

# Pattern recognition in pedestrian movement trajectories

Colin Kuntzsch

[Colin.Kuntzsch@ikg.uni-hannover.de](mailto:Colin.Kuntzsch@ikg.uni-hannover.de)

Monika Sester

[Monika.Sester@ikg.uni-hannover.de](mailto:Monika.Sester@ikg.uni-hannover.de)

Institute of Cartography and Geoinformatics (ikg)  
Hannover, Germany

# Overview

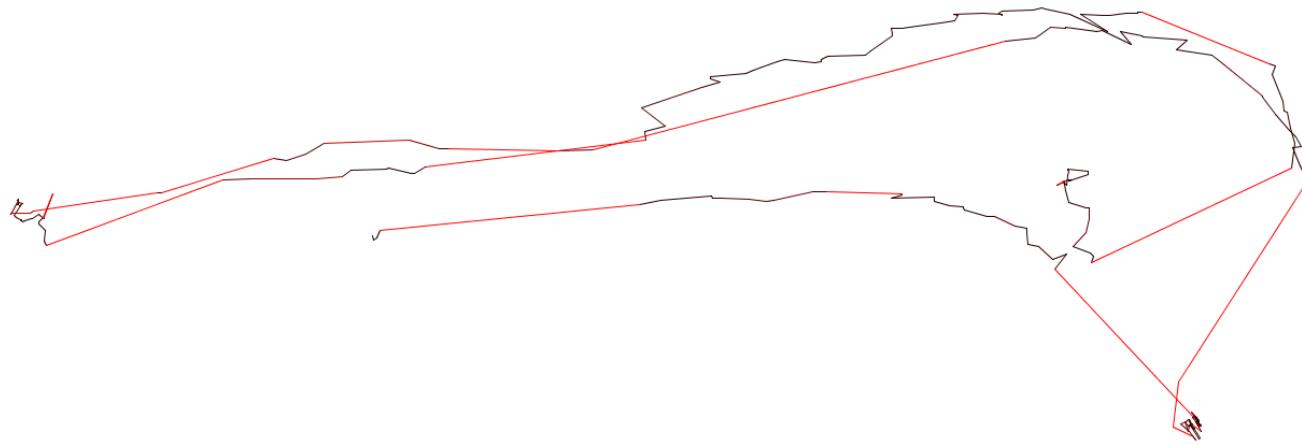
- ▶ BMBF project CamInSens



- ▶ Self-organized smart-camera network in a surveillance scenario
- ▶ pattern analysis on trajectory data
- ▶ collaborative camera tracking: generation of 3D-models
- ▶ user interface: visualization of observed anomalous behaviour (large amounts of spatial data)
- ▶ investigation of legal boundary conditions

# Challenges

- ▶ work with huge amounts of spatially distributed trajectory data
- ▶ real-time processing → need for incremental algorithms
- ▶ deal with limited precision, temporal/spatial resolution, short-term loss of tracking
- ▶ identify anomalous behaviour from small sample sizes
- ▶ build a scene-specific, spatio-temporal model of common behaviour



# Geometric analysis: trajectory attributes

- ▶ position, heading
- ▶ speed
- ▶ periodic lateral movement:  
frequency, step length

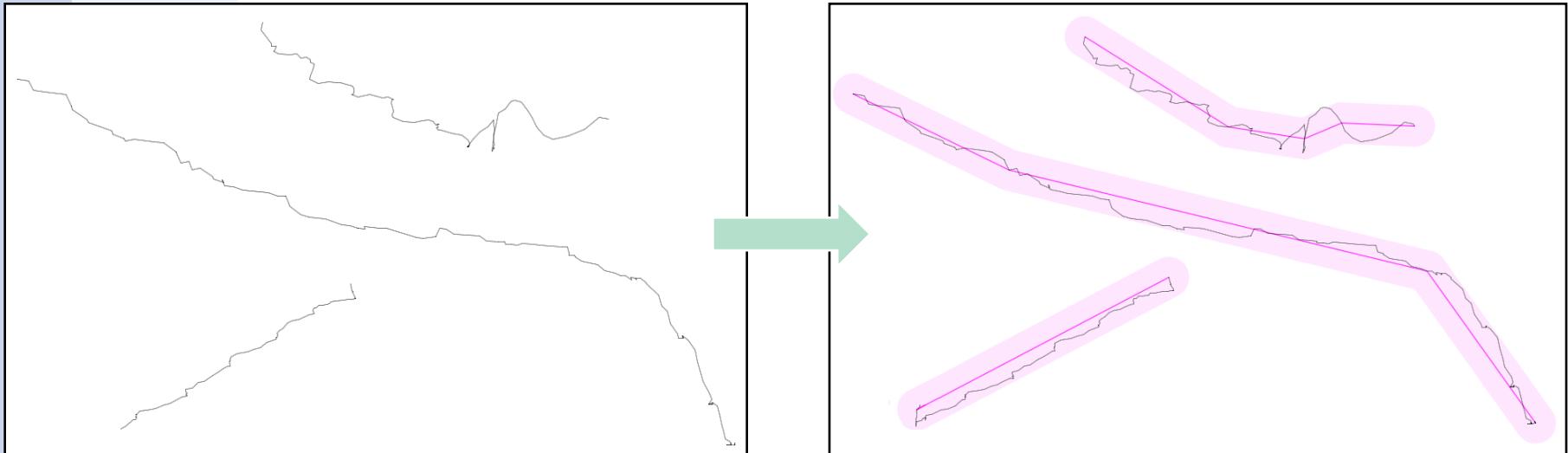


movement prediction

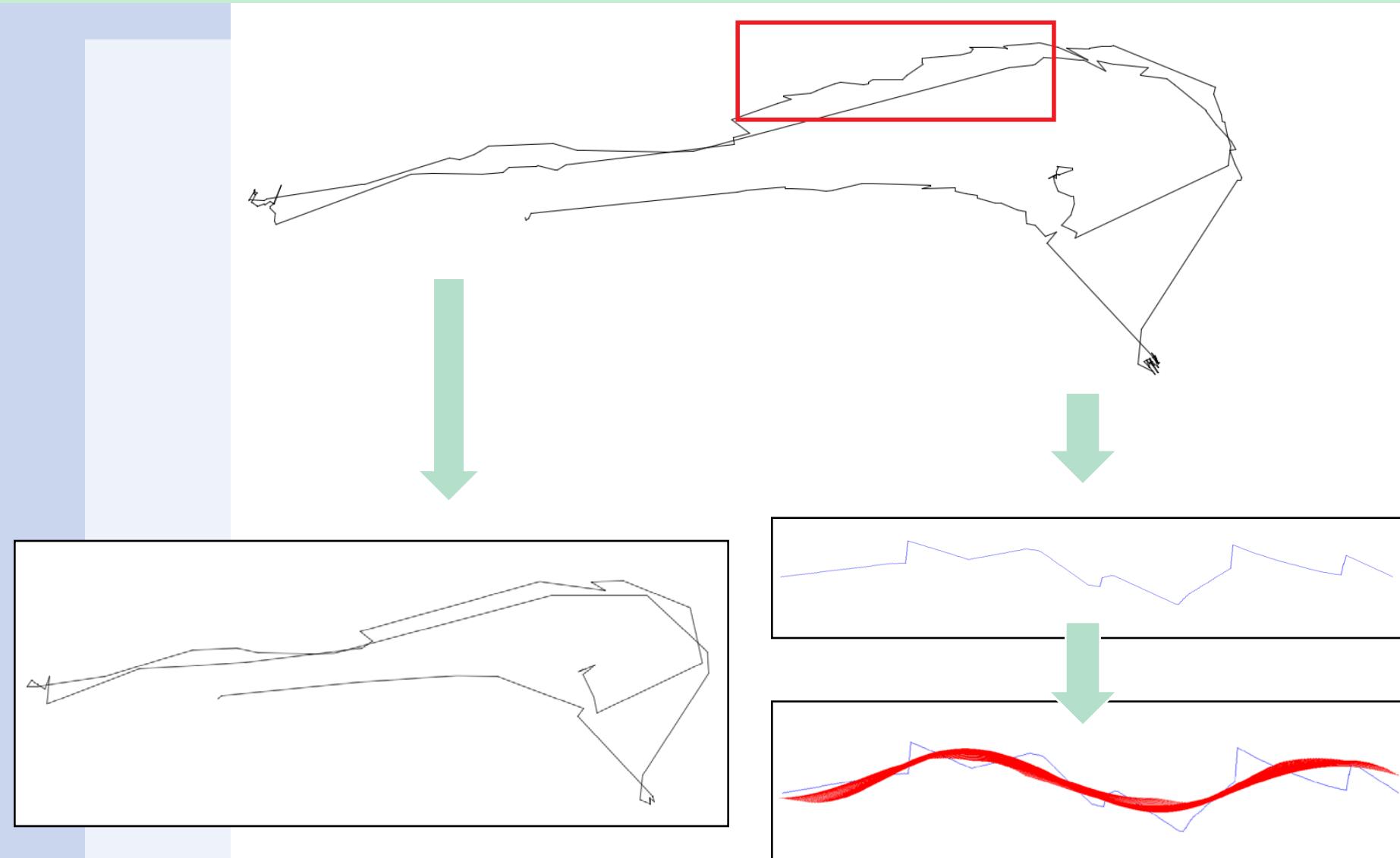
matching of unconnected  
trajectory segments

# Geometric analysis: trajectory pre-processing

- ▶ reduction of noise from trajectory data
- ▶ separate significant movement from fine-granular movement
  - piece-wise linearization utilizing a corridor width resembling the average width of human pedestrian movement (0.71 m)
  - swaying: lateral oscillation of trajectories due to alternating foot movement
  - indexing of trajectory with piecewise linear approximation



# Geometric analysis: trajectory pre-processing

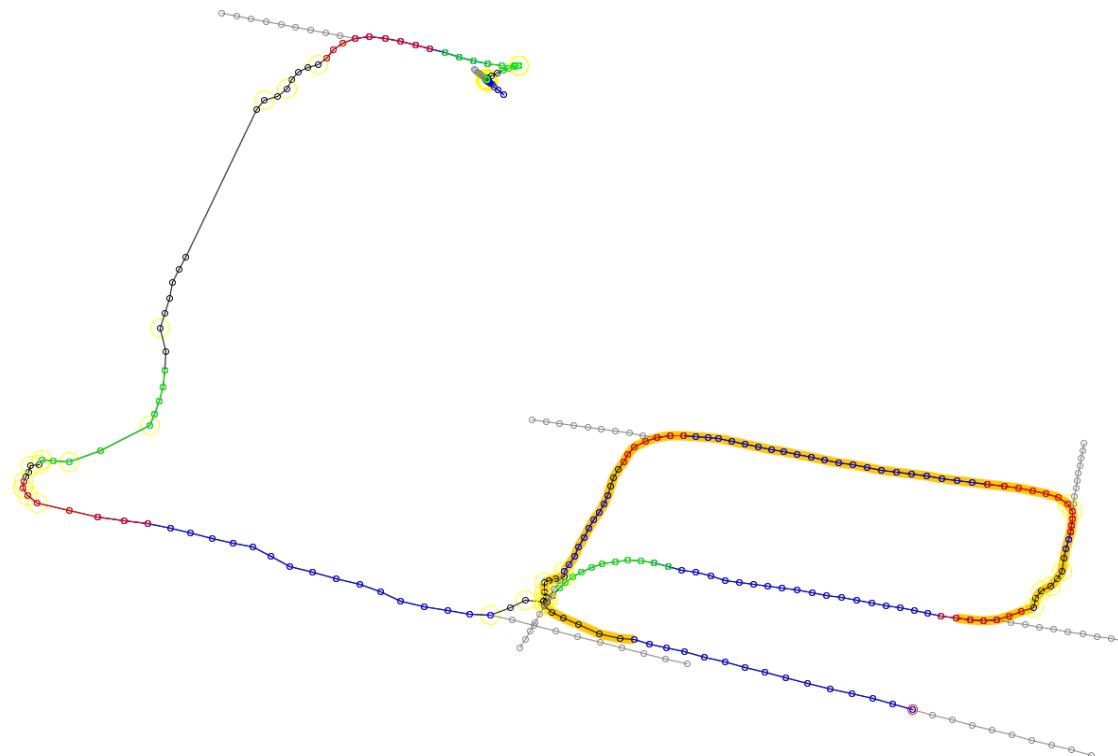


# Geometric analysis: segmentation

- ▶ Split trajectory up into
  - left/right curves
  - straight movement
  - circular movement
  - stops



semantic interpretation in combination with prior knowledge and other trajectories



# Geometric analysis: search for circular structures

- ▶ our approach: utilize list of cumulative turn angles
  - sum greater 360 degrees between fixes i and j: at least one (full) circle contained in trajectory segment  $t[i,j]$
  - search innermost circle
  - remove circle from turn angle list
  - repeat until no more circles are found
  - use angle and distance between first/last circle segments for classification of circle



# Geometric analysis: search for turns

- ▶ similar for turns
  - cumulative angles greater 45 degrees labeled as left/right turns
  - straight segments do not contain turns or circles
  - additional length criterion

# Outlook: trajectories within spatio-temporal context

- ▶ very few pre-defined patterns to actively look for (hard to identify most patterns from short trajectory samples)
- ▶ unsupervised learning of common behaviour within scene
  - typical trajectory attributes (space and time dependant)
  - typical low level patterns (e.g. stops, circles, turns, exits and entries)
  - detection of uncommon behaviour: raise visual notification
  - feedback-mechanism: security personnel manually classifies specific uncommon behaviour as relevant/irrelevant