

Trajectory Segmentation

Maxime Hauwaert

Université Libre de Bruxelles

30th August of 2022

Outline

1. Introduction
2. Trajectory Segmentation
3. Implementation
4. Experimentation
5. Conclusion



- ▶ Number of GPS powered devices increases
- ▶ Large quantity of data generated
- ▶ Efficient databases which support GPS data required
- ▶ MobilityDB, a PostgreSQL extension



Trajectory Segmentation Definition

Segmentation

A segmentation of a trajectory τ is a collection of subtrajectories which, when put end-to-end, reforms the original trajectory

$$S(\tau) = [\tau_1, \tau_2, \dots, \tau_n]$$

Goals:

- ▶ Meaningful subtrajectories, domain dependent
- ▶ Homogenous points within each subtrajectory
- ▶ Balance between the number of segments and their length

- ▶ Biology:
 - ▶ Study the characteristics of the migrations of certain animals
- ▶ Traffic dynamics:
 - ▶ Study traffic to improve the traffic flow
- ▶ Path planning:
 - ▶ Determine the best transportation mode to take
- ▶ Tourism:
 - ▶ Identify the most visited places by tourists

1. Introduction

2. Trajectory Segmentation

Stable Criteria Segmentation

Sliding Window Segmentation

3. Implementation

4. Experimentation

5. Conclusion



Types of Algorithms

Criteria based

- ▶ Characteristics of subtrajectories explicitly stated
- ▶ Great knowledge required
- ▶ Lots of possibilities

Model based

- ▶ Hidden underlying criteria
- ▶ Only a few parameters to set
- ▶ Easier to use

Stable Criteria Segmentation

Key points:

- ▶ Stable criteria
- ▶ Combinations
- ▶ Trajectory \rightarrow Compressed start-stop matrix \rightarrow Segmentation tree \rightarrow Segmentation
- ▶ Parameter:
 - ▶ Propositional formula
- ▶ Complexity $O(n \log n)$
- ▶ Competitors: smaller number of criteria supported, $O(n^2)$

Types of Criteria

Decreasing monotone criteria

If the criterion is met for a trajectory, it is also met for every subtrajectory

Example: speed range

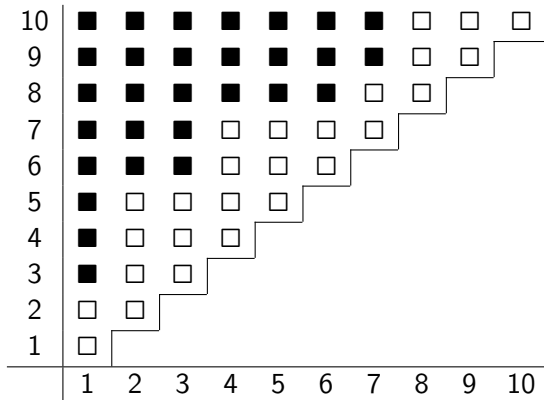
Increasing monotone criteria

If the criterion is met for a trajectory, it is also met for every supertrajectory

Example: time

⚠ An increasing monotone criterion is useless on its own, but can be combined with a decreasing monotone criterion

Start-Stop Matrix

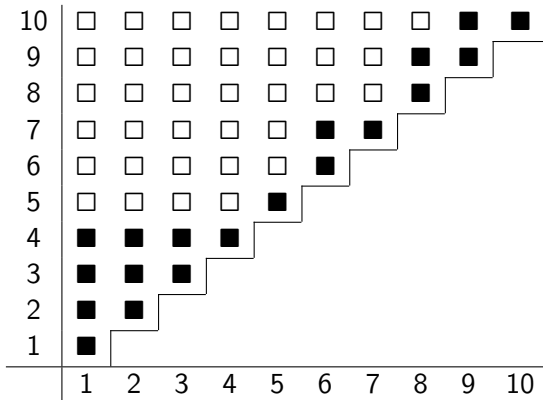


Example of a start-stop matrix obtained with a decreasing monotone criterion

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Start-Stop Matrix

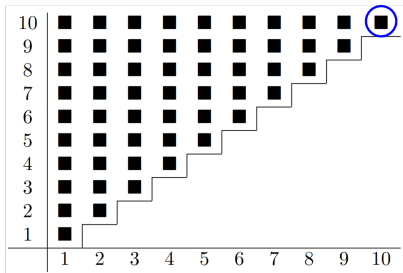


Example of a start-stop matrix obtained with an increasing monotone criterion

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation



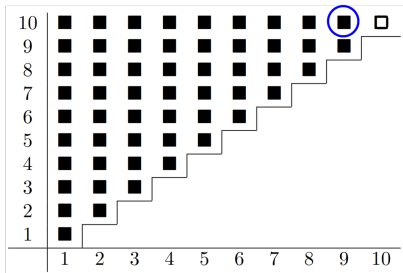
For decreasing monotone criteria:

- ▶ Start by placing a pointer at the top right of the matrix

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation

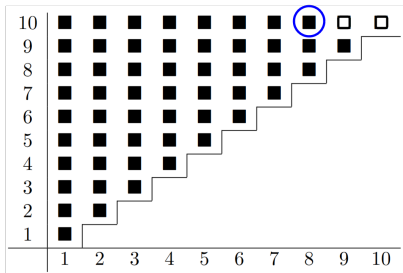


- ▶ While the criterion is met for $\tau_{j,i} \rightarrow$ set i,j to true and decrement j

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation

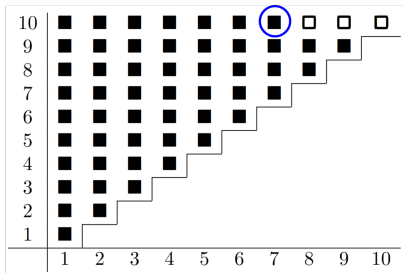


- ▶ While the criterion is met for $\tau_{j,i} \rightarrow$ set i,j to true and decrement j

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation

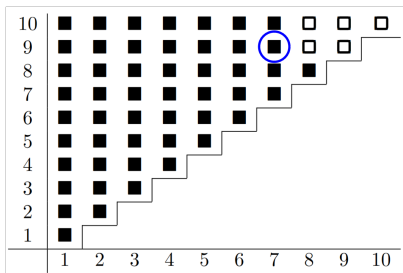


- ▶ If the criterion is not met for $\tau_{j,i} \rightarrow$ decrement i , set every element at the right of i,j to true and continue the execution

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation

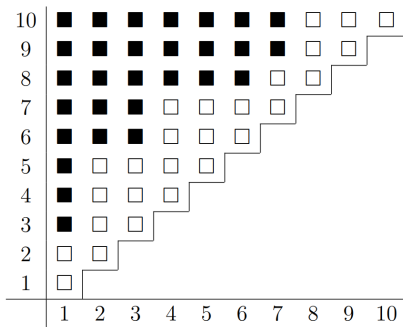


- If the criterion is not met for $\tau_{j,i} \rightarrow$ decrement i , set every element at the right of i,j to true and continue the execution

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation



Resulting start stop matrix

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Compressed Start-Stop Matrix Generation

For increasing monotone criteria:

Decreasing \leftrightarrow Increasing

- ▶ Negation of increasing criterion = decreasing criterion
- ▶ Negation of decreasing criterion = increasing criterion

→ Negate the criteria and then negating the resulting start-stop matrix

Run length encoding $\rightarrow O(n \log n)$ space complexity

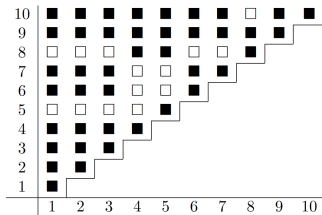


Segmentation Tree Generation

Optimal segmentation of $\tau_{0,i}$:

- ▶ Either $\tau_{0,i}$
- ▶ Or optimal segmentation of $\tau_{0,j} + \tau_{j,i}$

Segmentation Tree Generation

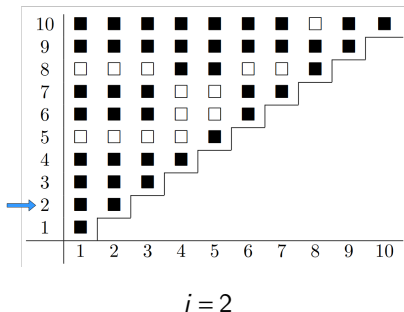


- ▶ Iteration on each row of the start-stop matrix
- ▶ Keeping a list of tuples:
 - ▶ index: the index of a trajectory point
 - ▶ count: the number of subtrajectories required to get to the current point
 - ▶ last: the last segmentation point
- ▶ Adding tuple (index = 1, count = 0, last = N/A)
- ▶ Starting at row 2

2. Trajectory Segmentation

a) Stable Criteria Segmentation

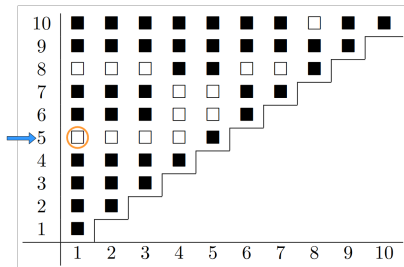
Segmentation Tree Generation



- ▶ No true value at row 2, thus 2 cannot be a segmentation point



Segmentation Tree Generation



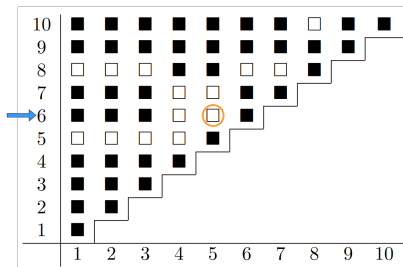
$i = 5$

- ▶ There are some true values, thus 5 might be a segmentation point
- ▶ Point 5 can be reached from point 1, 2, 3 and 4
- ▶ Select the element with the minimum count.
- ▶ Add (index = 5, count = 1, last = 1) to the list

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Segmentation Tree Generation

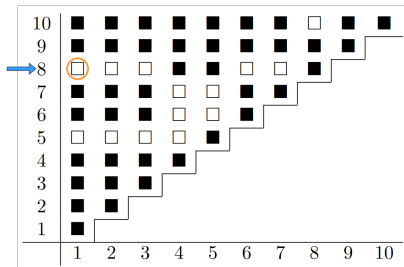


$i = 6$

- ▶ Point 6 can be reached from point 4 and 5
- ▶ Select the minimum count
- ▶ Add (index = 6, count = 2, last = 5) to the list



Segmentation Tree Generation



$$i = 8$$

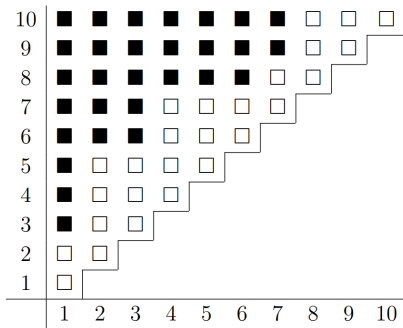
To speed up:

- ▶ Store the tuples in a search binary tree sorted on the index
- ▶ Each node points the node with the lowest count in its subtree
- ▶ For each block of consecutive true values, search only the first and last index

2. Trajectory Segmentation

a) Stable Criteria Segmentation

Segmentation Tree Generation

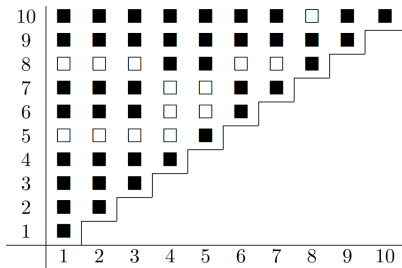


To retrieve the segmentation:

- ▶ Take the tuple with the highest index
- ▶ Loop on the last field until the end is reached
- ▶ Outputting the index at each iteration



Segmentation Tree Generation



$\{(5,1,1), (6,2,5), (7,2,5), (8,1,1), (10,2,8)\}$

$\text{Segmentation}(\tau) = [\tau_{1,8}, \tau_{8,10}]$



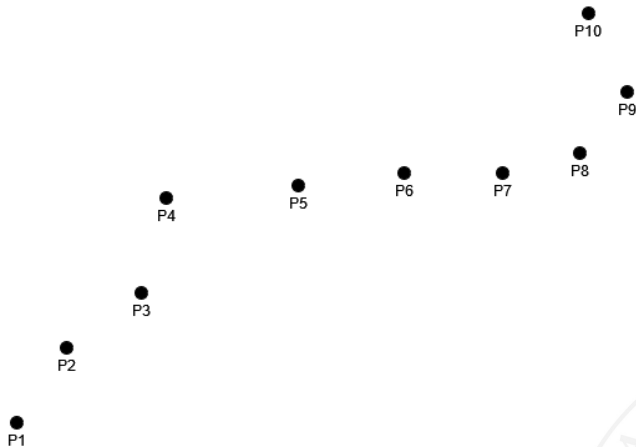
Sliding Window Segmentation

Key points:

- ▶ Sliding window
- ▶ Interpolation
- ▶ Trajectory → Error signal → Segmentation
- ▶ Parameters:
 - ▶ Window length
 - ▶ Kernel
 - ▶ Threshold / Percentile
- ▶ Complexity $O(n)$ / $O(n \log n)$
- ▶ Competitors: smaller harmonic mean of purity and coverage metrics



Error Signal Generation

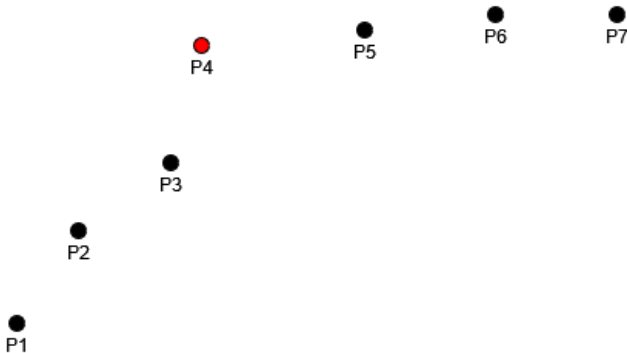


Base trajectory

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

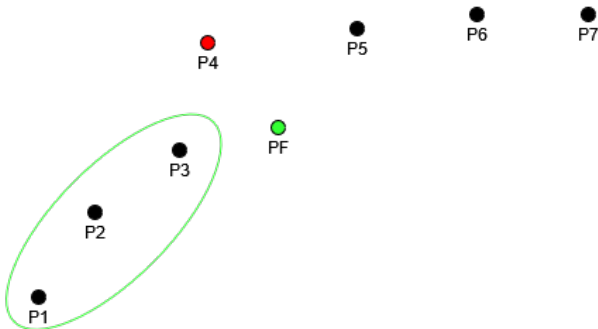


Window creation (length 7)

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

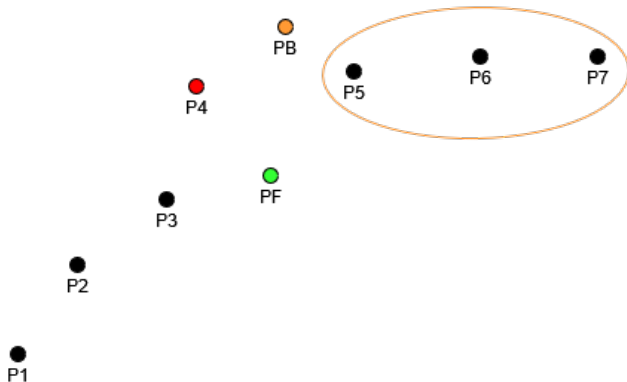


Forward extrapolation

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

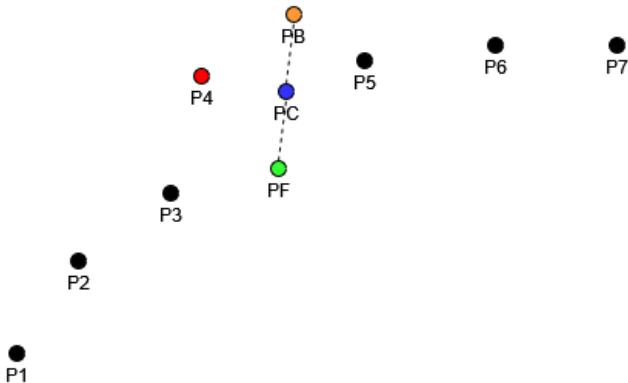


Backward extrapolation

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

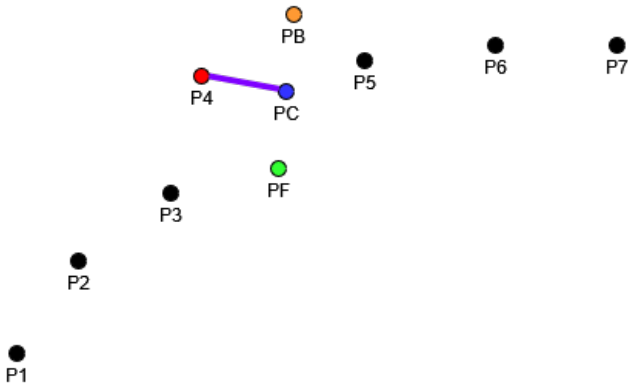


Middle point computation

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

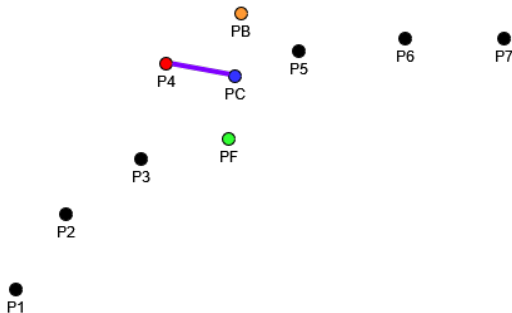


Error value computation

2. Trajectory Segmentation

b) Sliding Window Segmentation

Error Signal Generation

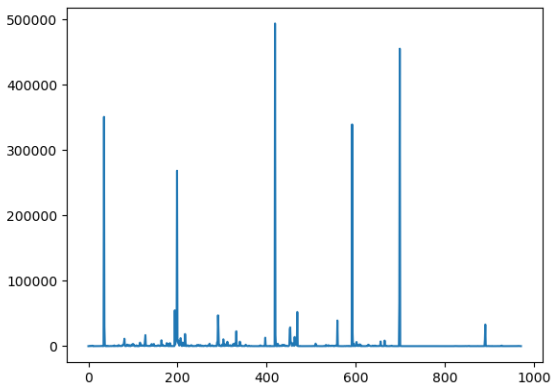


Error value computation

- First and last $\frac{\text{window length}}{2}$ points impossible to extrapolate
→ error value = 0

2. Trajectory Segmentation

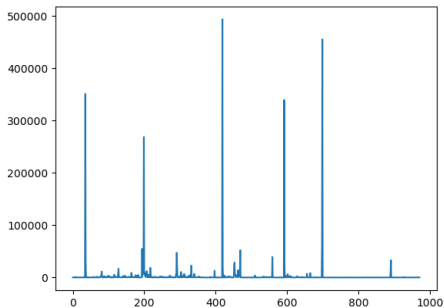
b) Sliding Window Segmentation



Error signal

2. Trajectory Segmentation

b) Sliding Window Segmentation

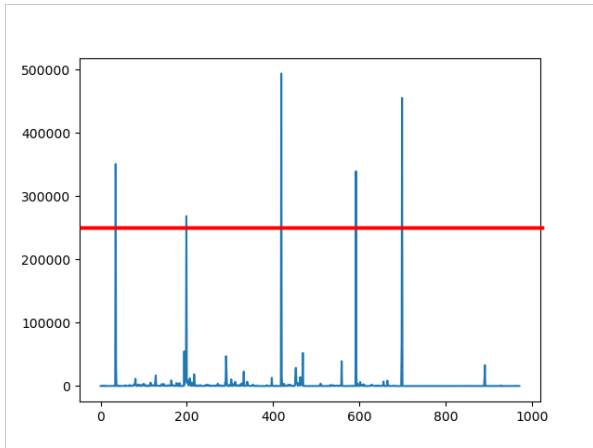


Error signal

For the percentile variant: set threshold t such that $x\%$ error values $< t$

2. Trajectory Segmentation

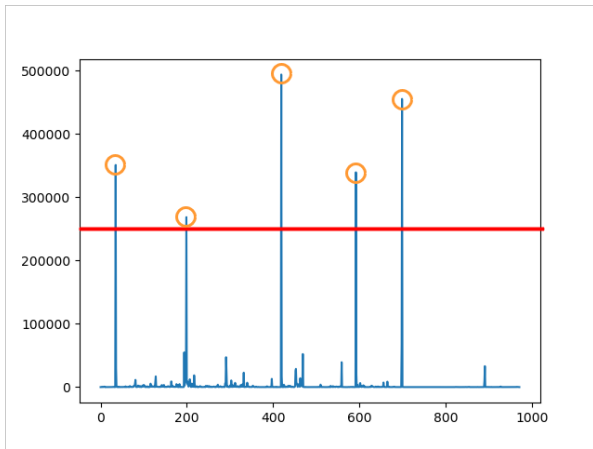
b) Sliding Window Segmentation



Threshold = 250000

2. Trajectory Segmentation

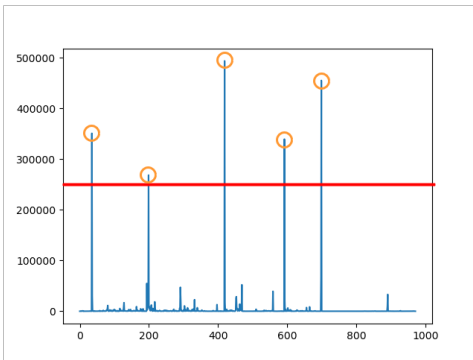
b) Sliding Window Segmentation



Segmentation points

2. Trajectory Segmentation

b) Sliding Window Segmentation



Segmentation points

$$\text{Segmentation}(\tau) = [\tau_{0,25}, \tau_{25,200}, \tau_{200,410}, \tau_{410,590}, \tau_{590,700}, \tau_{700,980}]$$

2. Trajectory Segmentation

b) Sliding Window Segmentation

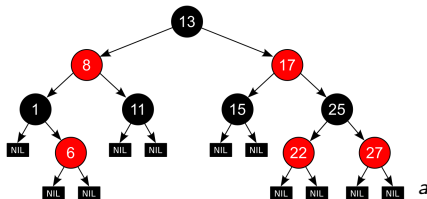
Outline

1. Introduction
2. Trajectory Segmentation
3. Implementation
4. Experimentation
5. Conclusion



Red-Black trees

- ▶ Binary search tree
- ▶ Complexity $\log n$ search, insert, delete
- ▶ Augmentation

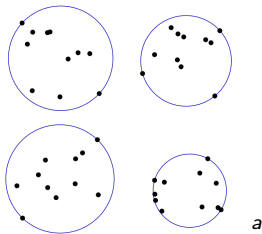


^aBy Nomen4Omen - Own work, CC BY-SA 4.0,
<https://commons.wikimedia.org/w/index.php?curid=83966511>

3. Implementation

Welzl's algorithm

- ▶ Smallest enclosing disk
- ▶ Randomized procedure
- ▶ Linear expected time complexity



^aBy Claudio Rocchini - Own work, CC BY-SA 3.0,
<https://commons.wikimedia.org/w/index.php?curid=12562519>

3. Implementation

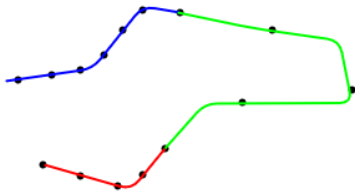
Stable Criteria Segmentation

Speed range criterion

For each subtrajectory, the speed of the object should not vary more than the specified amount



Before



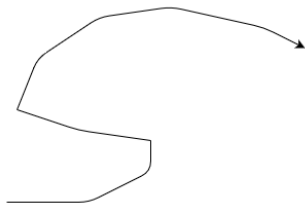
After

3. Implementation

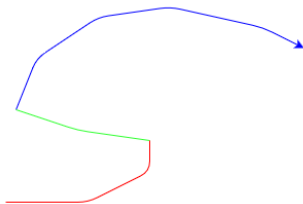
Stable Criteria Segmentation

Angle criterion

Within each subtrajectory, the direction the object is heading does not change by more than the specified angle between two consecutive points



Before



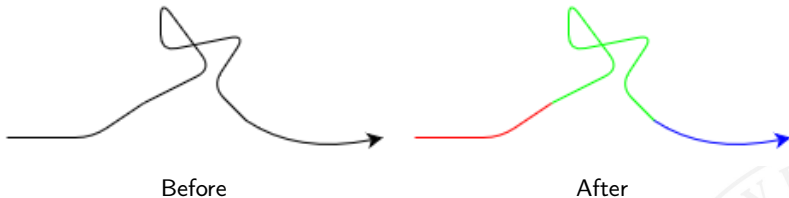
After

3. Implementation

Stable Criteria Segmentation

Disk criterion

For each subtrajectory, there exists a disk of the specified perimeter which covers all of their points



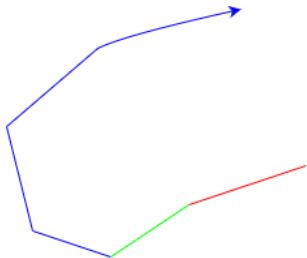
3. Implementation

Stable Criteria Segmentation

Time criterion

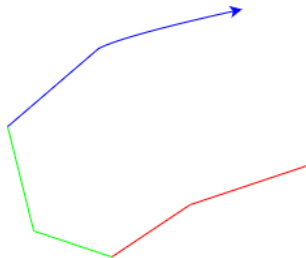
The subtrajectories should last at least the specified amount of time

⚠ Needs to be combined with a decreasing monotone criterion



Before

3. Implementation



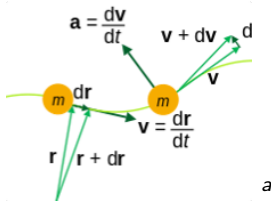
After

Sliding Window Segmentation

Kinematic:

$$x_i = \frac{1}{2} a_x(2) (t_i - t_2)^2 + v_x(2) (t_i - t_2) + x_2$$

$$v_x(a) = \frac{x_a - x_{a-1}}{t_a - t_{a-1}}, \quad a_x(a) = \frac{v_x(a) - v_x(a-1)}{t_a - t_{a-1}}$$



^aBy Maschen - Own work, CC0, <https://commons.wikimedia.org/w/index.php?curid=21275429>

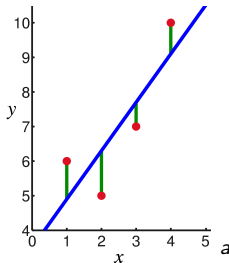
3. Implementation

Sliding Window Segmentation

Linear regression:

- Ordinary least squares

$$y = mx + p, \quad m = \frac{\Sigma((x - \bar{x})(y - \bar{y}))}{\Sigma((x - \bar{x})^2)}, \quad p = \bar{y} - m\bar{x}$$



^aBy Oleg Alexandrov - self-made with MATLAB, Public Domain,
<https://commons.wikimedia.org/w/index.php?curid=4099808>

3. Implementation

Outline

1. Introduction
2. Trajectory Segmentation
3. Implementation
4. Experimentation
5. Conclusion



Queries:

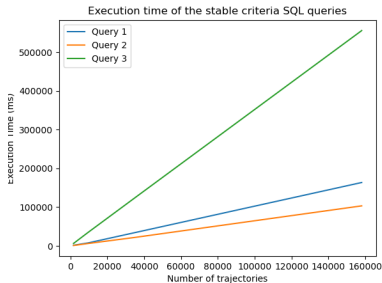
▶ Stable criteria segmentation:

- 1) `select segment_by_stable_criteria(trip, '(s[20]))' from trips;`
- 2) `select segment_by_stable_criteria(trip, '(a[90]))' from trips;`
- 3) `select segment_by_stable_criteria(trip, '(d[100]))' from trips;`

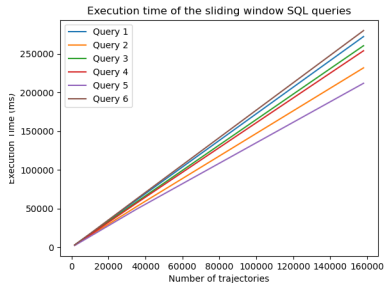
▶ Sliding window segmentation:

- 1) `select segment_by_sws_t(trip, 400000, 7, 'K') from trips;`
- 2) `select segment_by_sws_t(trip, 400000, 5, 'L') from trips;`
- 3) `select segment_by_sws_p(trip, 85, 7, 'K') from trips;`
- 4) `select segment_by_sws_p(trip, 95, 7, 'K') from trips;`
- 5) `select segment_by_sws_p(trip, 95, 5, 'L') from trips;`
- 6) `select segment_by_sws_p(trip, 95, 9, 'L') from trips;`

Dataset: MobilityDB Berlin mod with four scale factors (0.005,0.05,0.2,1)
Hardware: Windows 11 64 bits, Intel®Core™ i7-7700HQ at 2.8 GHz, 16 GB RAM
Software: PostgreSQL 13, PostGIS 3 and MobilityDB develop



Stable Criteria Segmentation

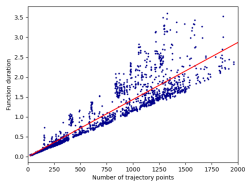


Sliding Window Segmentation

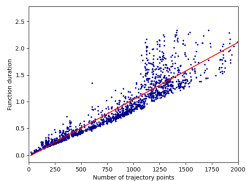
4. Experimentation

Stable criteria segmentation:

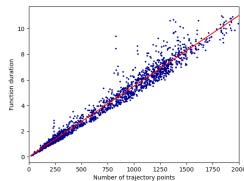
Function duration(y) in relation to the number of trajectory points(x)



Speed range



Angle

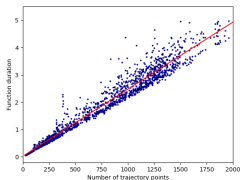


Disk

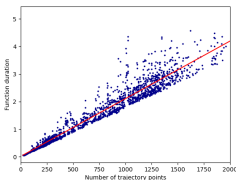
4. Experimentation

Sliding window segmentation:

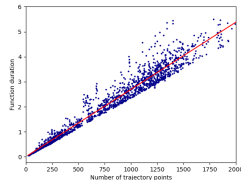
Function duration(y) in relation to the number of trajectory points(x)



Percentile kinematic

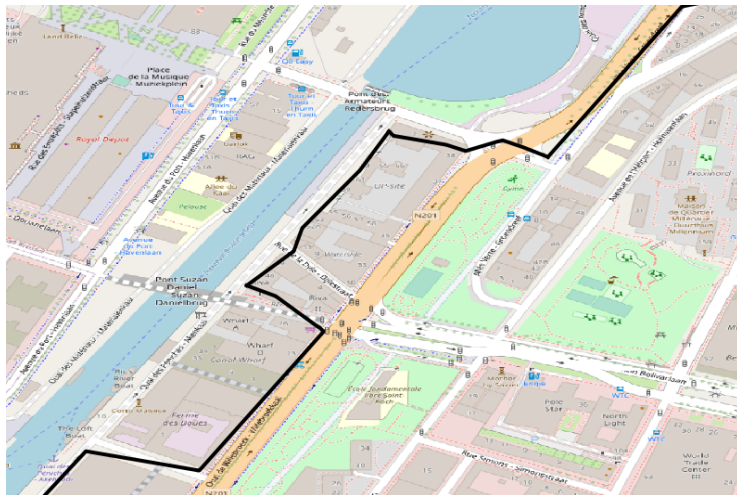


Percentile linear



Threshold kinematic

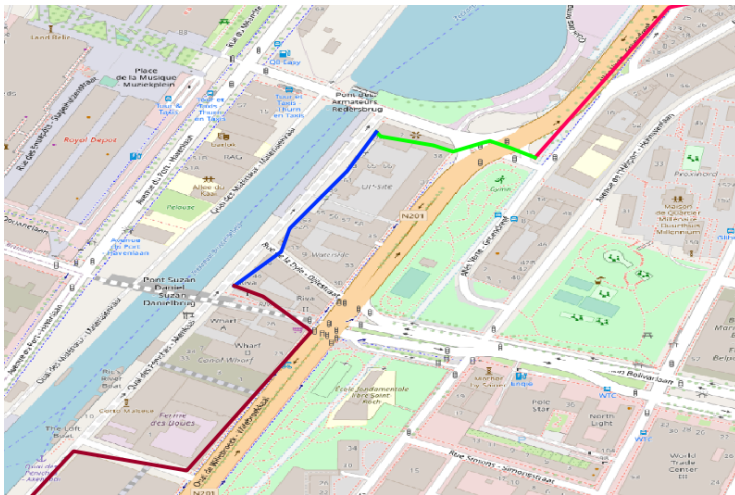
4. Experimentation



Base trajectory

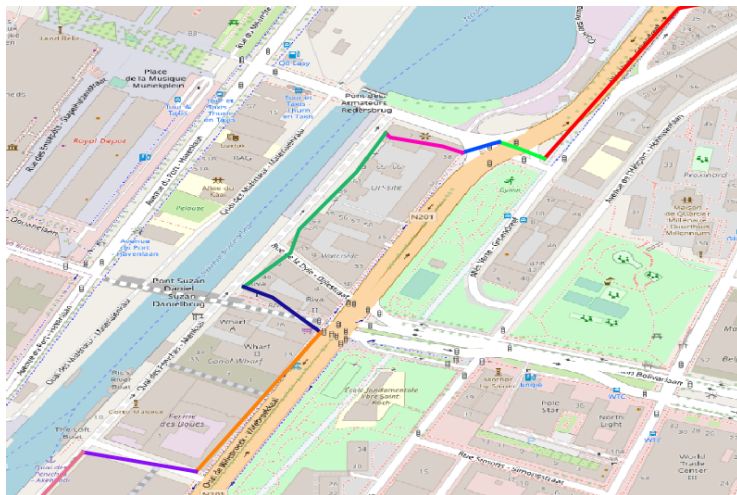
4. Experimentation

Maxime Hauwaert



Angle criterion 90°

4. Experimentation



Outline

1. Introduction
2. Trajectory Segmentation
3. Implementation
4. Experimentation
5. Conclusion



- ▶ A wide state-of-the-art about the trajectory segmentation
- ▶ An in-depth analysis of two trajectory segmentation algorithms
- ▶ An implementation in MobilityDB of these two algorithms
- ▶ An analysis of the performance of the implementation

- ▶ Improve implemented algorithms:
 - ▶ Stable criteria segmentation:
 - ▶ Add outlier tolerance
 - ▶ Add angle range criterion
 - ▶ Sliding window segmentation:
 - ▶ Add random walk and cubic spline kernel
- ▶ Use external data:
 - ▶ Heading, speed from sensors
 - ▶ Road and public transportation network
- ▶ Add temporal sequence set support

- [1] Maïke Buchin et al. "Segmenting trajectories: A framework and algorithms using spatiotemporal criteria". In: *Journal of Spatial Information Science* 2011.3 (2011), pp. 33–63.
- [2] Sander PA Alewijnse et al. "A framework for trajectory segmentation by stable criteria". In: *Proceedings of the 22nd ACM SIGSPATIAL international conference on advances in geographic information systems*. 2014, pp. 351–360.
- [3] Sander PA Alewijnse. "A framework for trajectory segmentation by stable criteria and Brownian bridge movement model". PhD thesis. Master's thesis: Eindhoven University of Technology, 2013.
- [4] Boris Aronov et al. "Segmentation of trajectories on nonmonotone criteria". In: *ACM Transactions on Algorithms (TALG)* 12.2 (2015), pp. 1–28.
- [5] Mohammad Etemad et al. "SWS: an unsupervised trajectory segmentation algorithm based on change detection with interpolation kernels". In: *Geoinformatica* (2020), pp. 1–21.