

2IL76

Algorithms for Geographic Data

Spring 2015

Lecture 6: Segmentation

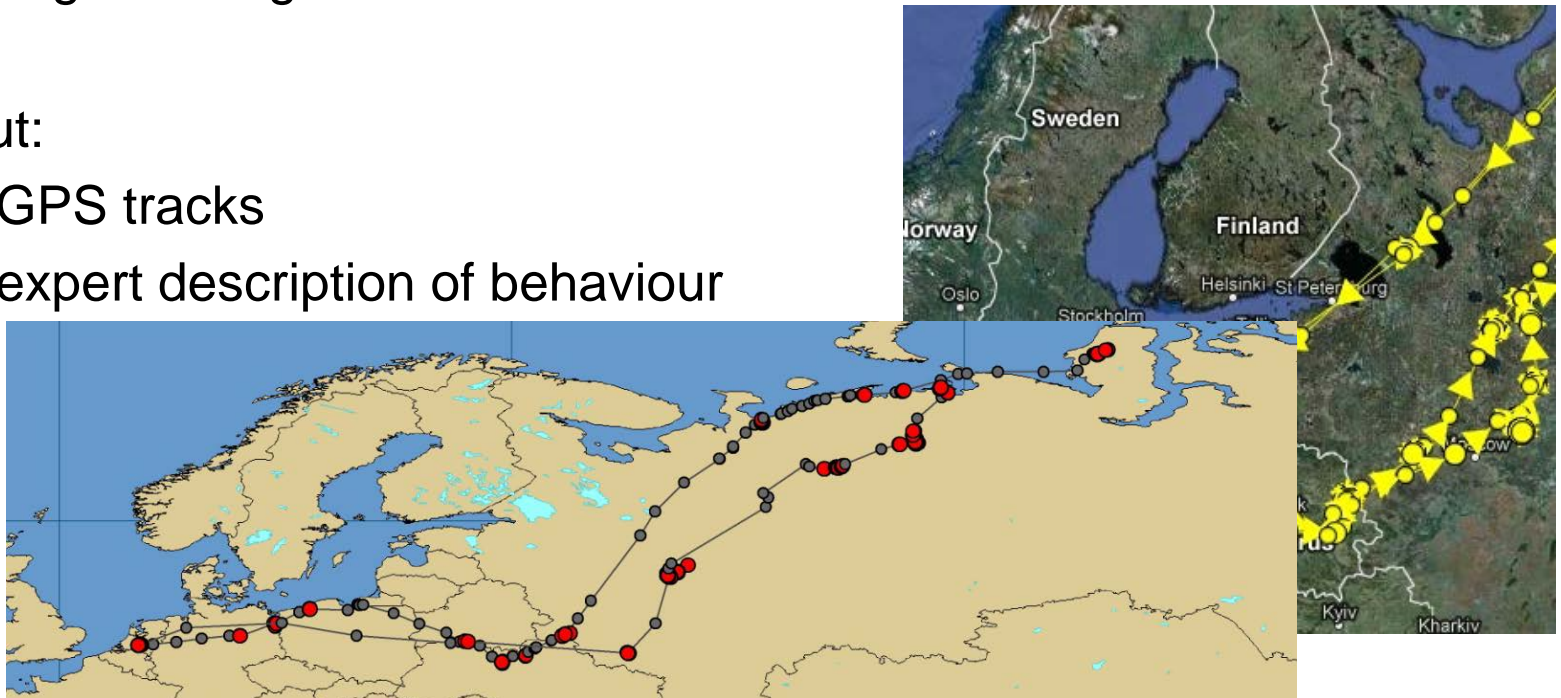
Motivation: Geese Migration

- Two behavioural types

- stopover
- migration flight

- Input:

- GPS tracks
- expert description of behaviour



- Goal: Delineate stopover sites of migratory geese

Abstract / general purpose questions

Single trajectory

- simplification, cleaning
- *segmentation into semantically meaningful parts*
- finding recurring patterns (repeated subtrajectories)

Two trajectories

- similarity computation
- subtrajectory similarity

Multiple trajectories

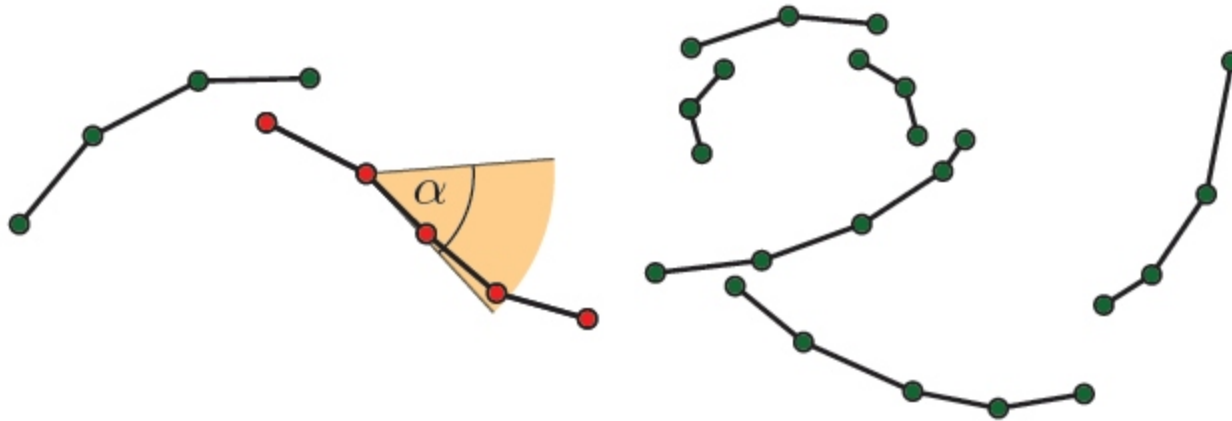
- clustering, outliers
- flocking/grouping pattern detection
- finding a typical trajectory or computing a mean/median trajectory
- visualization

Problem

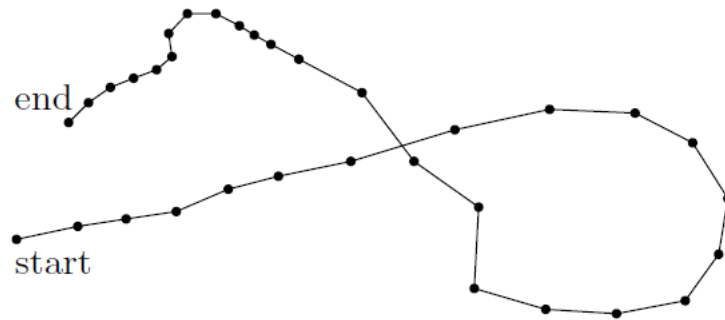
- ❑ For analysis, it is often necessary to break a trajectory into pieces according to the behaviour of the entity (e.g., walking, flying, ...).
- ❑ **Input:** A trajectory T , where each point has a set of attribute values, and a set of criteria.
- ❑ **Attributes:** speed, heading, curvature...
- ❑ **Criteria:** bounded variance in speed, curvature, direction, distance...
- ❑ **Aim:** Partition T into a minimum number of subtrajectories (so-called *segments*) such that each segment fulfils the criteria.
- ❑ *“Within each segment the points have similar attribute values”*

Criteria-Based Segmentation

Goal: Partition trajectory into a small number of segments such that a given criterion is fulfilled on each segment

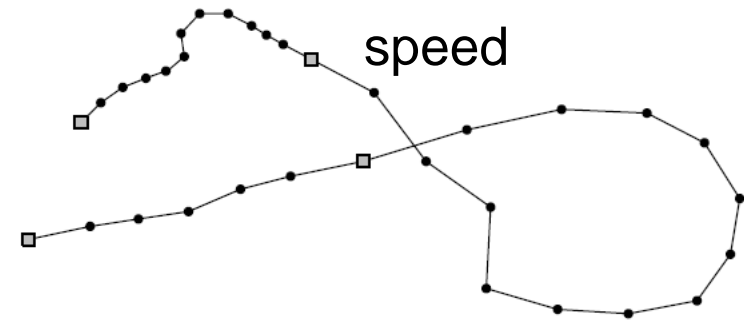
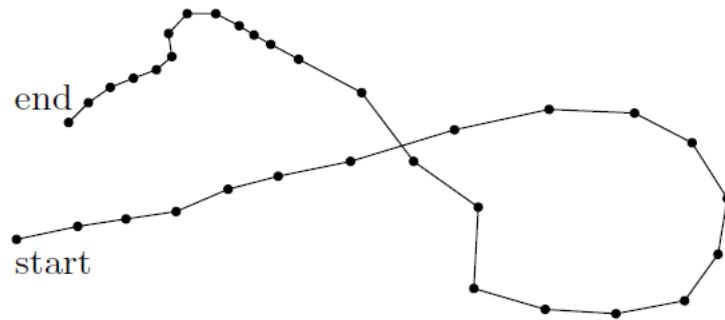


Example



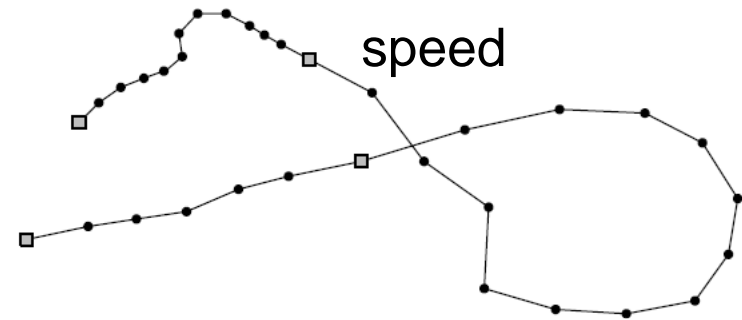
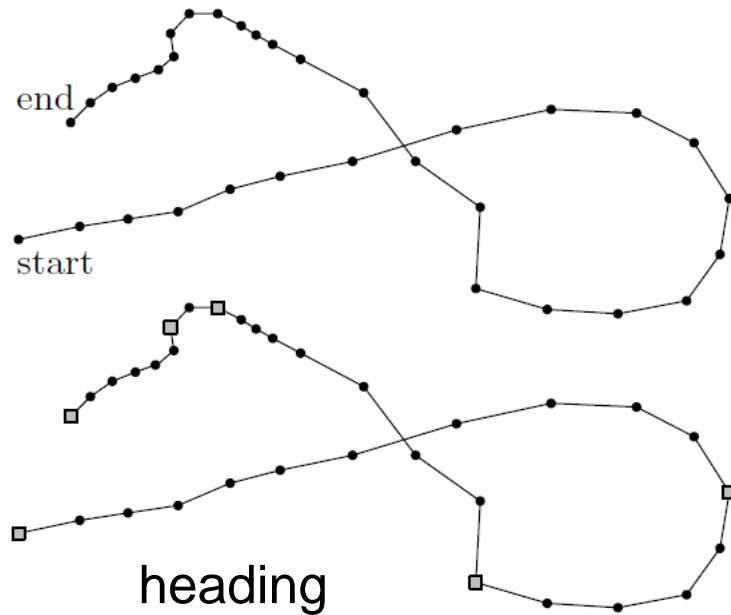
- ❑ Trajectory T sampled with equal time intervals
- ❑ **Criterion:** speed cannot differ more than a factor 2?

Example



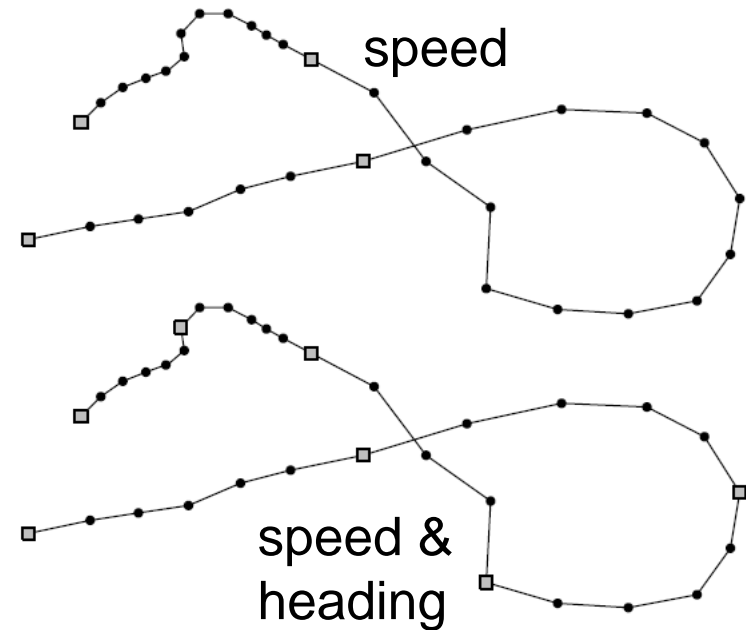
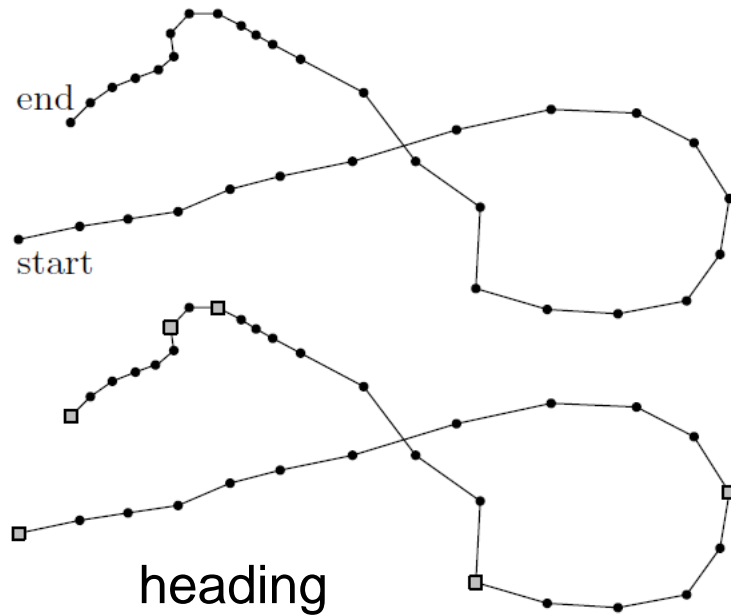
- ❑ Trajectory T sampled with equal time intervals
- ❑ **Criterion:** speed cannot differ more than a factor 2?
- ❑ **Criterion:** direction of motion differs by at most 90° ?

Example



- ❑ Trajectory T sampled with equal time intervals
- ❑ **Criterion:** speed cannot differ more than a factor 2?
- ❑ **Criterion:** direction of motion differs by at most 90° ?

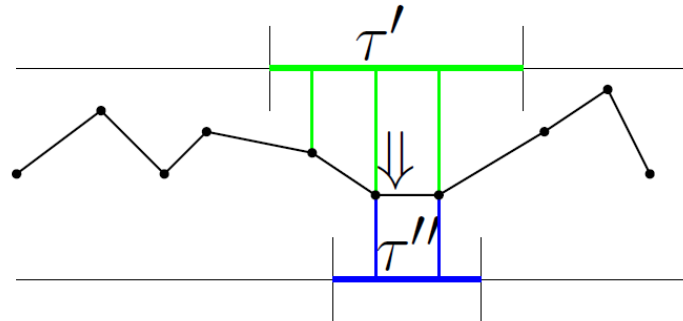
Example



- ❑ Trajectory T sampled with equal time intervals
- ❑ **Criterion:** speed cannot differ more than a factor 2?
- ❑ **Criterion:** direction of motion differs by at most 90° ?

Decreasing monotone criteria

- **Definition:** A criterion is decreasing monotone, if it holds on a segment, it holds on any subsegment.

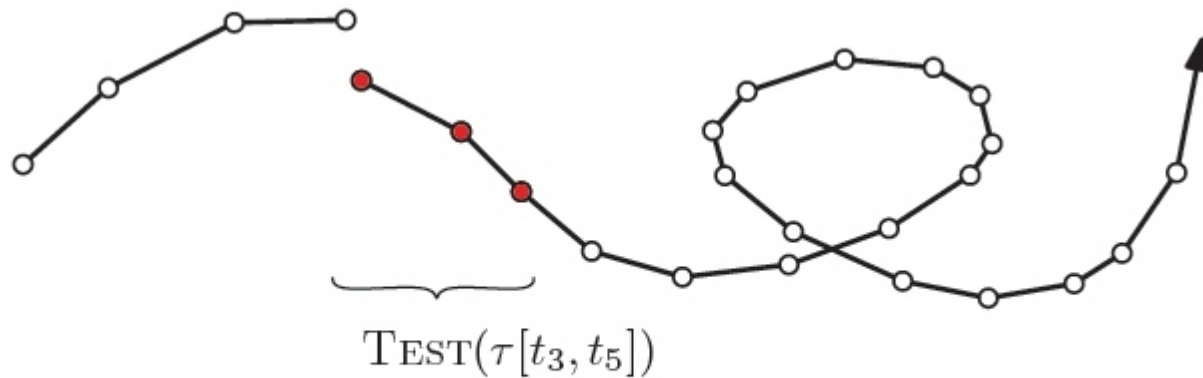


- **Examples:** disk criterion (location), angular range (heading), speed...
- **Theorem:** A combination of conjunctions and disjunctions of decreasing monotone criteria is a decreasing monotone criterion.

Greedy Algorithm

Observation:

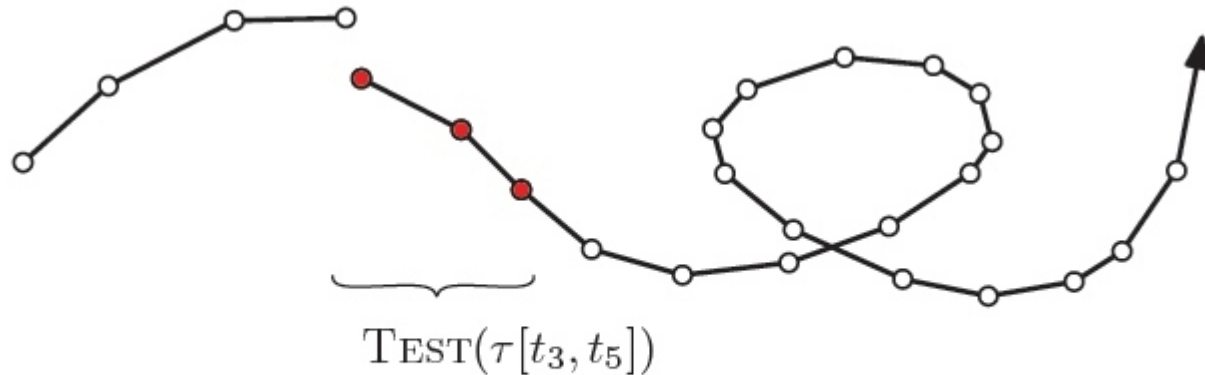
If criteria are decreasing monotone, a greedy strategy works.



Greedy Algorithm

Observation:

If criteria are decreasing monotone, a greedy strategy works.

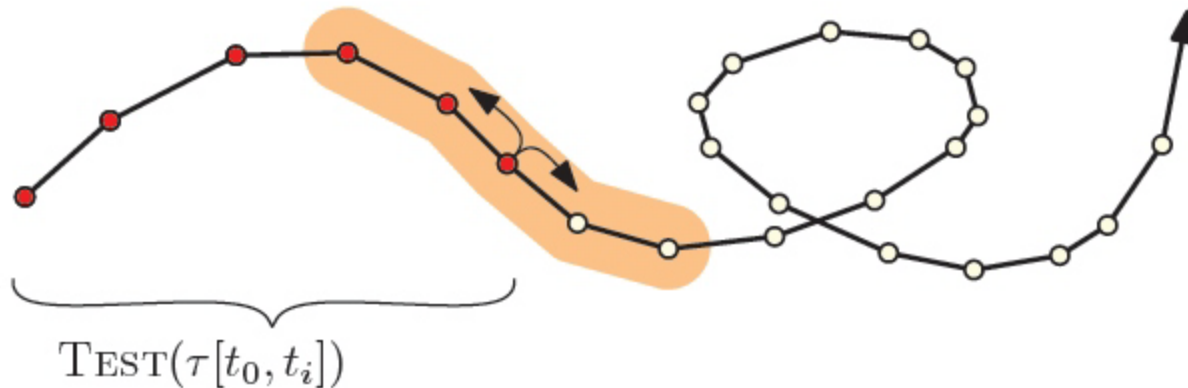


For many decreasing monotone criteria Greedy requires $O(n)$ time, e.g. for speed, heading...

Greedy Algorithm

Observation:

For some criteria, iterative double & search is faster.



Double & search: An exponential search followed by a binary search.

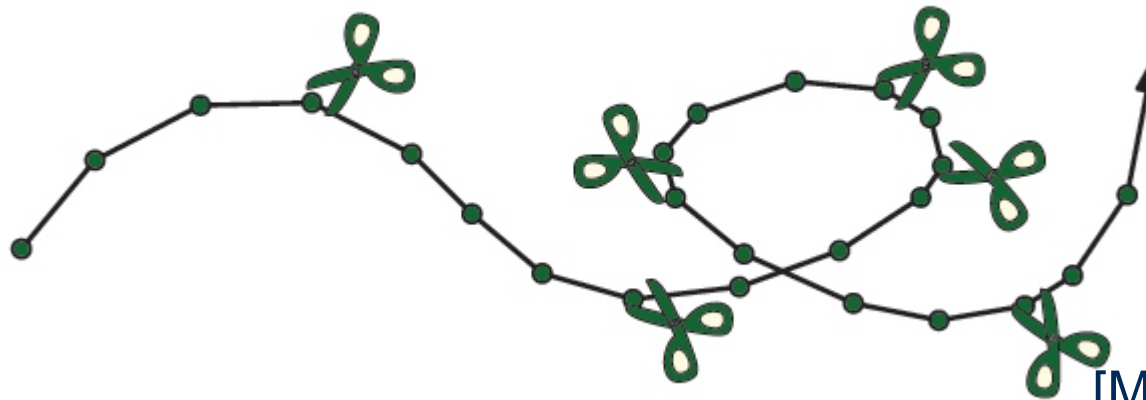
Criteria-Based Segmentation

Boolean or linear combination of decreasing monotone criteria

Greedy Algorithm

- incremental in $O(n)$ time or constant-update criteria e.g. bounds on speed or heading
- double & search in $O(n \log n)$ time for non-constant update criteria e.g. staying within some radius

Thus, solves Assignment 2, Ex.1)



[M.Buchin et al.'11]

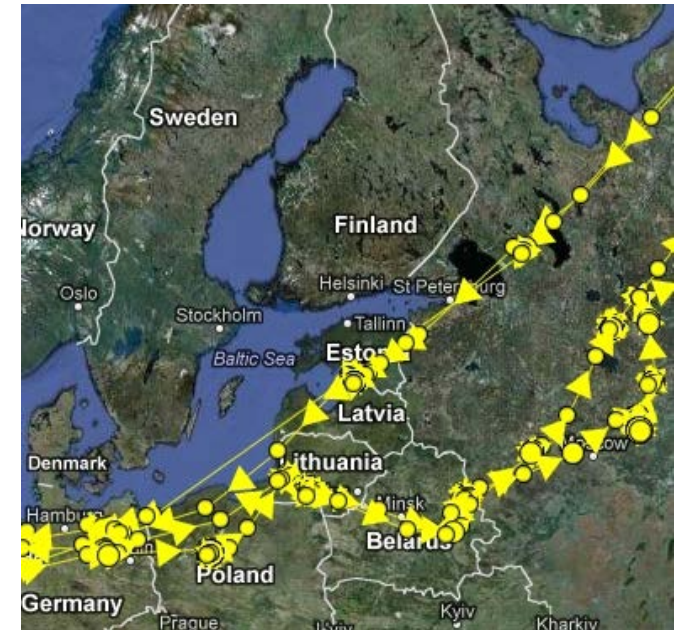
Motivation: Geese Migration

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Case Study: Geese Migration

Data

- ❑ Spring migration tracks
 - White-fronted geese
 - 4-5 positions per day
 - March – June

- ❑ Up to 10 stopovers during spring migration
 - Stopover: 48 h within radius 30 km
 - Flight: change in heading $<120^\circ$



Comparison

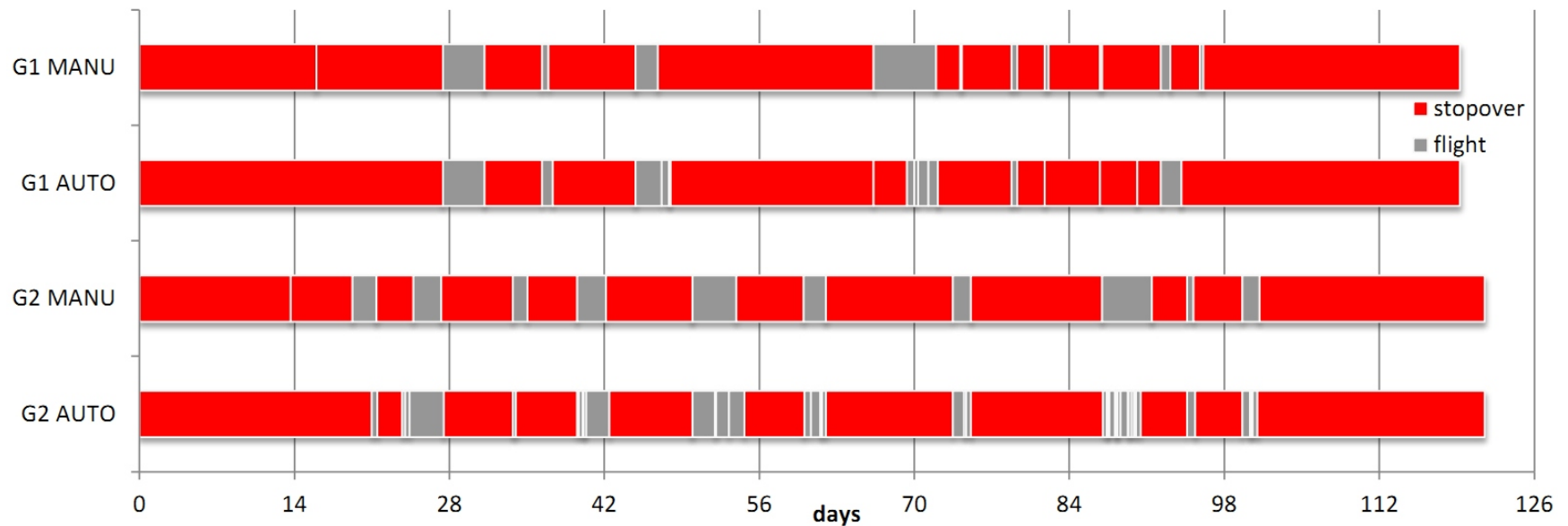


manual



computed

Evaluation



Few local differences:

Shorter stops, extra cuts in computed segmentation

Criteria

Within
radius 30km

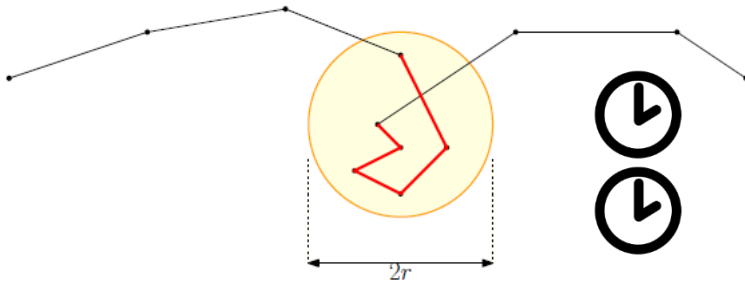
AND

At least
48h

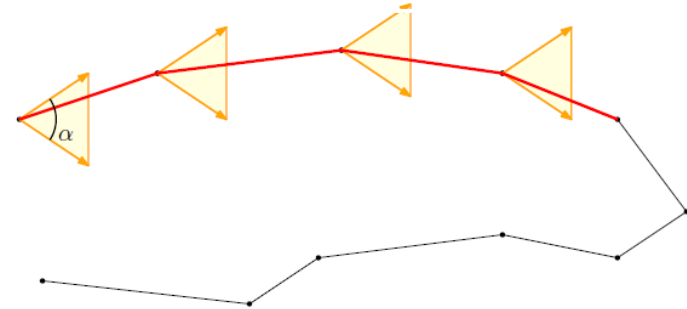
OR

Change in
heading $< 120^\circ$

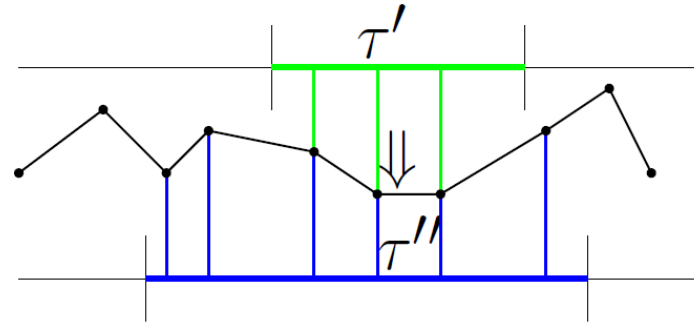
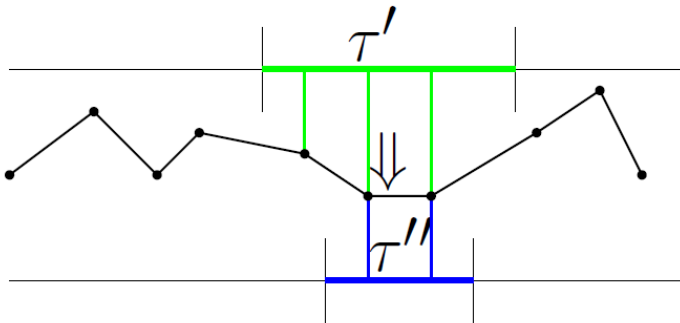
stopover



migration flight



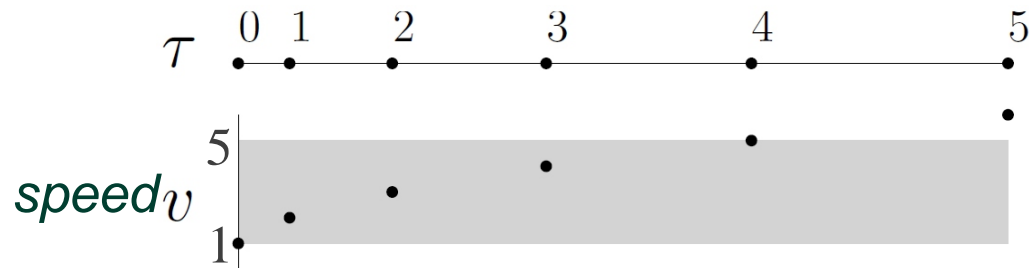
- A combination of decreasing and increasing monotone criteria



Decreasing and Increasing Monotone Criteria

- **Observation:** For a combination of decreasing and increasing monotone criteria the greedy strategy does not always work.

- **Example:** Min duration 2 AND Max speed range 4

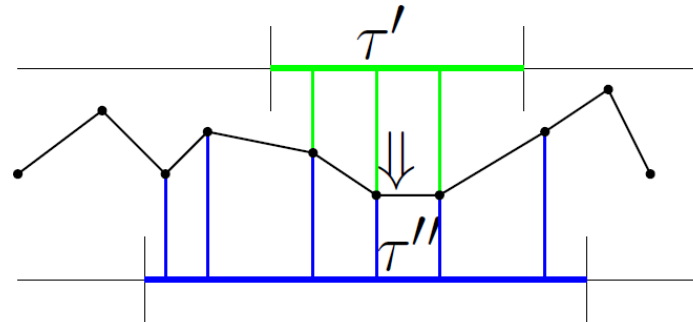
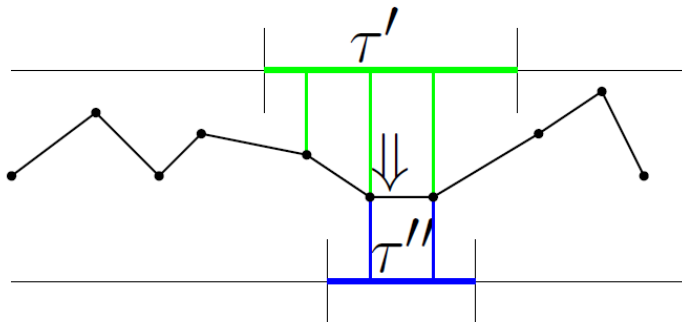


Non-Monotone Segmentation

Many Criteria are not (decreasing) monotone:

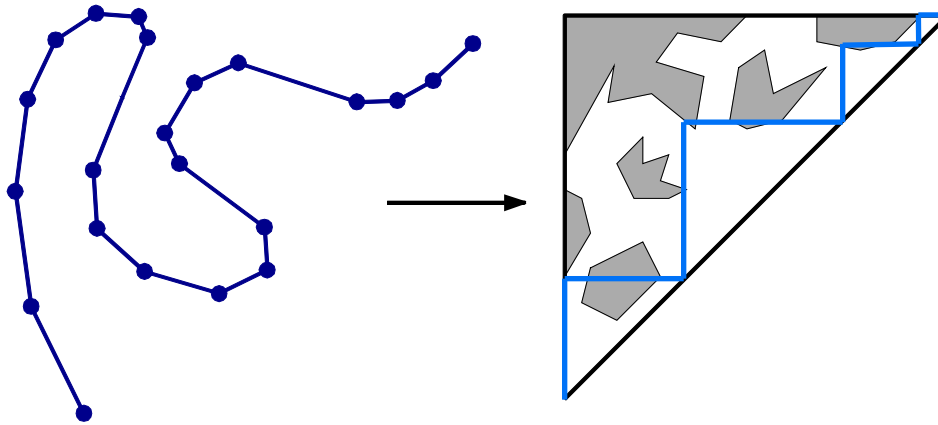
- ❑ Minimum time
- ❑ Standard deviation
- ❑ Fixed percentage of outliers
- ❑ For these Aronov et al. introduced the start-stop diagram

Example: Geese Migration



Start-Stop Diagram

Algorithmic approach



input trajectory

compute start-stop diagram

Start-Stop Diagram

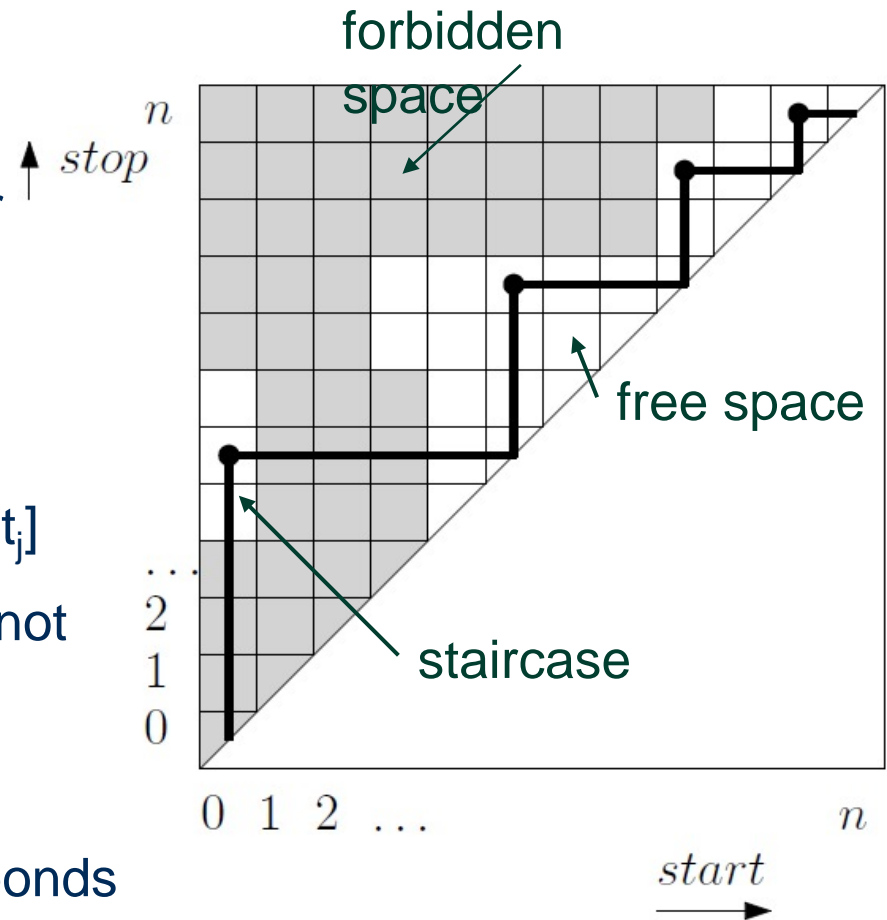
Given a trajectory T over time interval $I = \{t_0, \dots, t_\tau\}$ and criterion C

The **start-stop diagram** D is (the upper diagonal half of) the $n \times n$ grid,

where each point (i, j) is associated to segment $[t_i, t_j]$ with

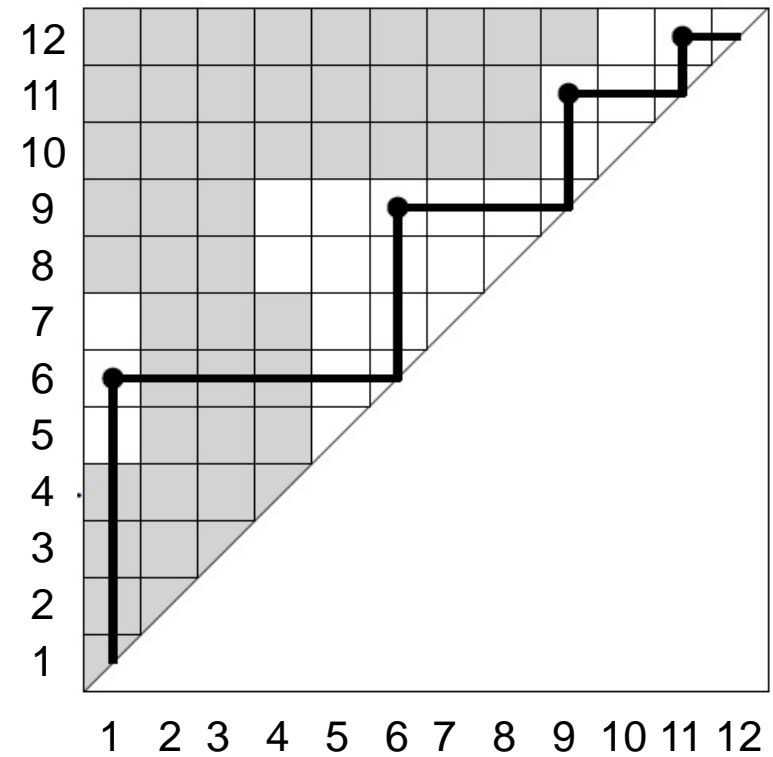
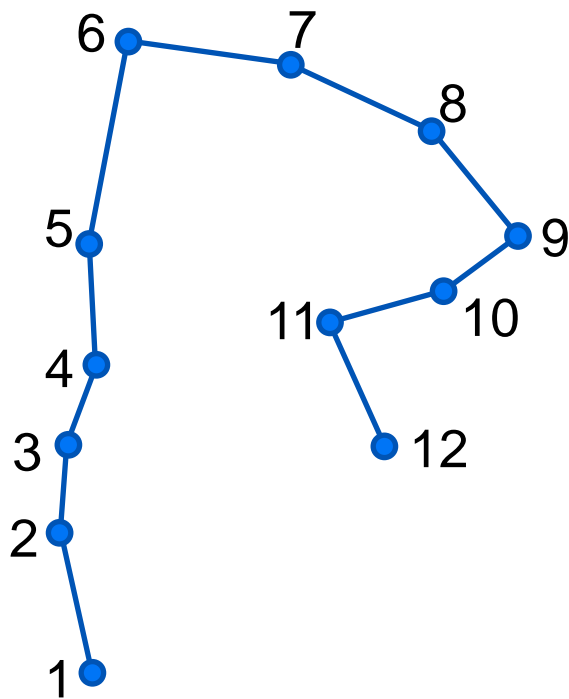
- (i, j) is in free space if C holds on $[t_i, t_j]$
- (i, j) is in forbidden space if C does not hold on $[t_i, t_j]$

A (minimal) segmentation of T corresponds to a (min-link) staircase in D



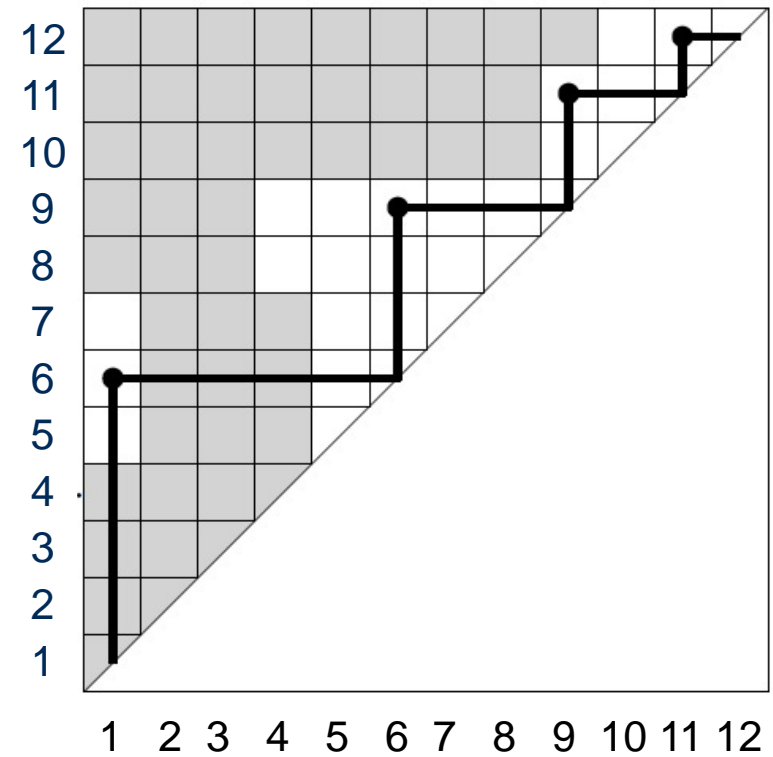
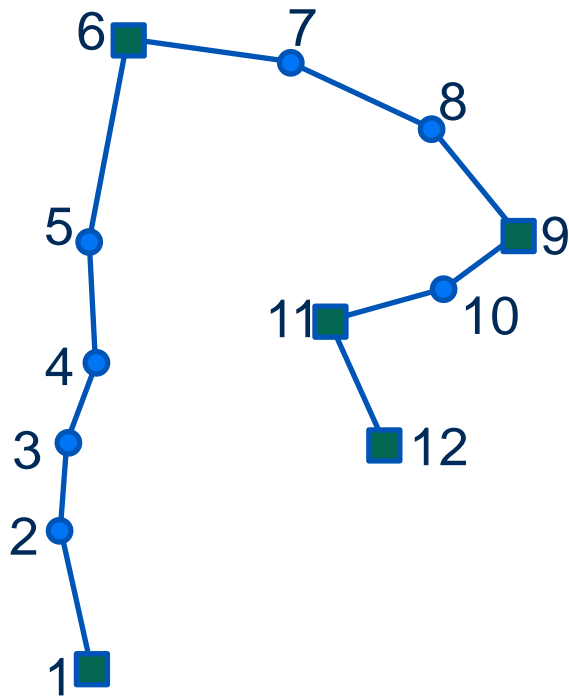
Start-Stop Diagram

A (minimal) segmentation of T corresponds to a (min-link) staircase in D .



Start-Stop Diagram

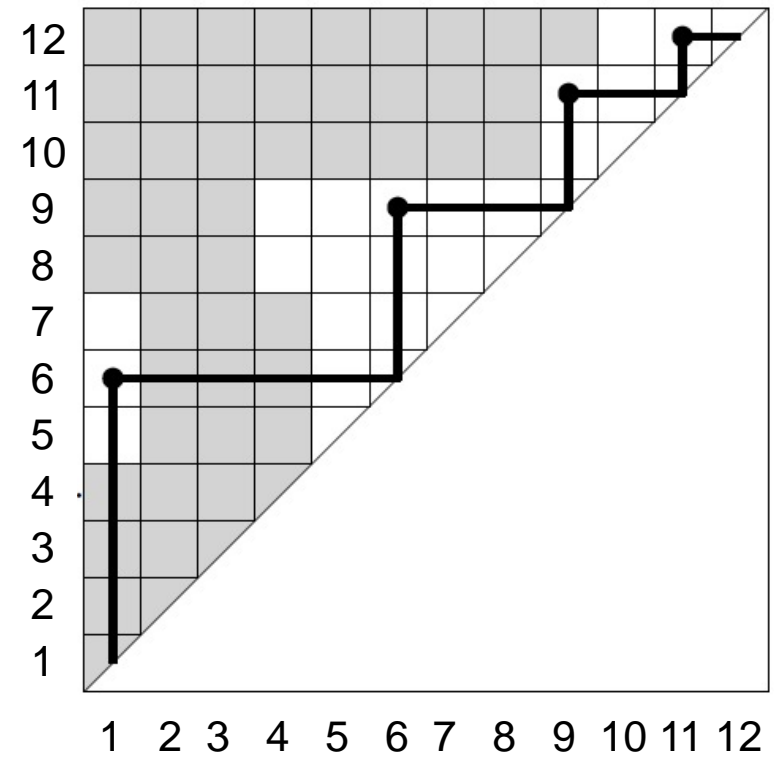
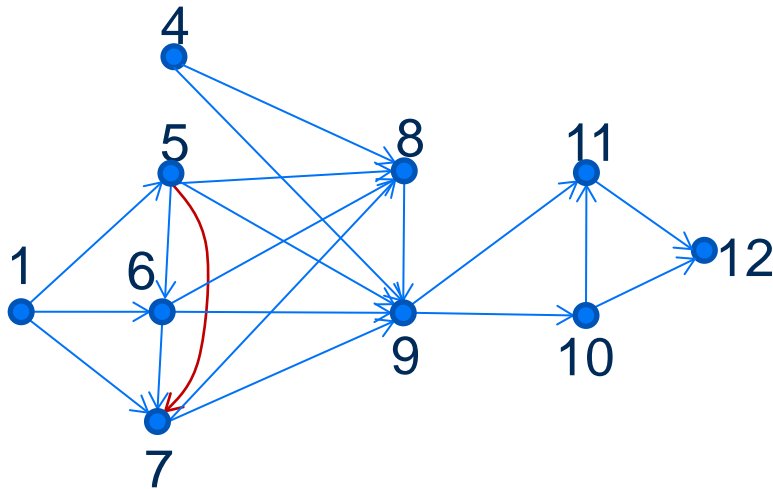
A (minimal) segmentation of T corresponds to a (min-link) staircase in D .



Start-Stop Diagram

Discrete case:

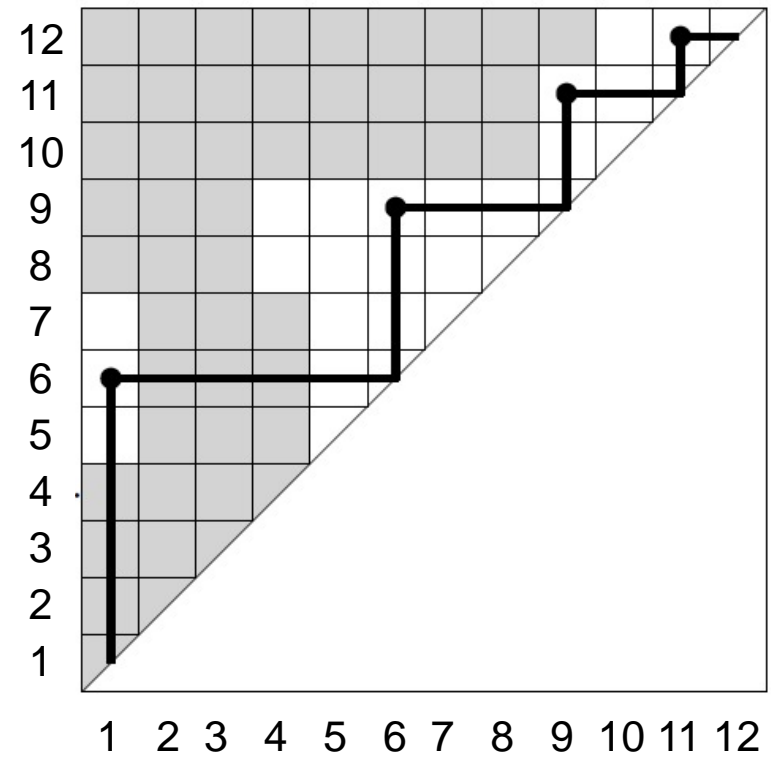
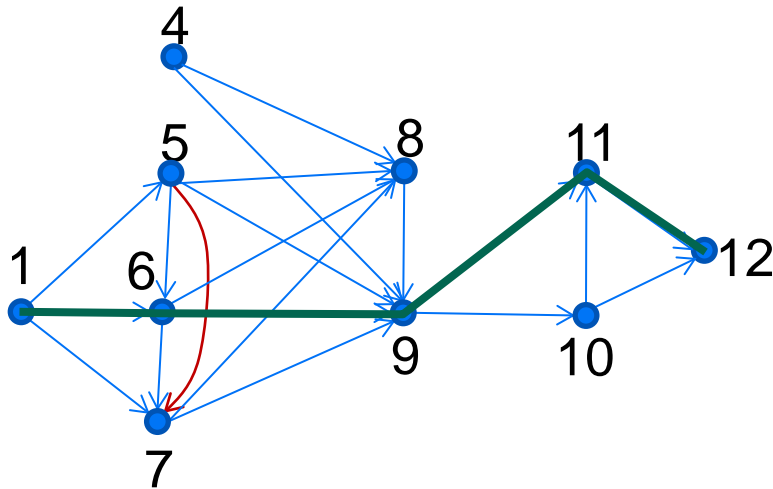
- A non-monotone segmentation can be computed in $O(n^2)$ time.



Start-Stop Diagram

Discrete case:

- A non-monotone segmentation can be computed in $O(n^2)$ time.



[Aronov et al.13]

Stable Criteria

Definition:

A criterion is **stable** if and only if $\sum_{i=0}^n v(i) = O(n)$ where

$v(i)$ = number of changes of validity on segments $[0,i]$, $[1,i]$, \dots , $[i-1,i]$

Stable Criteria

Definition:

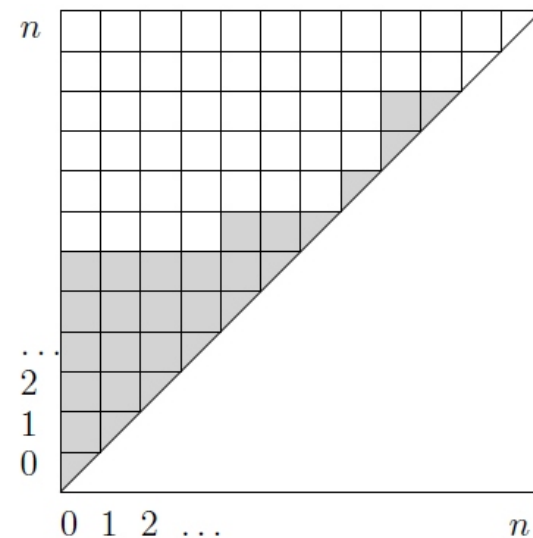
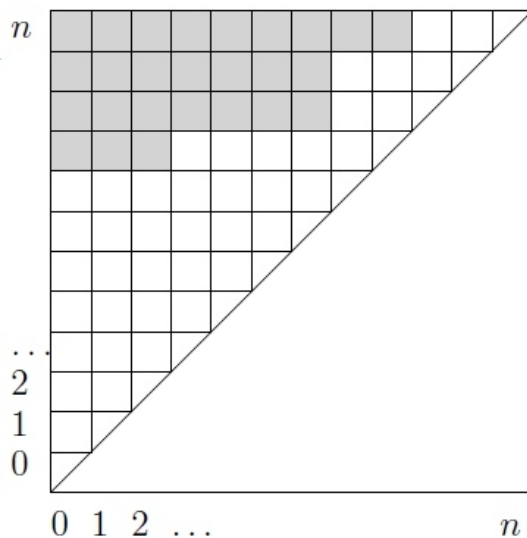
A criterion is **stable** if and only if $\sum_{i=0}^n v(i) = O(n)$ where

$v(i)$ = number of changes of validity on segments $[0,i]$, $[1,i]$, ..., $[i-1,i]$

Observations:

Decreasing and increasing monotone criteria are stable.

A conjunction or disjunction of stable criterion are stable.

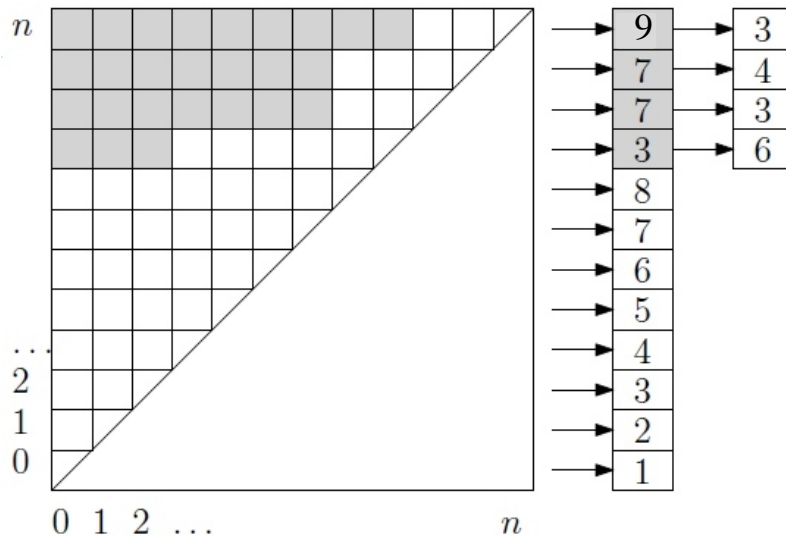


Compressed Start-Stop Diagram

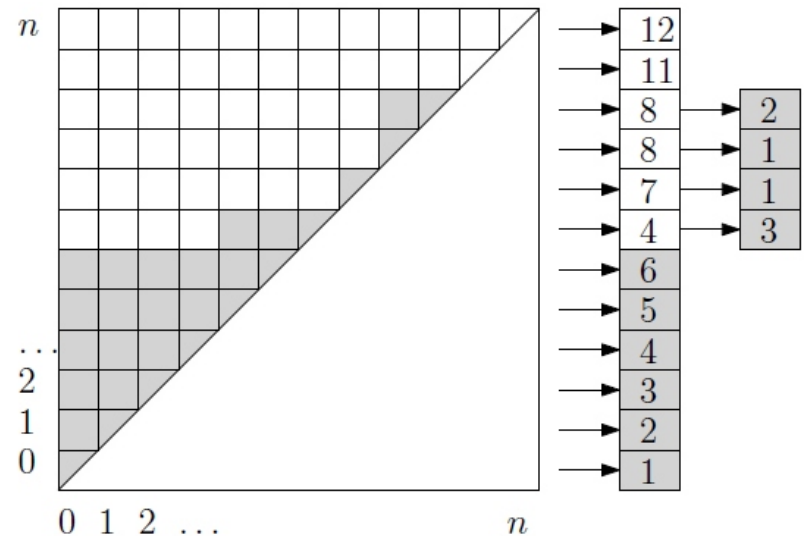
For stable criteria the start-stop diagram can be compressed by applying run-length encoding.

Examples:

decreasing monotone



increasing monotone



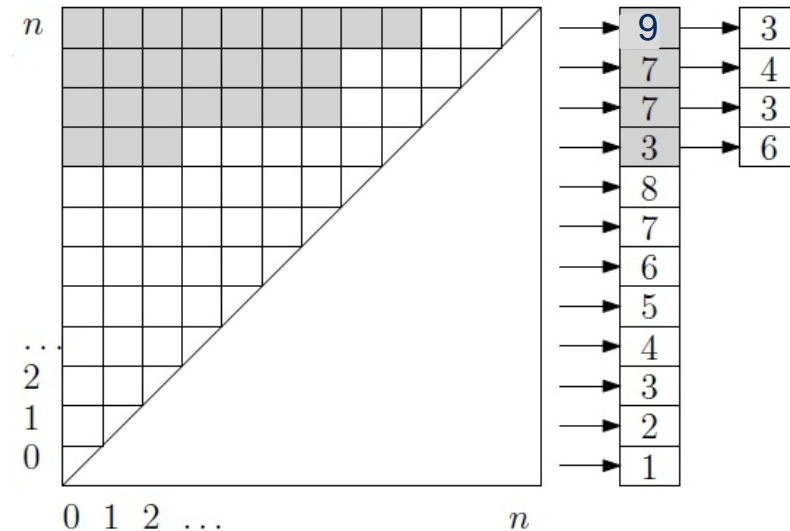
Computing the Compressed Start-Stop Diagram

For a decreasing criterion consider the algorithm:

ComputeLongestValid(crit C, traj T)

Algorithm:

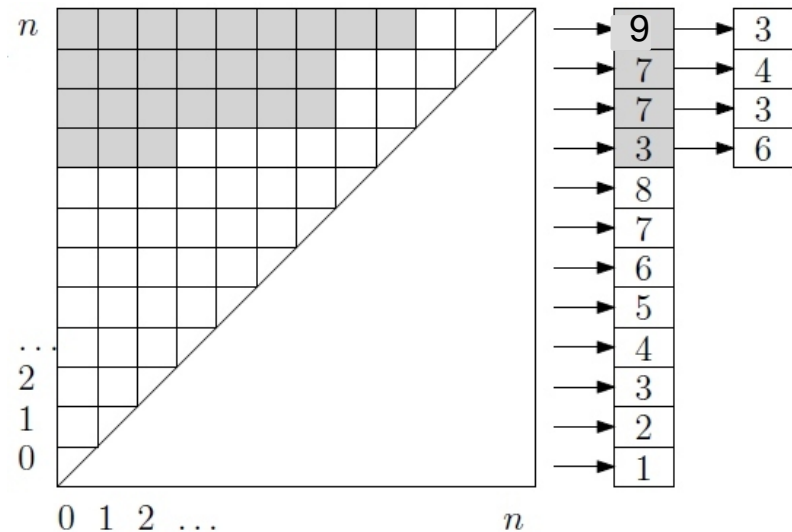
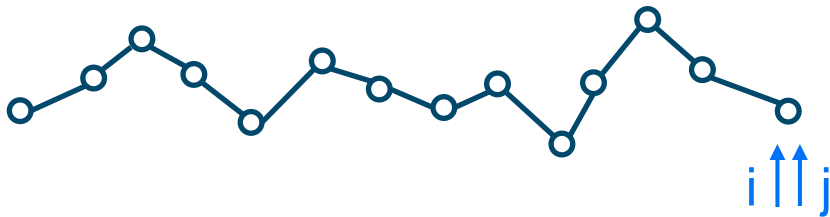
Move two pointers i, j from n to 0 over the trajectory. For every trajectory index j the smallest index i for which $\pi[i, j]$ satisfies the criterion C is stored.



Computing the Compressed Start-Stop Diagram

For a **decreasing criterion** *ComputeLongestValid*(crit *C*, traj *T*):

Move two pointers *i, j* from *n* to 0 over the trajectory



Requires a data structure for segment $[i, j]$ allowing the operations *isValid*, *extend*, and *shorten*, e.g., a balanced binary search tree on attribute values for range or bound criteria.

Runs in $O(n \cdot c(n))$ time where $c(n)$ is the time to update & query.

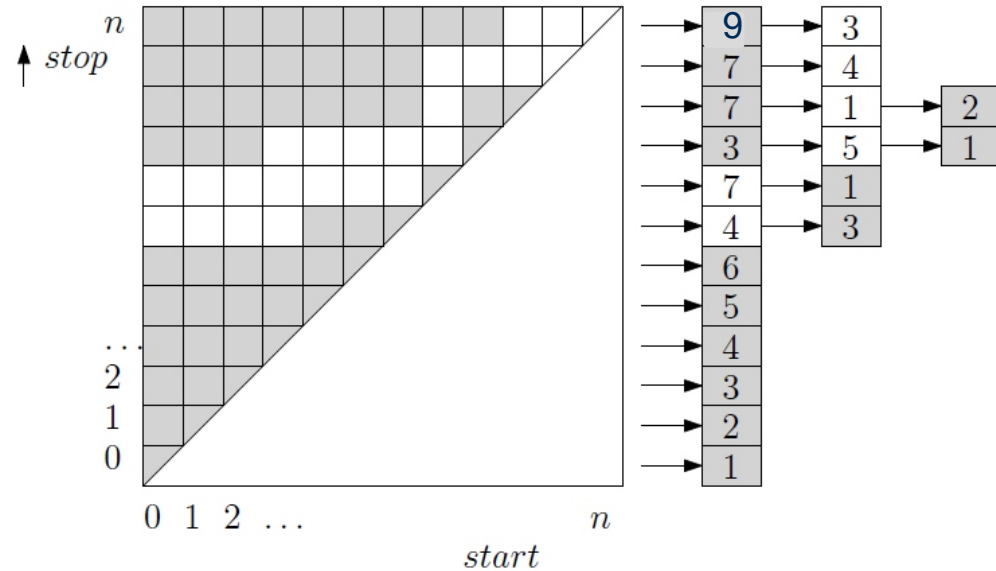
Analogously for **increasing criteria**.

Computing the Compressed Start-Stop Diagram

The start-stop diagram of a conjunction (or disjunction) of two stable criteria is their intersection (or union).

The start-stop diagram of a negated criteria is its inverse.

The corresponding compressed start-stop diagrams can be computed in $O(n)$ time.



Attributes and criteria

Examples of stable criteria

- ❑ Lower bound/Upper bound on attribute
- ❑ Angular range criterion
- ❑ Disk criterion
- ❑ Allow a fraction of outliers
- ❑ ...

Computing the Optimal Segmentation

Observation: The optimal segmentation for $[0,i]$ is either one segment, or an optimal sequence of segments for $[0,j<i]$ appended with a segment $[j,i]$, where j is an index such $[j,i]$ is valid.

Dynamic programming algorithm

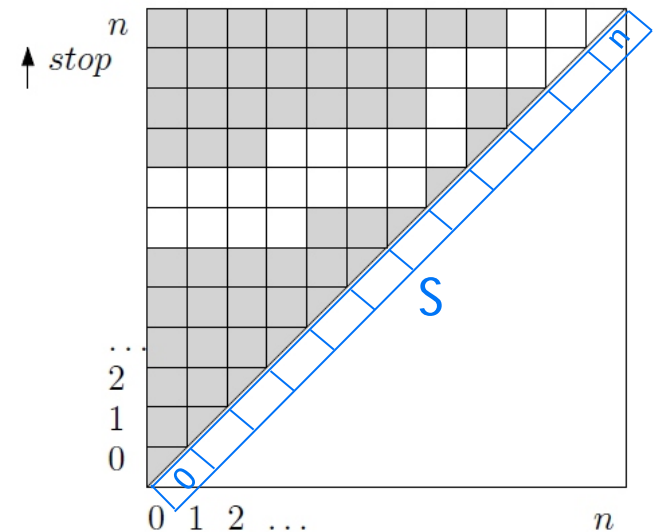
for each row from 0 to n

find white cell with min-link

That is, iteratively compute a table $S[0,n]$ where entry $S[i]$ for row i stores

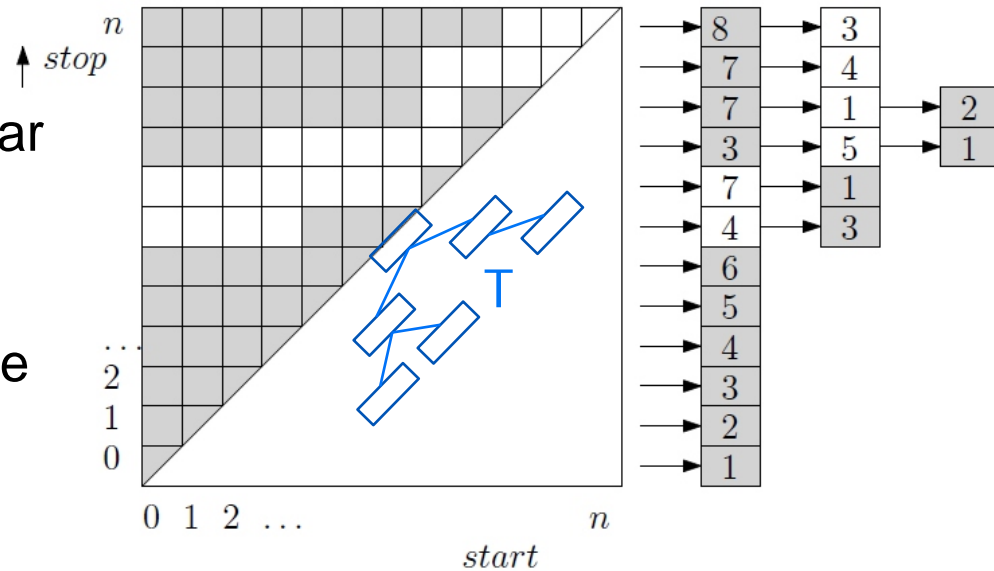
- **last:** index of last link
- **count:** number of links so far

□ runs in $O(n^2)$ time



Computing the Optimal Segmentation

- More efficient dynamic programming algorithm for compressed diagrams
- Process blocks of white cells using a range query in a binary search tree T (instead of table S) storing
 - index**: row index
 - last**: index of last link
 - count**: number of links so far
- augmented by minimal **count** in subtree
- runs in $O(n \log n)$ time



[Alewijns et al.'14]

Beyond criteria-based segmentation

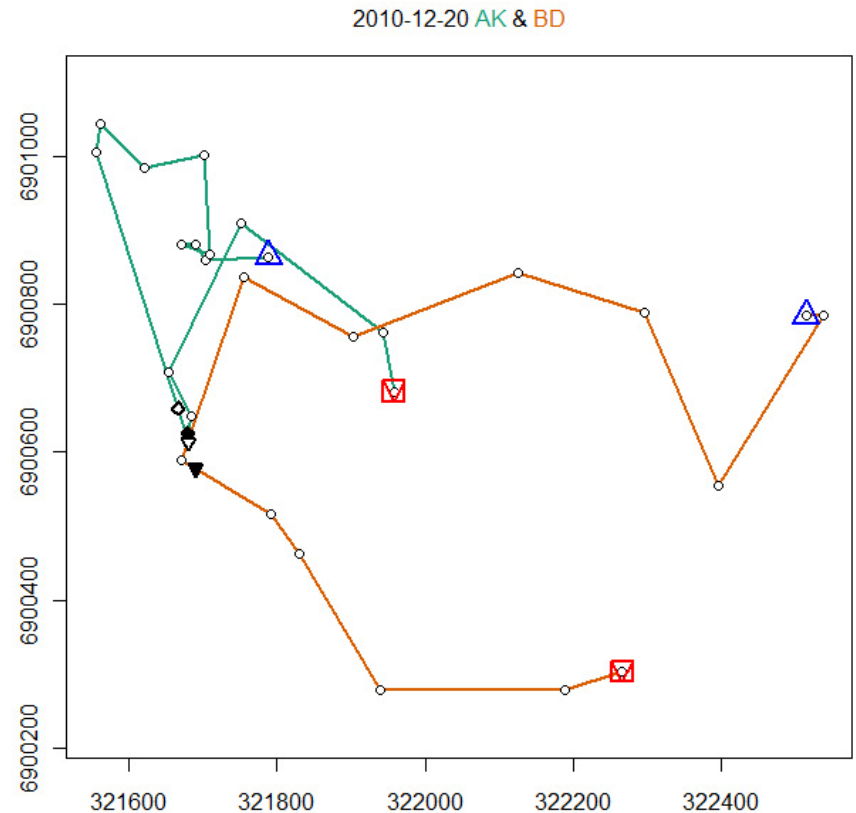
- ❑ When is criteria-based segmentation applicable?
- ❑ What can we do in other cases?

Excursion

MOVEMENT MODELS

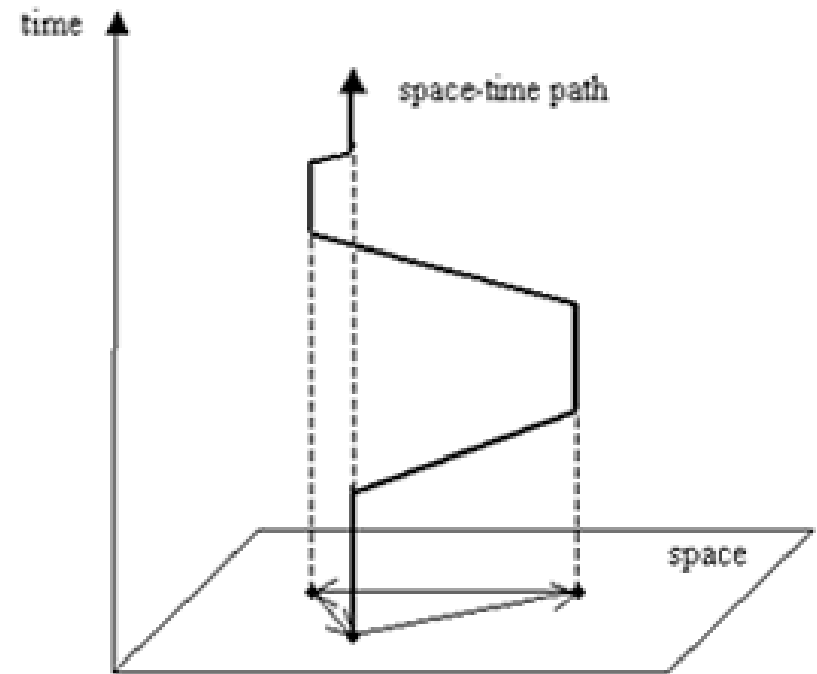
Movement Models

- ▣ Movement data: sequence of observations, e.g. (x_i, y_i, t_i)
- ▣ Linear movement realistic?



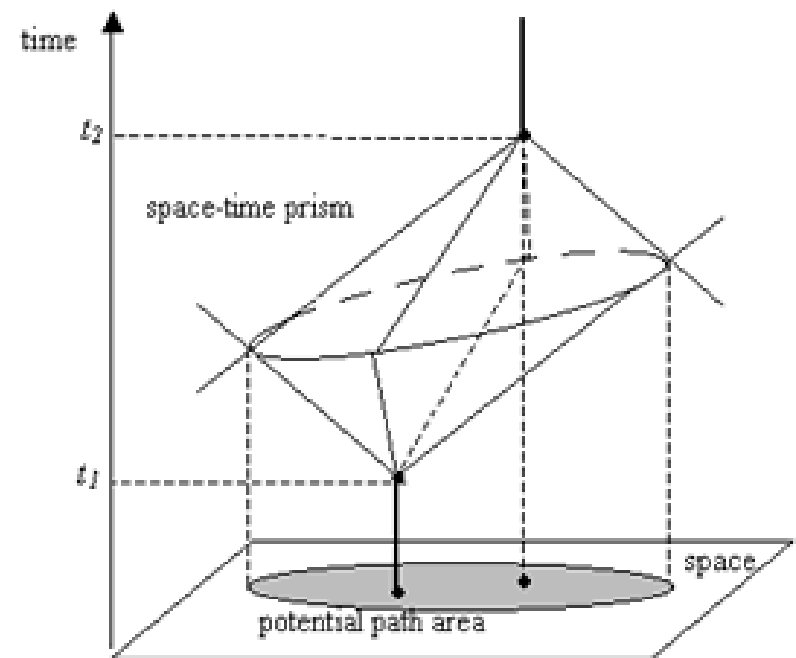
Movement Models

- ❑ Movement data: sequence of observations, e.g. (x_i, y_i, t_i)
- ❑ Linear movement realistic?
- ❑ Space-time Prisms



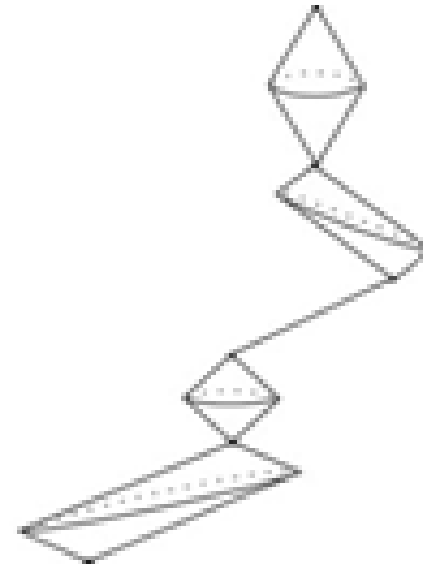
Movement Models

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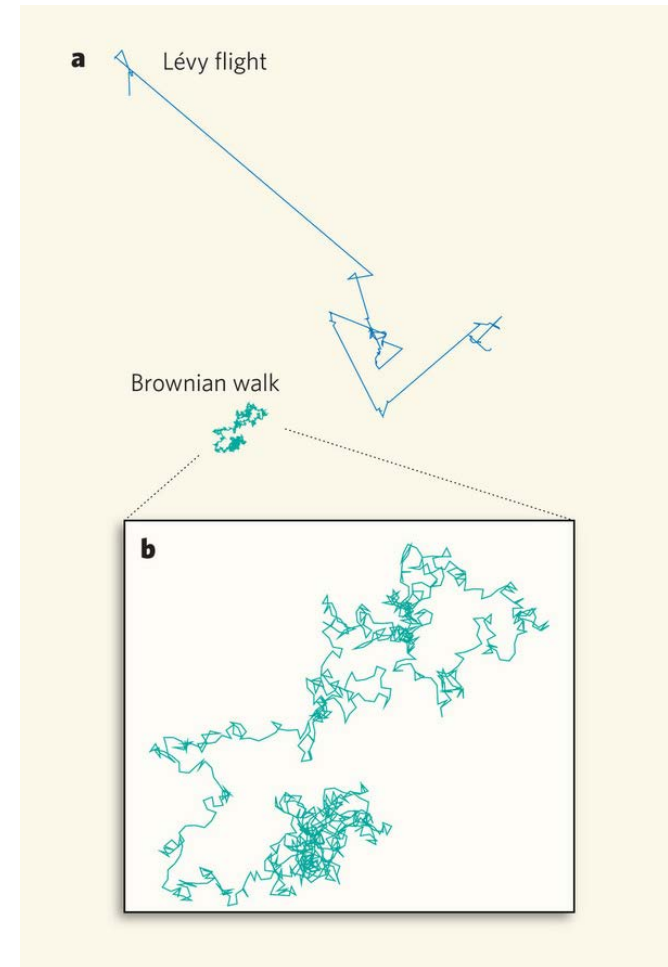
Movement Models

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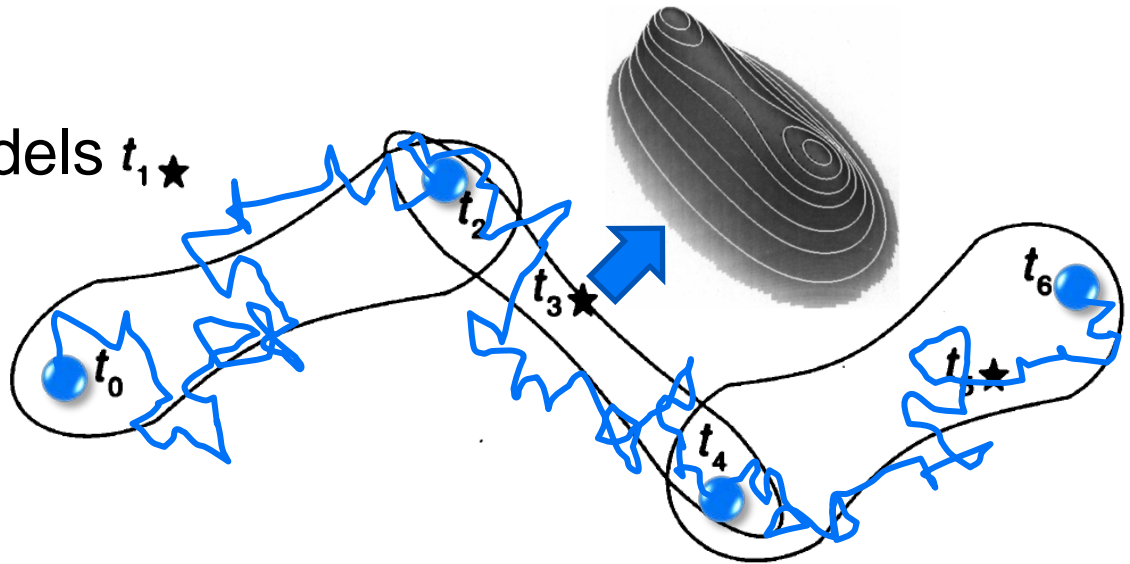
Movement Models

- ▣ Movement data: sequence of observations, e.g. (x_i, y_i, t_i)
- ▣ Linear movement realistic?
- ▣ Space-time Prisms
- ▣ Random Motion Models (Brownian bridges)

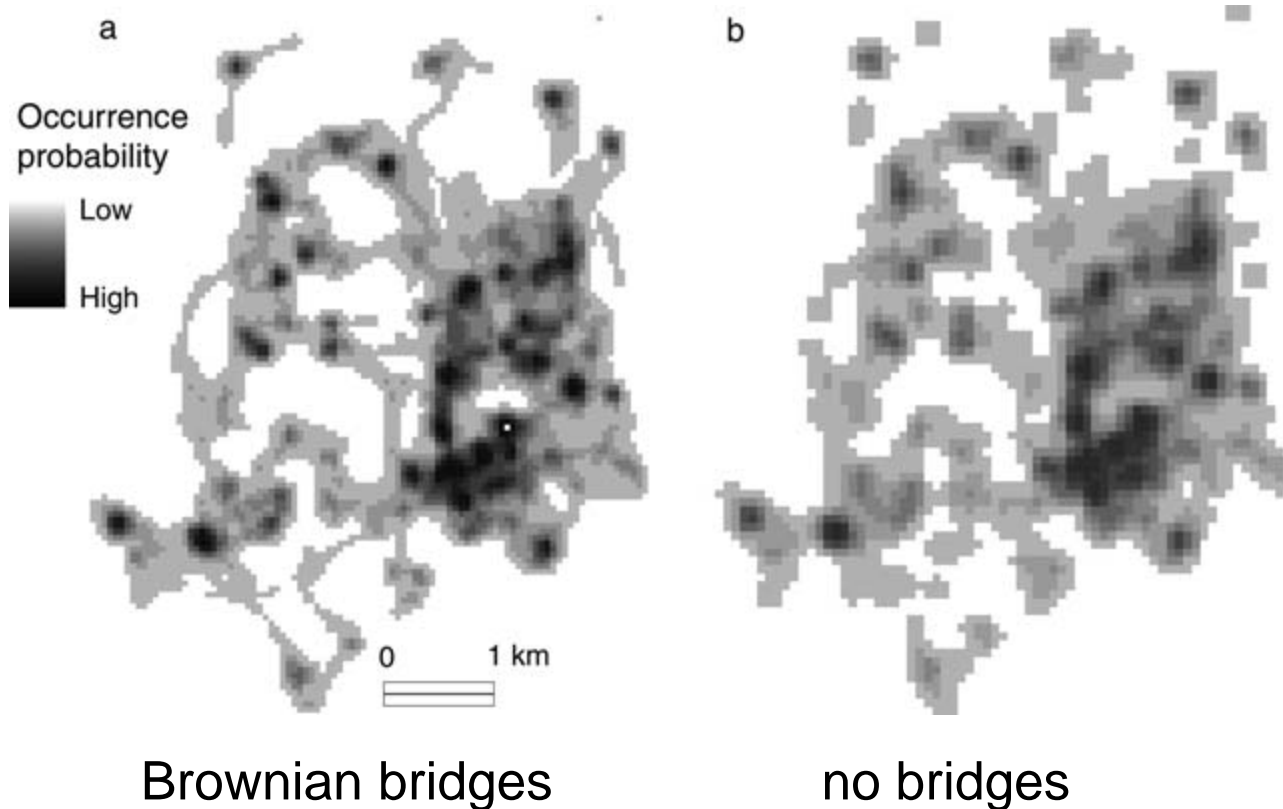


Movement Models

- ▣ Movement data: sequence of observations, e.g. (x_i, y_i, t_i)
- ▣ Linear movement realistic?
- ▣ Space-time Prisms
- ▣ Random Motion Models $t_1 \star$
(Brownian bridges)

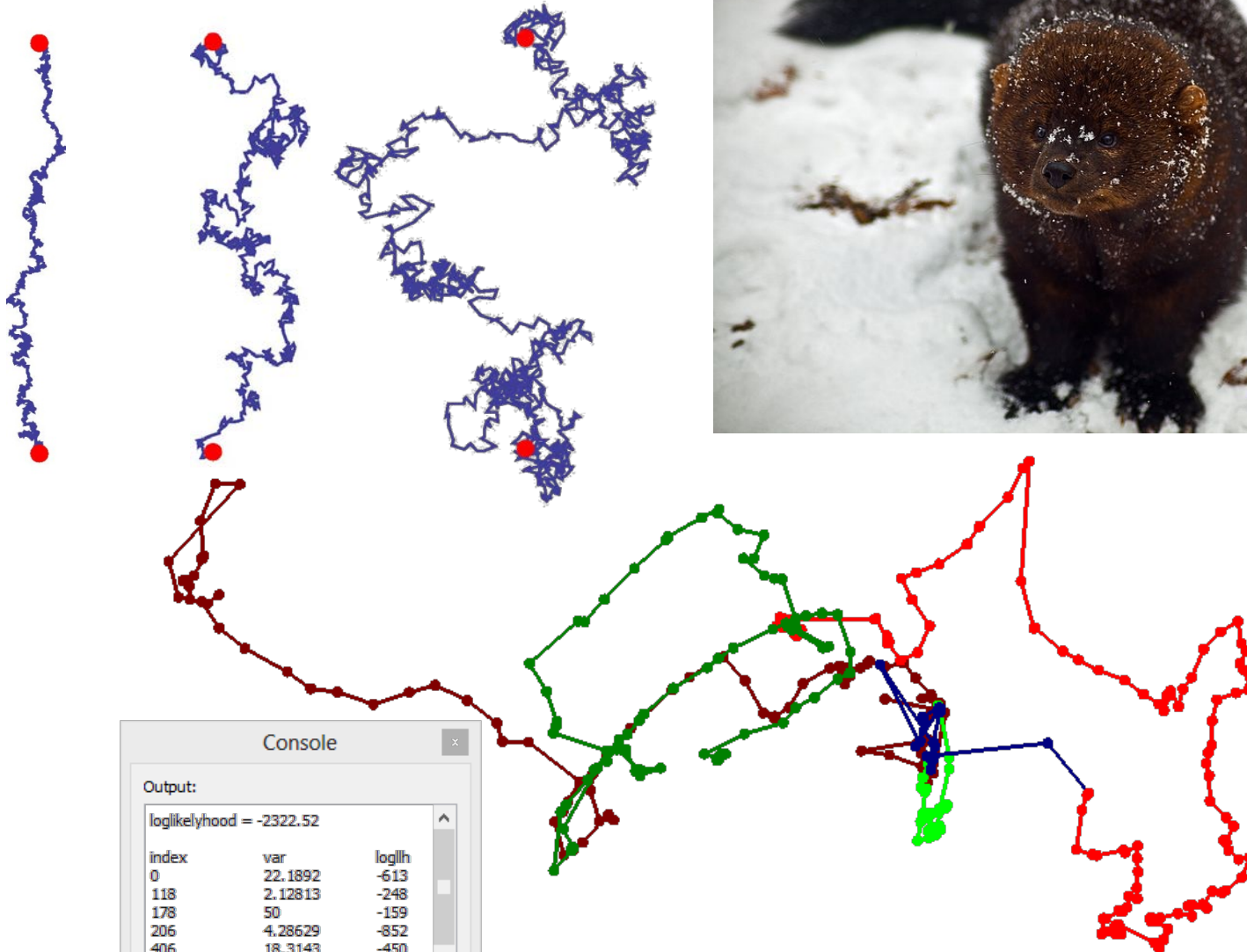


Brownian Bridges - Examples





Segment by diffusion coefficient



Console

Output:

loglikelihood = -2322.52

index	var	loglikh
0	22.1892	-613
118	2.12813	-248
178	50	-159
206	4.28629	-852
406	18.3143	-450

Summary

- Greedy algorithm for decreasing monotone criteria $O(n)$ or $O(n \log n)$ time

[M.Buchin, Driemel, van Kreveld, Sacristan, 2010]

- Case Study: Geese Migration

[M.Buchin, Kruckenberg, Kölzsch, 2012]

- Start-stop diagram for arbitrary criteria $O(n^2)$ time

[Aronov, Driemel, van Kreveld, Löffler, Staals, 2012]

- Compressed start-stop diagram for stable criteria $O(n \log n)$ time

[Alewijnse, Buchin, Buchin, Sijben, Westenberg, 2014]

References

- S. Alewijnse, T. Bagautdinov, M. de Berg, Q. Bouts, A. ten Brink, K. Buchin and M. Westenberg. Progressive Geometric Algorithms. SoCG, 2014.
- M. Buchin, A. Driemel, M. J. van Kreveld and V. Sacristan. Segmenting trajectories: A framework and algorithms using spatiotemporal criteria. Journal of Spatial Information Science, 2011.
- B. Aronov, A. Driemel, M. J. Kreveld, M. Löffler and F. Staals. Segmentation of Trajectories for Non-Monotone Criteria. SODA, 2013.
- M. Buchin, H. Kruckenberg and A. Kölzsch. Segmenting Trajectories based on Movement States. SDH, 2012.