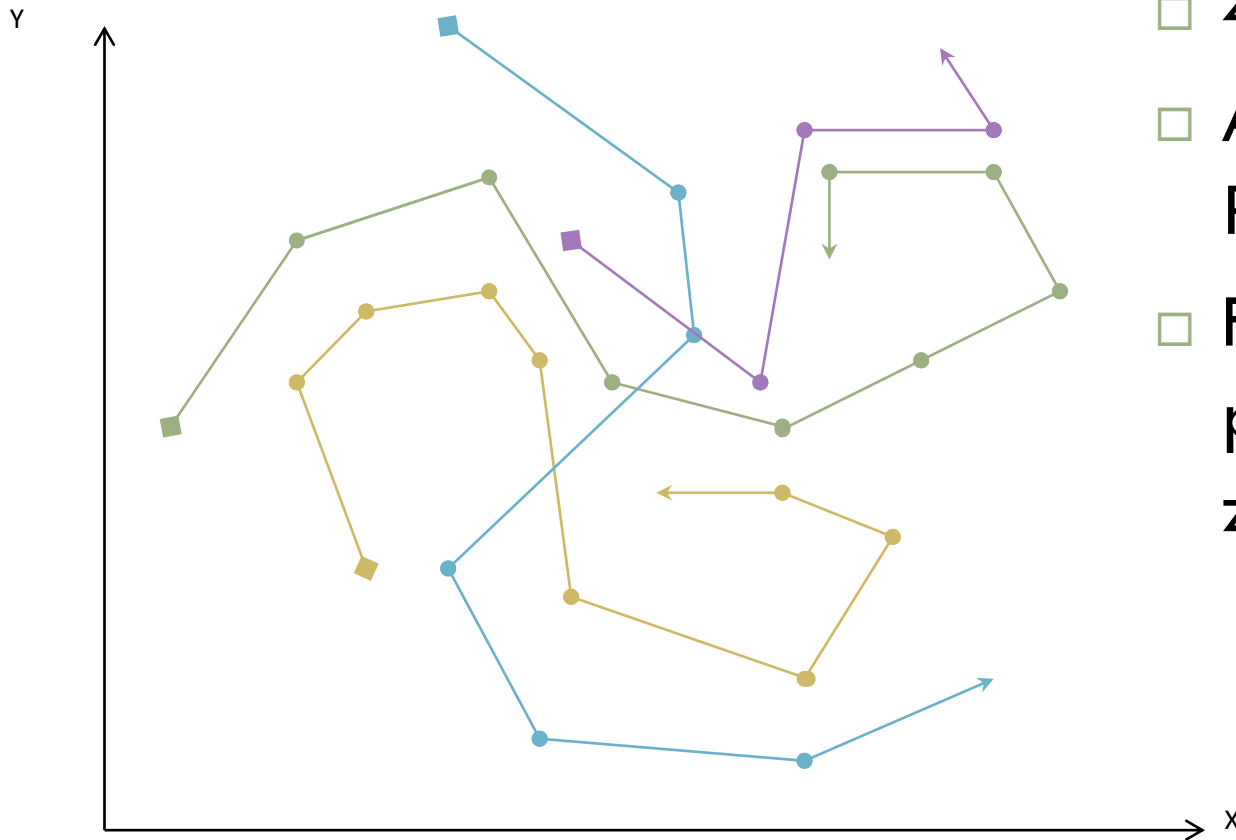
Static discovery of T-Patterns

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- Compute ROIs over the whole trajectories.
- Covert (x, y, t) trajectories to (ROI, t)
- Find sequences of ROIs which satisfy support condition.

Dynamic Discovery of T-Patterns (1)

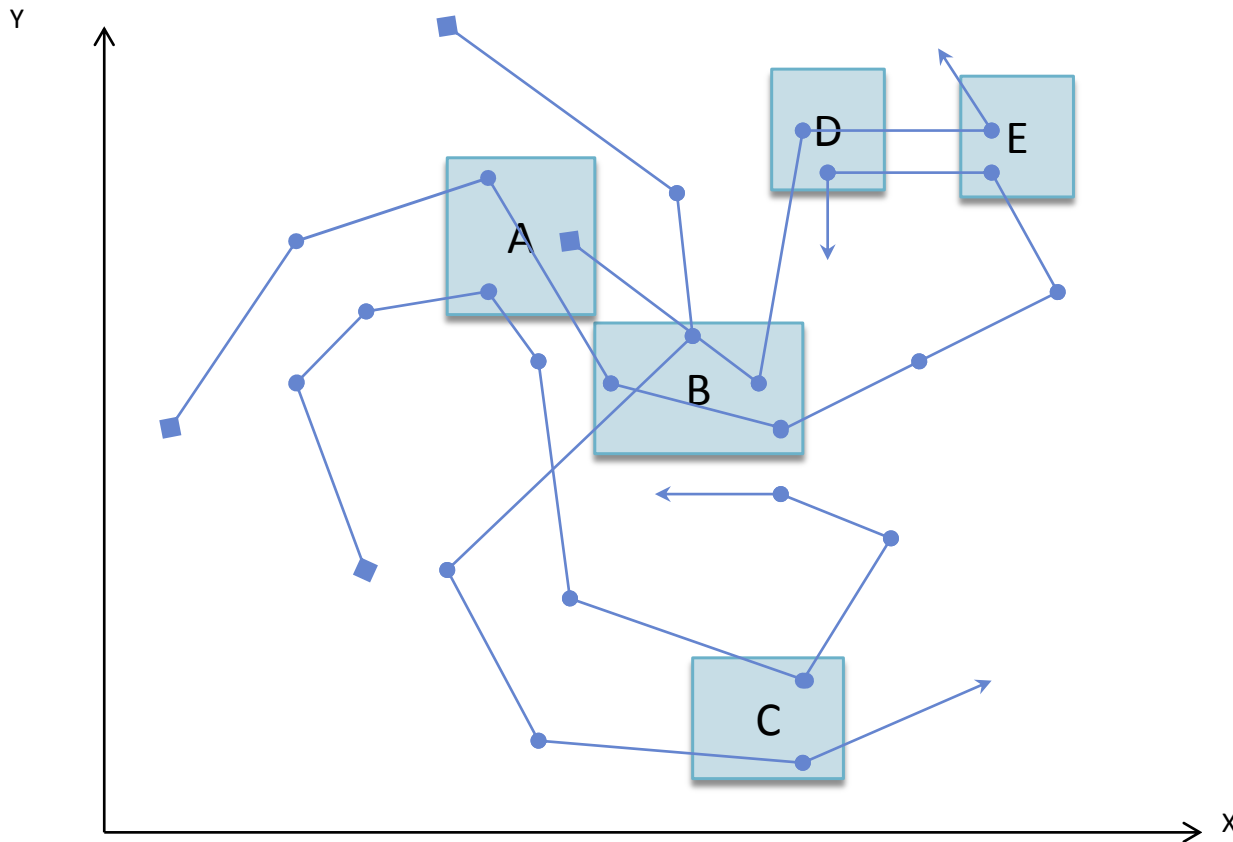
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- 4 trajectories.
- Aim is to find T-Patterns.
- First iteration prefix length is zero.

Dynamic Discovery of T-Patterns (2)

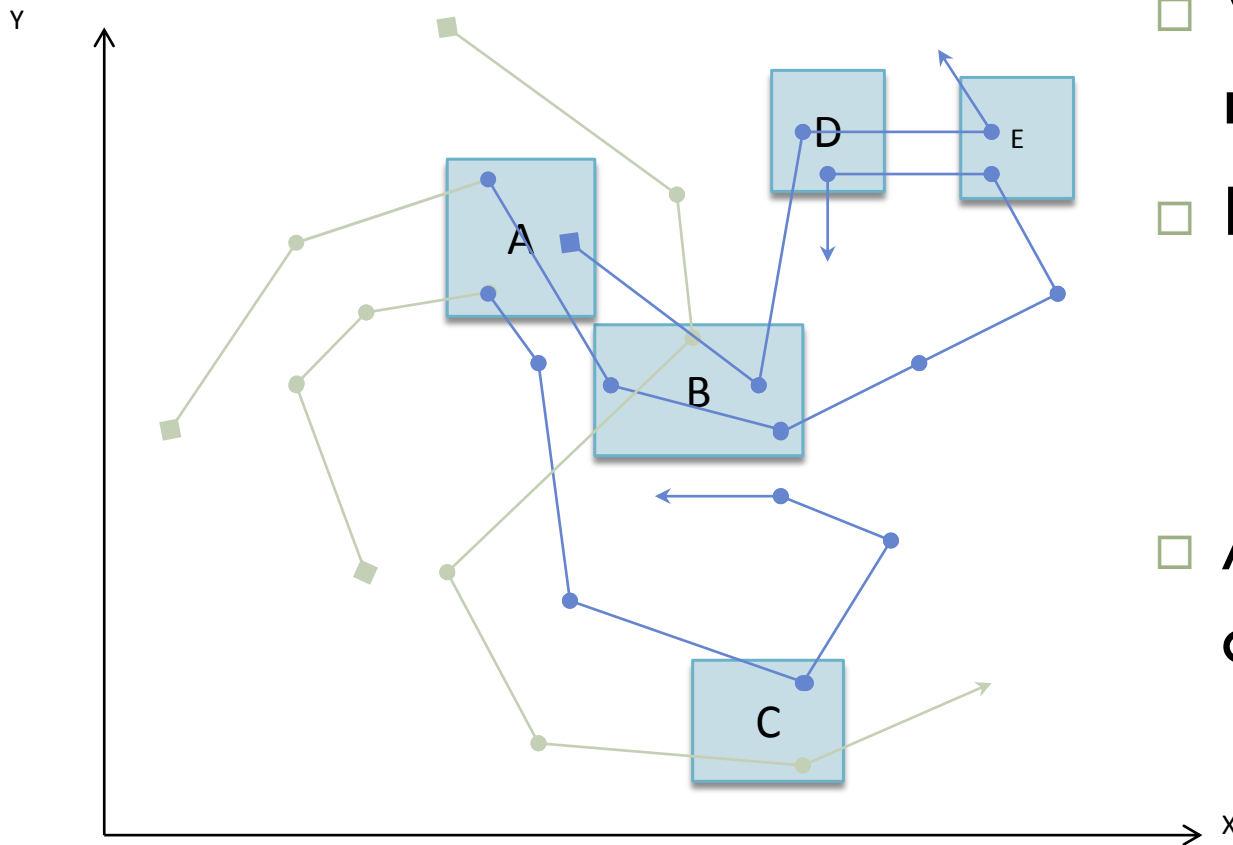
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- Detect Regions of Interest
- e.g. $support(r) \geq 0.5$
- For each region find possible projections to other regions

Dynamic Discovery of T-Patterns (3)

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□ With respect to region A.

□ Projections:

$A \rightarrow B$

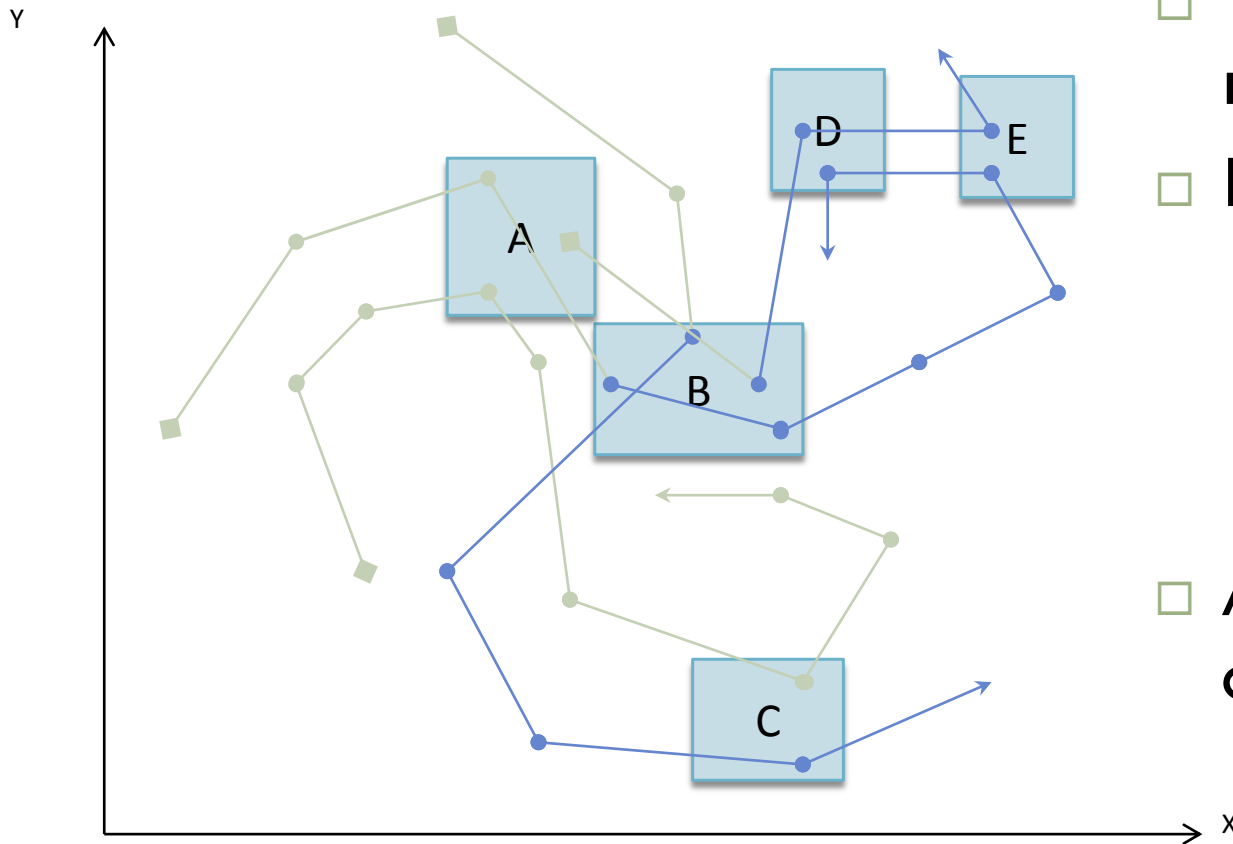
$A \rightarrow C$

□ Add new configuration

$(T', \langle A \rangle)$

Dynamic Discovery of T-Patterns (4)

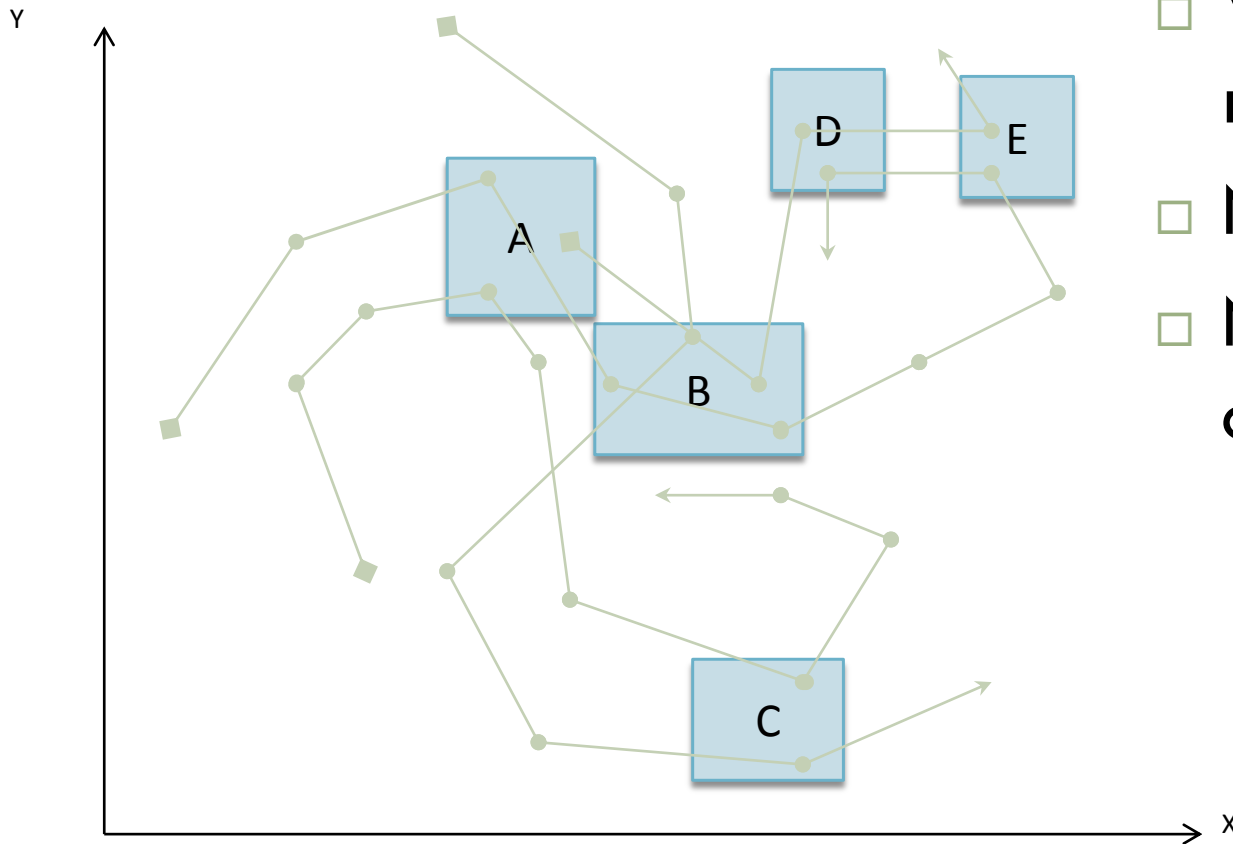
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- With respect to region B.
- Projections:
 - $B \rightarrow C$
 - $B \rightarrow D$
 - $B \rightarrow E$
- Add new configuration
 $(T', \langle B \rangle)$

Dynamic Discovery of T-Patterns (5)

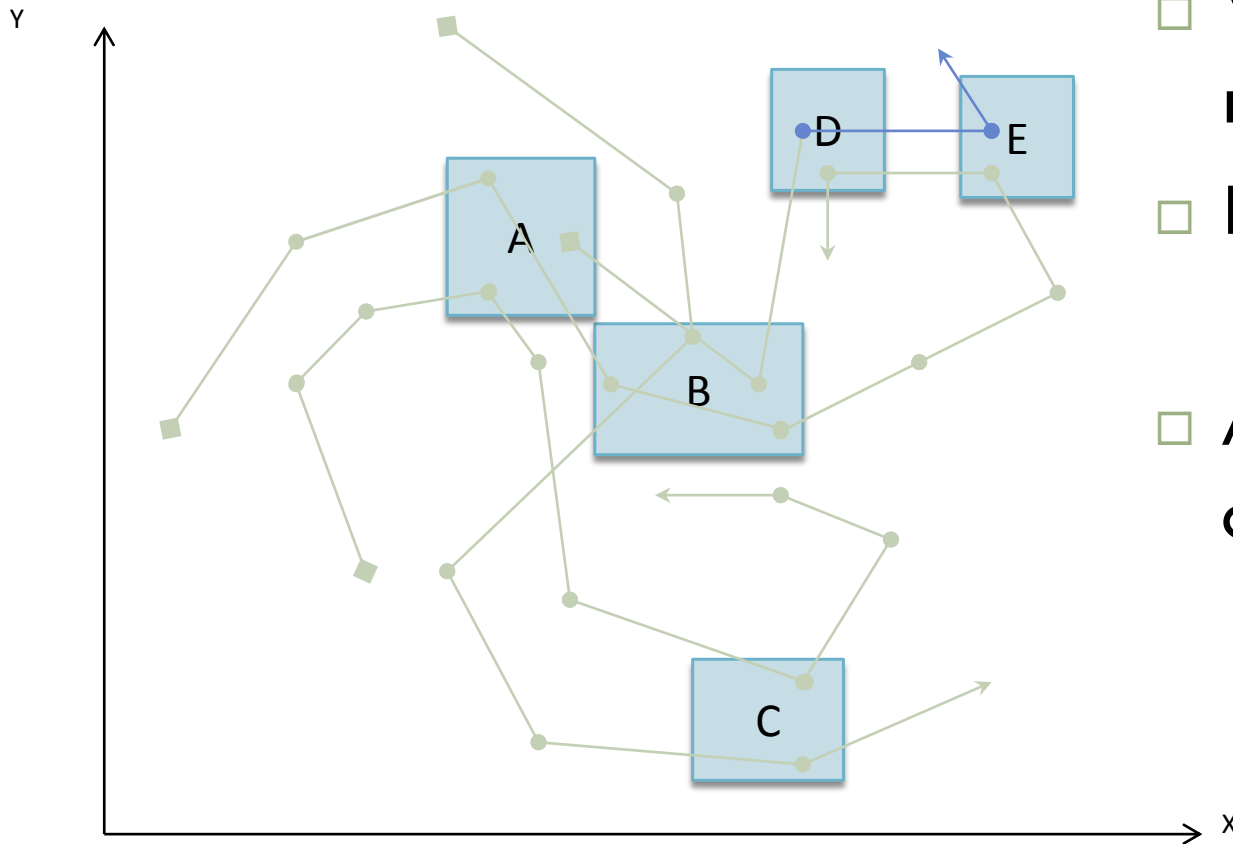
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- With respect to region C.
- No Projections
- No new configuration

Dynamic Discovery of T-Patterns (6)

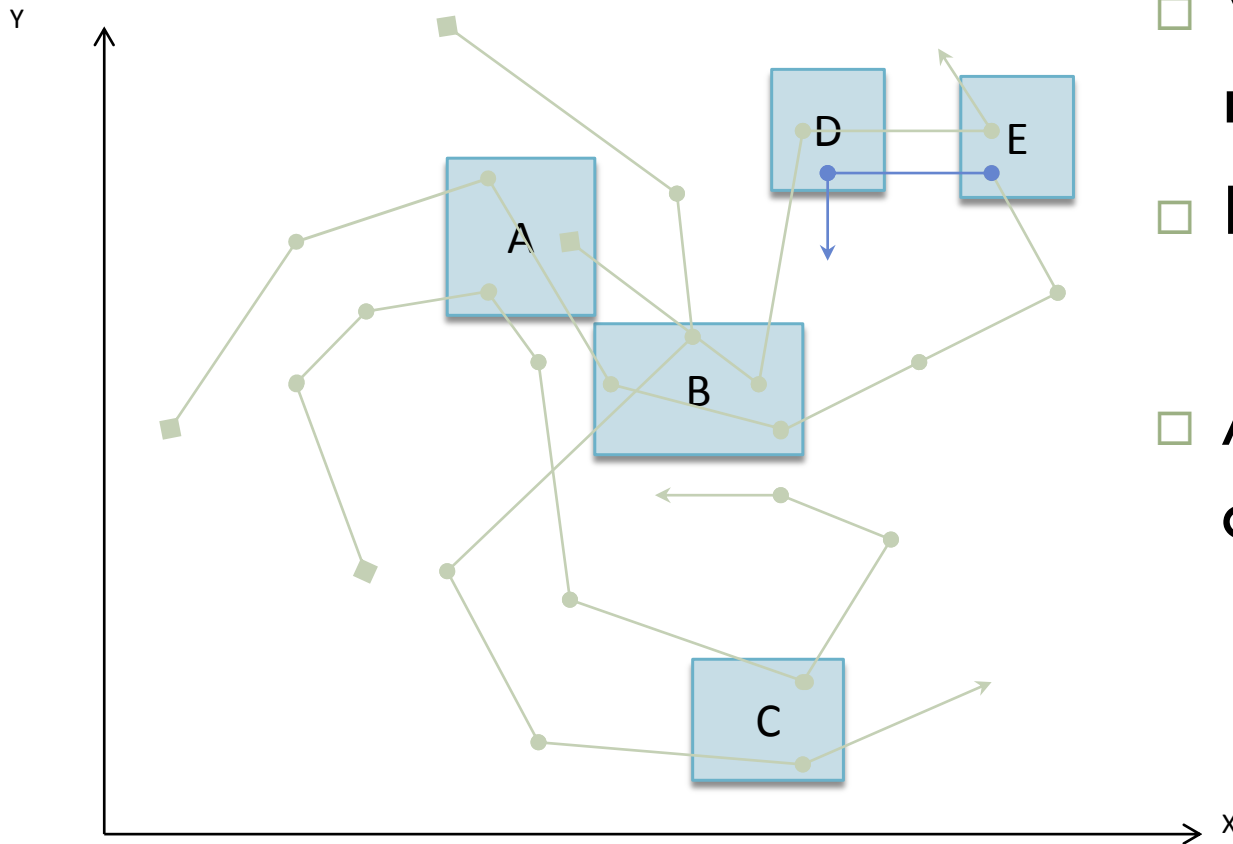
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- With respect to region D.
- Projection
 $D \rightarrow E$
- Add new configuration
 $(T', \langle D \rangle)$

Dynamic Discovery of T-Patterns (7)

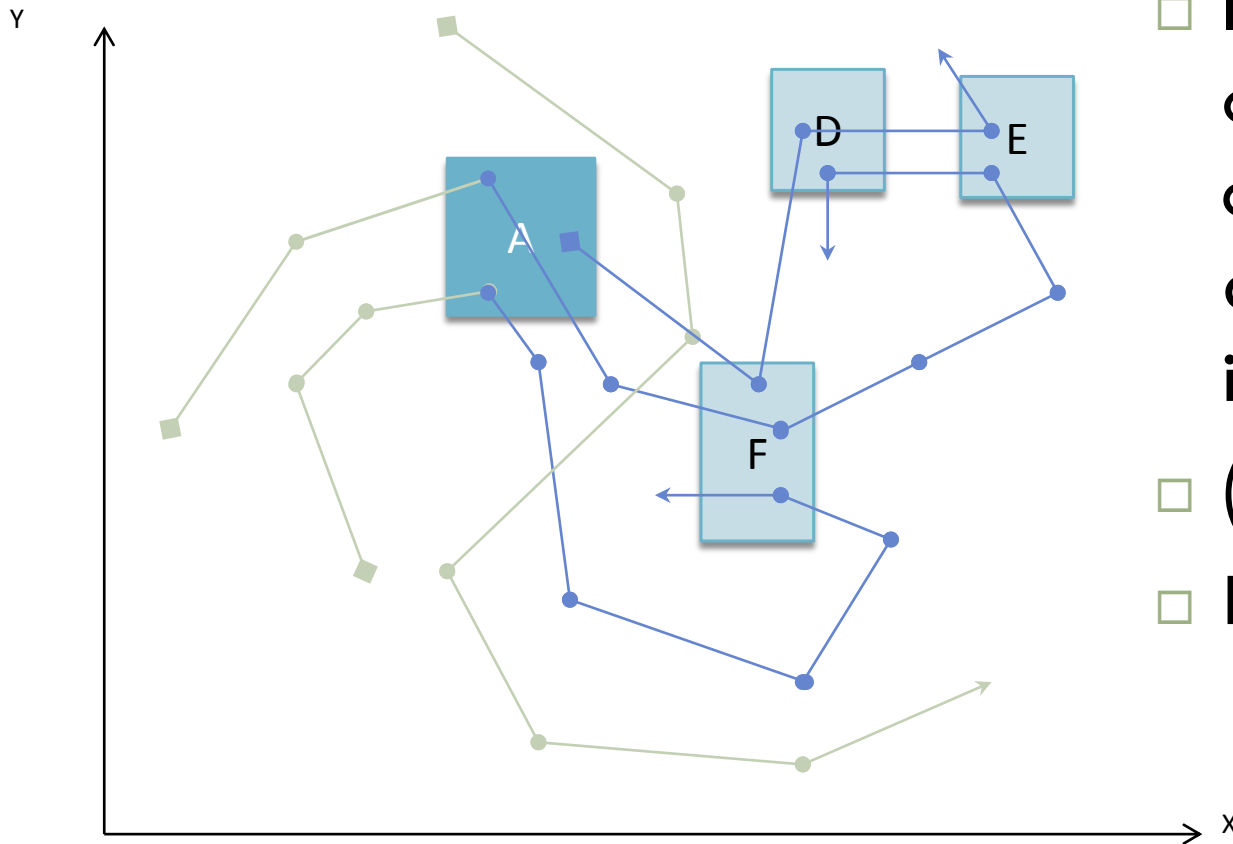
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- With respect to region E.
- Projection
 $E \rightarrow D$
- Add new configuration
 $(T', \langle E \rangle)$

Dynamic Discovery of T-Patterns (8)

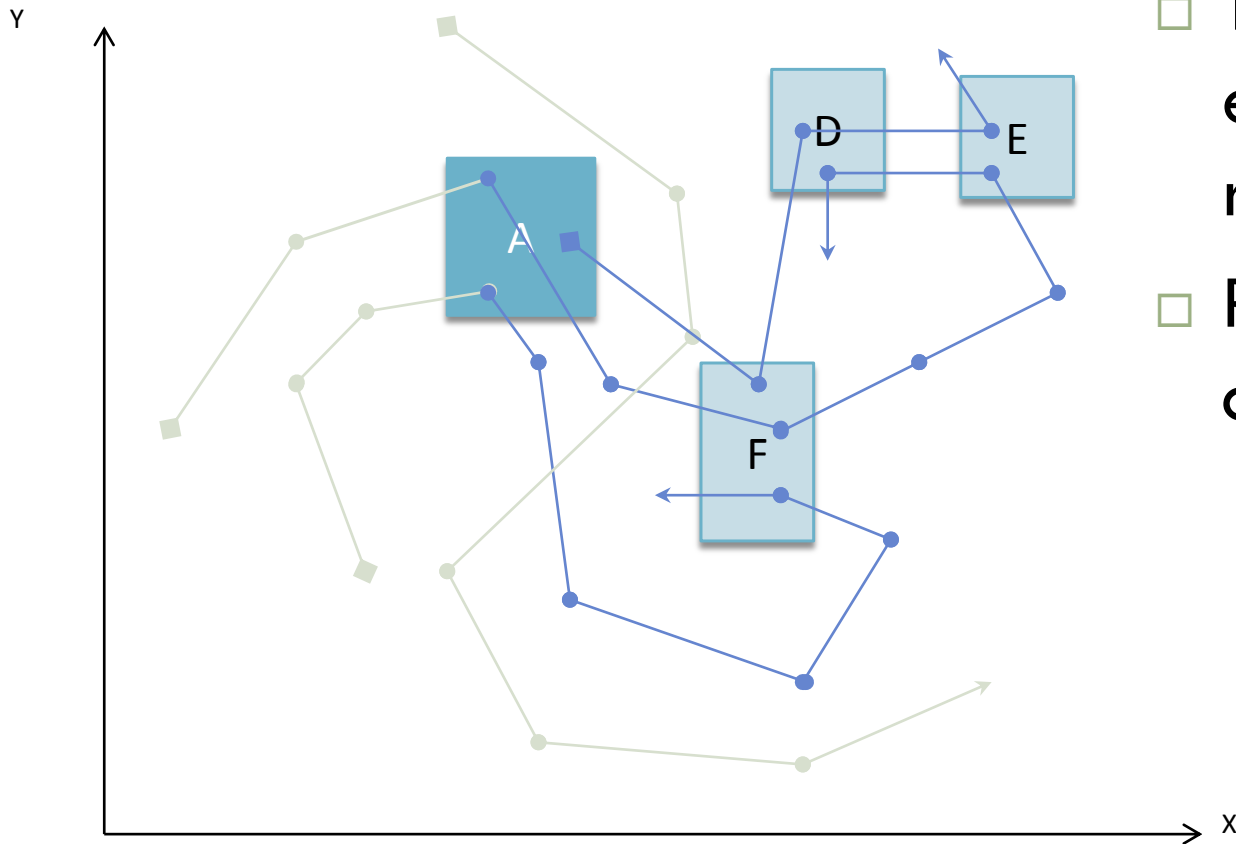
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- Iteration 2 works on each configurations added in prior iteration.
- $(T', \langle A \rangle)$
- Recompute ROIs.

Dynamic Discovery of T-Patterns (9)

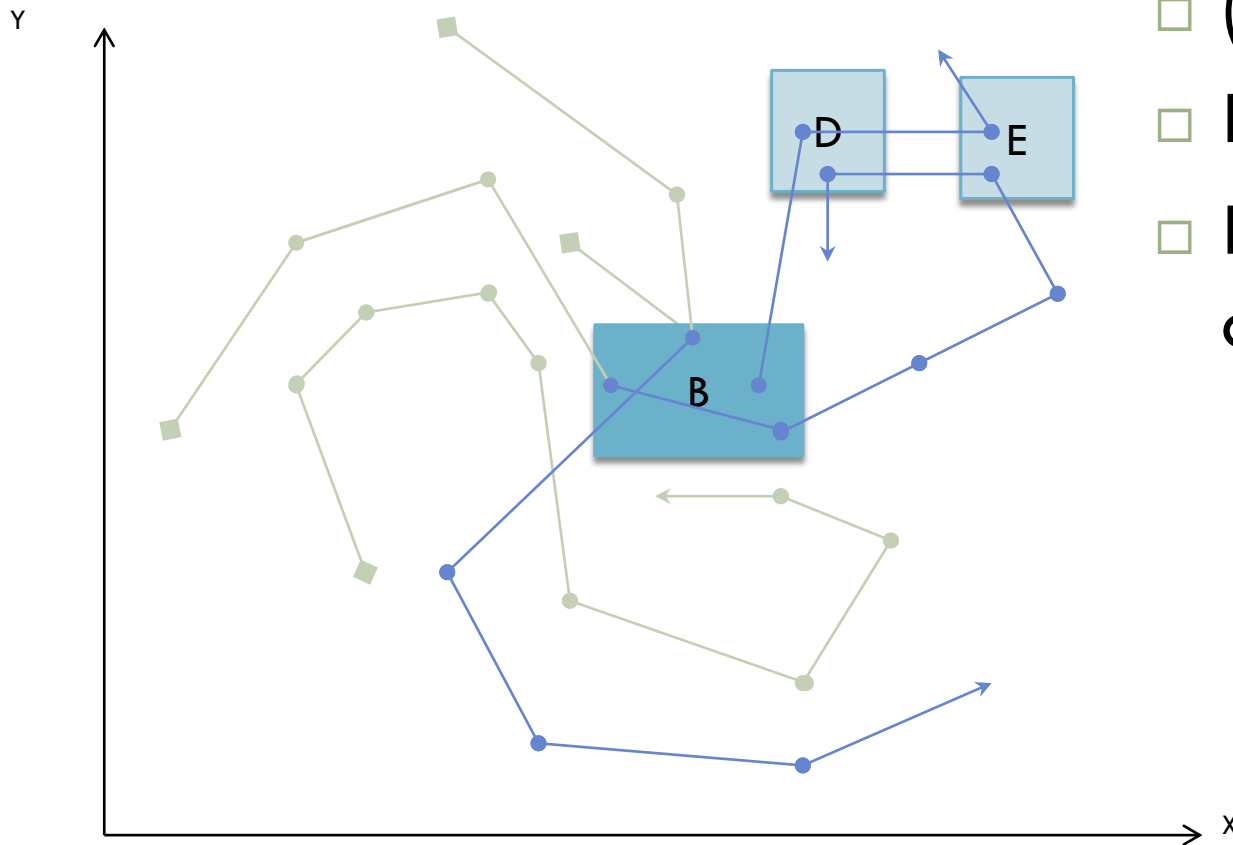
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- Test projects from each of the regions.
- Results in configuration $(T', \langle A \rightarrow F \rangle)$

Dynamic Discovery of T-Patterns (1 1)

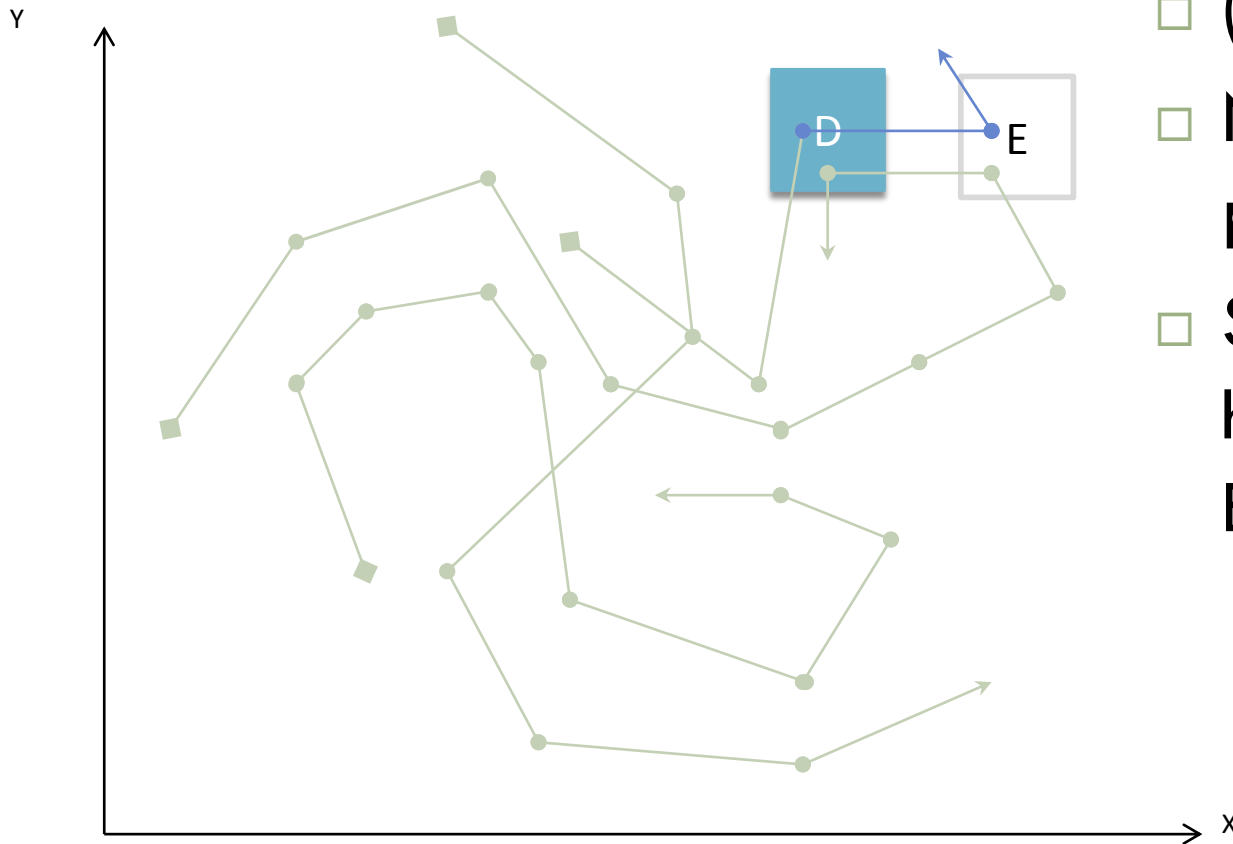
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- $(T', \langle B \rangle)$
- Recompute ROIs.
- Results in configuration
 $(T', \langle B \rightarrow D \rangle)$
 $(T'', \langle B \rightarrow E \rangle)$

Dynamic Discovery of T-Patterns (12)

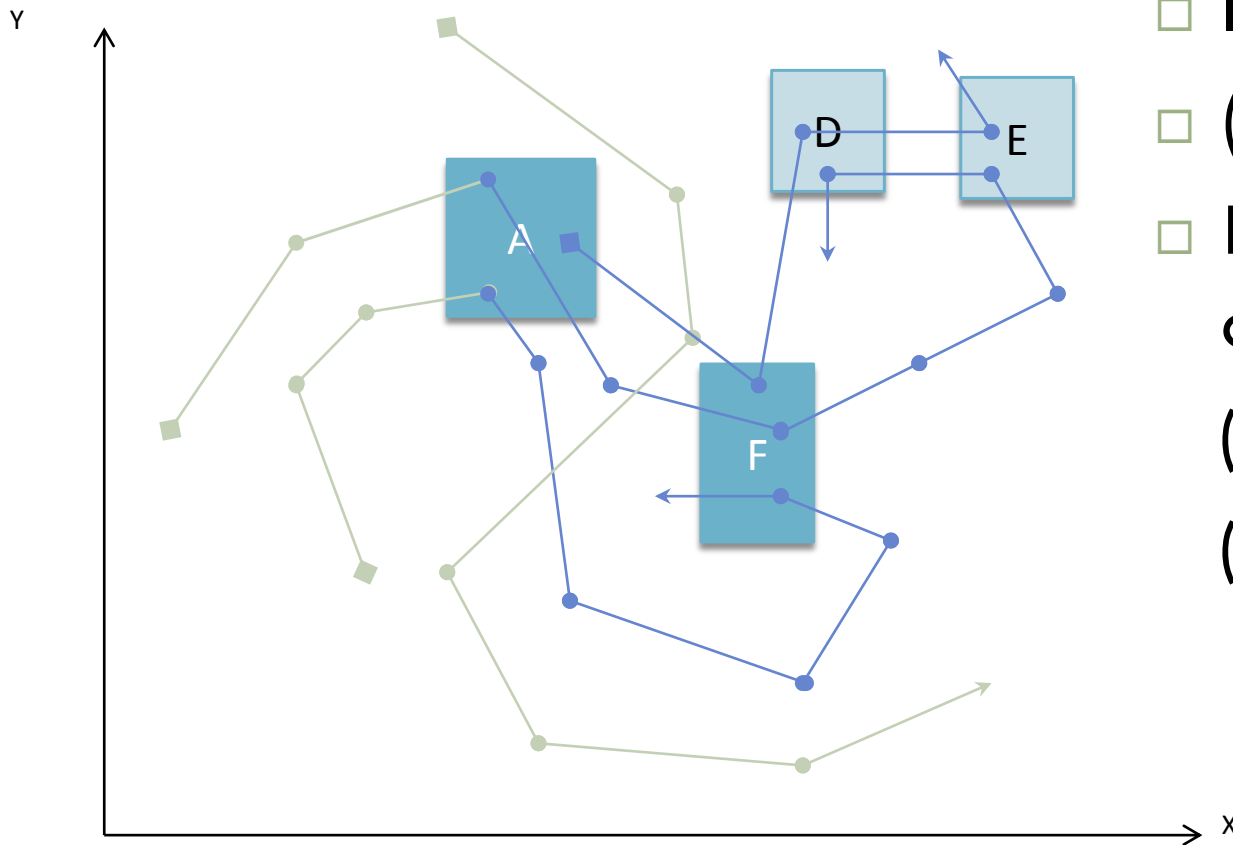
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- $(T', \langle D \rangle)$
- No possible projections.
- Same thing happens for region E.

Dynamic Discovery of T-Patterns (11)

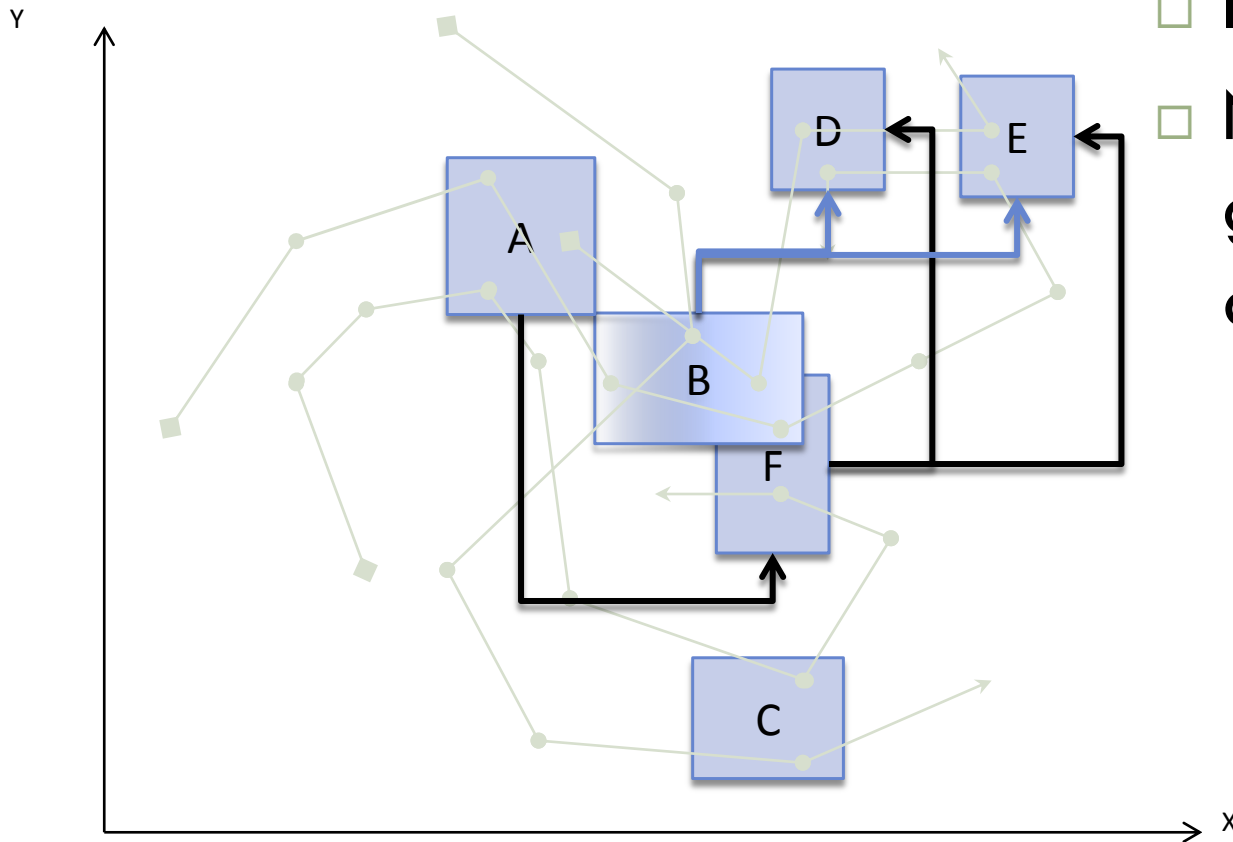
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- Iteration 3
- $(T', \langle A \rightarrow F \rangle)$
- Produces configurations
 $(T', \langle A \rightarrow F \rightarrow D \rangle)$
 $(T', \langle A \rightarrow F \rightarrow E \rangle)$

Dynamic Discovery of T-Patterns (11)

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□ Iteration 4

□ No more paths to go therefore stops and outputs paths:

$A \rightarrow F \rightarrow D$

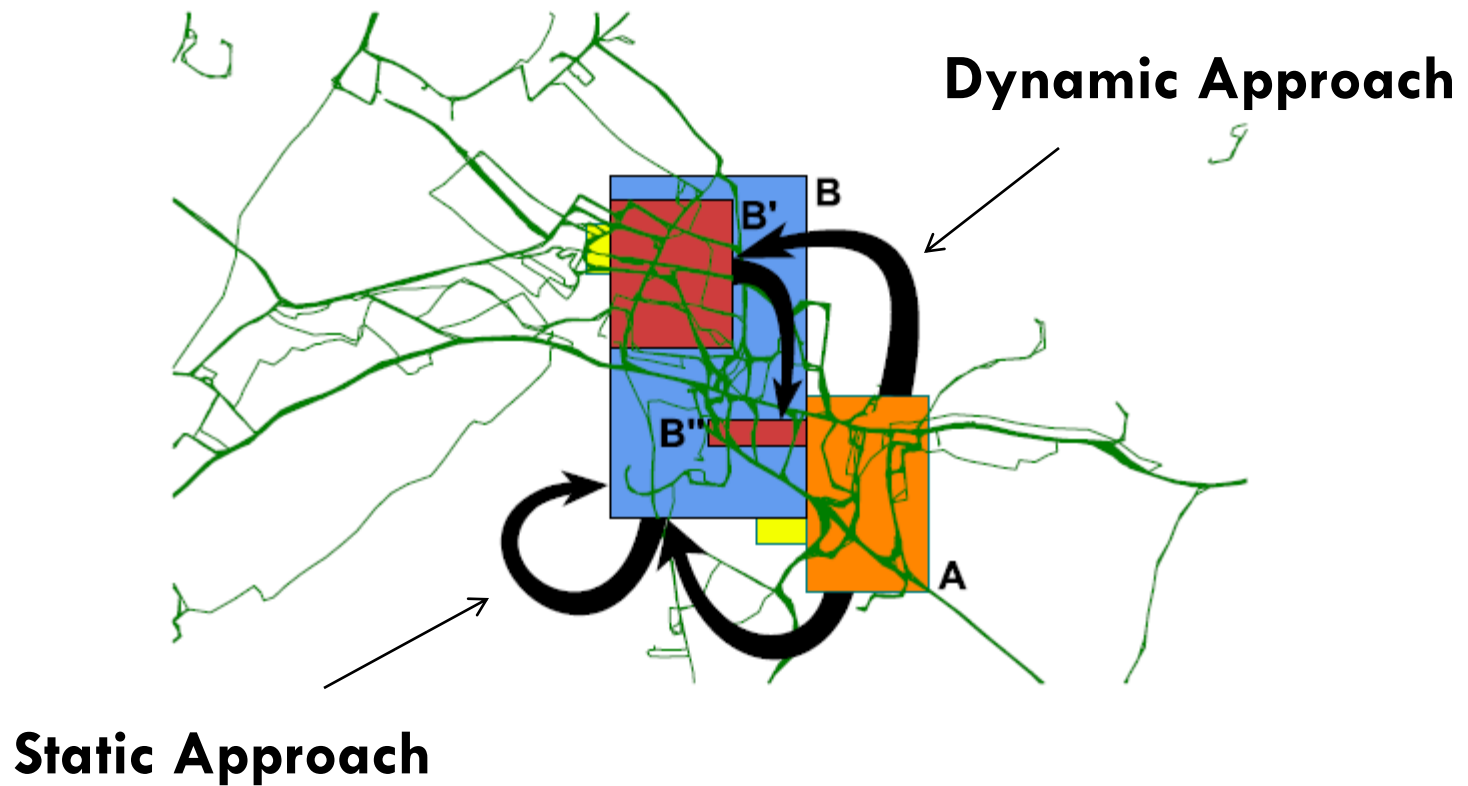
$A \rightarrow F \rightarrow E$

$B \rightarrow D$

$B \rightarrow E$

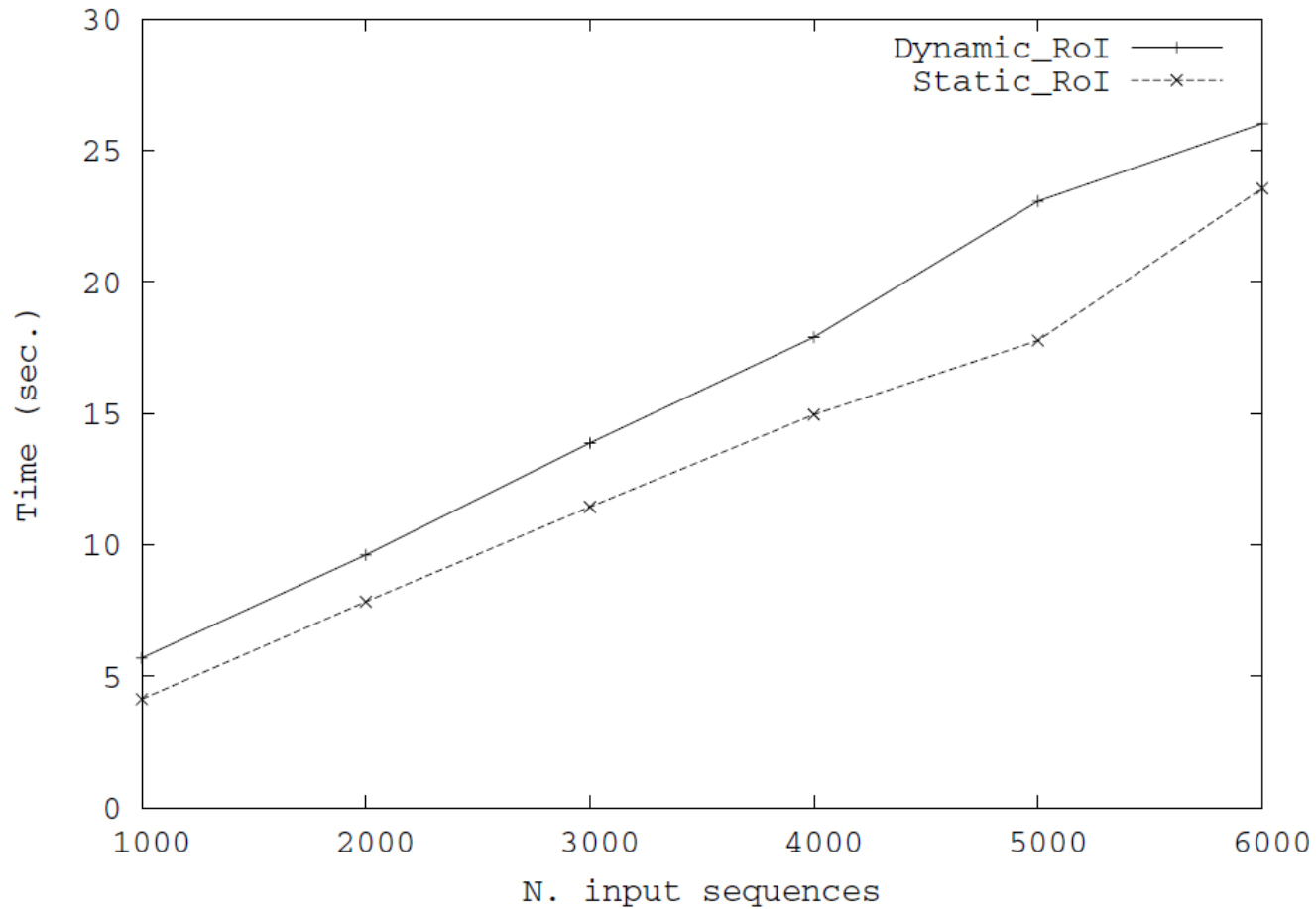
Results

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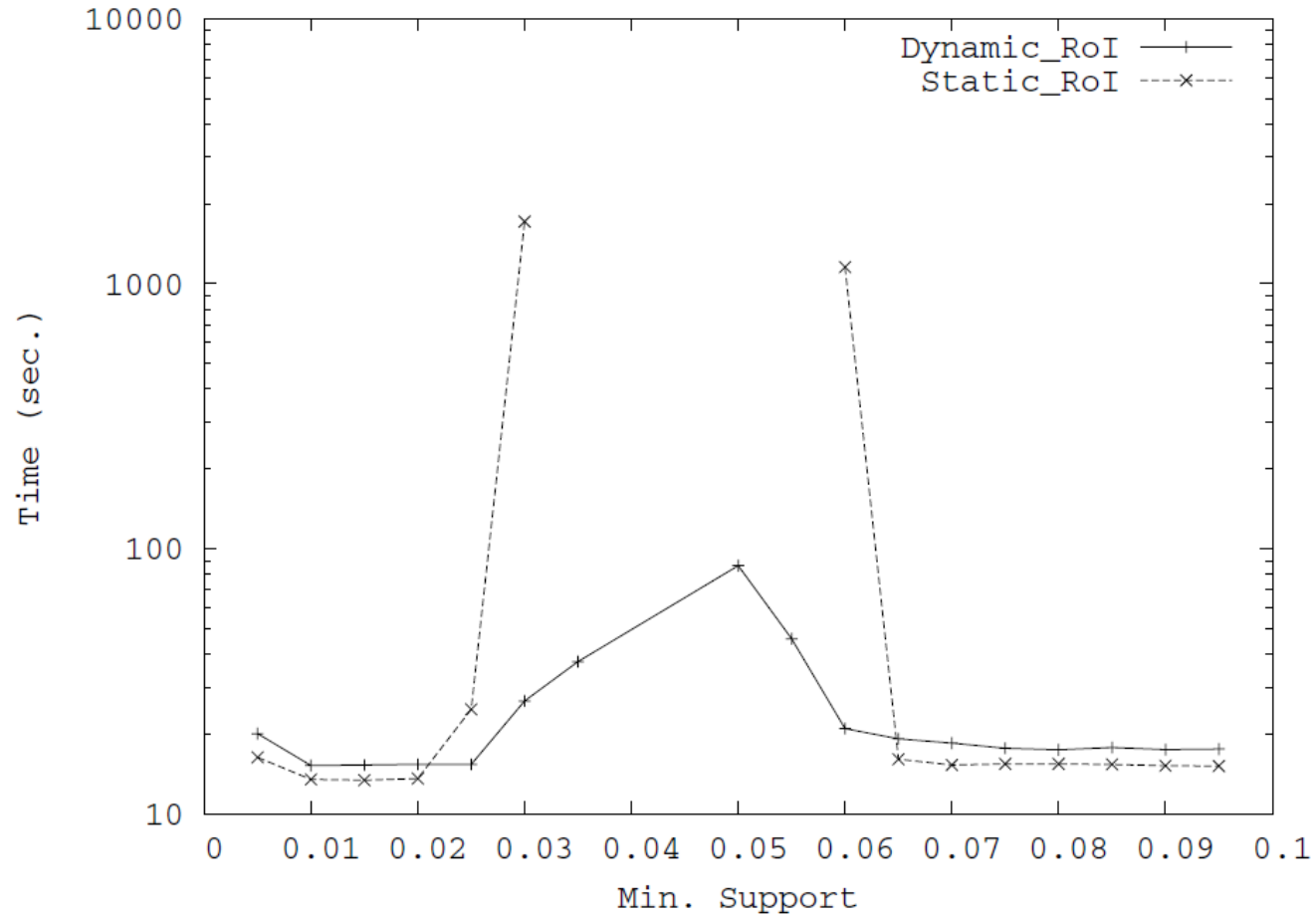
Results

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Results

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To conclude

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- Lack of comparative results
- Idea of ROI is good for applications looking for simple and course results.
- Cited many times for ROI discovery algorithm

- *Where Next*: T-Patterns used to predict where a person will move next, given the current sequence.
- DAEDALUS: Extends SQL to enable TAS queries.

References

- Giannotti, F., Nanni, M., Pinelli, F., and Pedreschi, D. 2007. Trajectory pattern mining. <http://doi.acm.org/10.1145/1281192.1281230>
- F. Giannotti, M. Nanni, and D. Pedreschi. Efficient mining of sequences with temporal annotations <http://www.siam.org/meetings/sdm06/proceedings/032giannottif.pdf>
- Monreale, A., Pinelli, F., Trasarti, R., and Giannotti, F. 2009. WhereNext: a location predictor on trajectory pattern mining <http://doi.acm.org/10.1145/1557019.1557091>
- Ortale, R., Ritacco, E., Pelekis, N., Trasarti, R., Costa, G., Giannotti, F., Manco, G., Renso, C., and Theodoridis, Y. 2008. The DAEDALUS framework: progressive querying and mining of movement data. <http://doi.acm.org/10.1145/1463434.1463497>

Thanks!