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UNIVERSITÄT
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Vienna University of Technology

*“EXTRACTION OF USER’S STAYS FROM GPS LOGS:
A COMPARISON OF THREE SPATIO-TEMPORAL CLUSTERING APPROACHES”*

International Master in Cartography

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Wien 16.12.2015

Overview

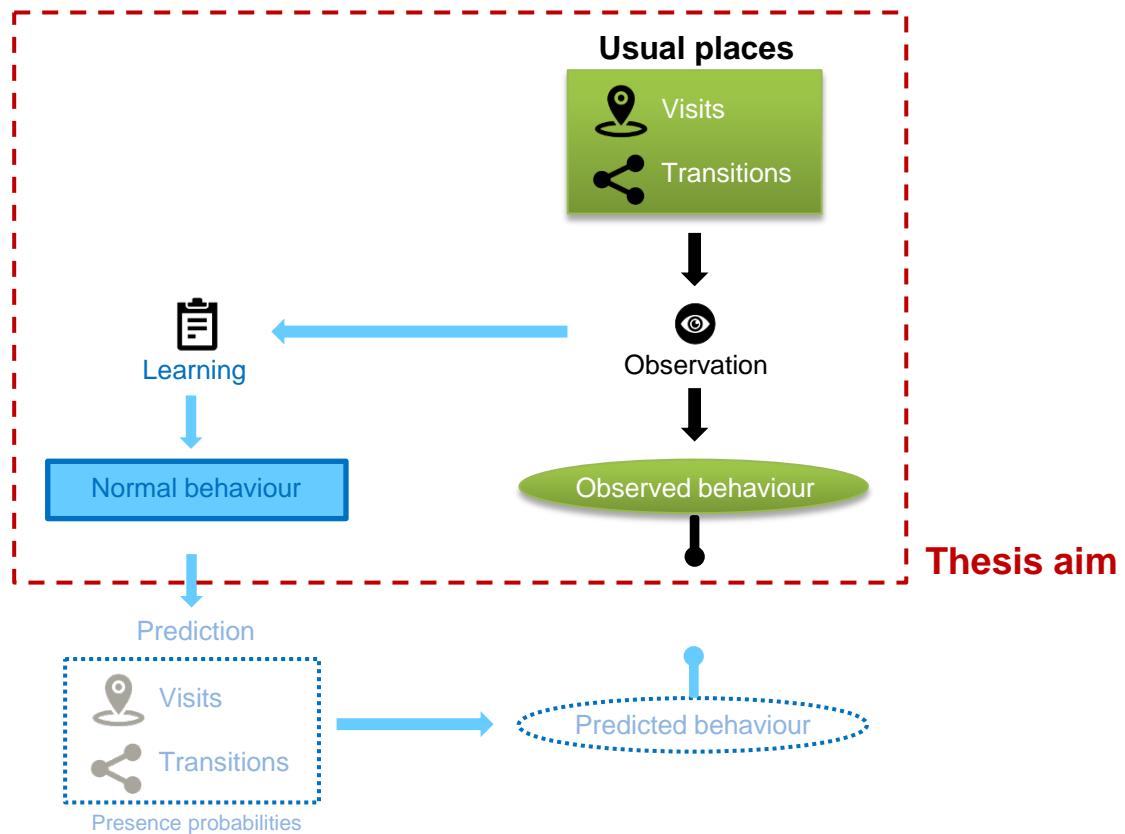
*“EXTRACTION OF USER’S STAYS FROM GPS LOGS:
A COMPARISON OF THREE SPATIO-TEMPORAL CLUSTERING APPROACHES”*

- 1. Introduction
- 2. Theory
- 3. Method
- 4. Implementation
- 5. Results
- 6. Conclusions

1. Introduction

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USE CASE



1. Introduction

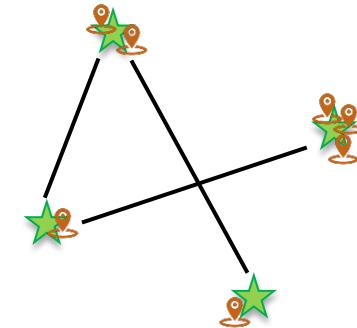
Research identification

Research objectives

- (i) Detecting **visited places** in user's life in an automated way



- (ii) Characterising **stays** at visited places and **transitions** between them



Visited place

Geographic **location** in which a **user** has been located during a **minimum period of time** (stay).



Stay

Physical **presence** of a **user** at a geographic location during a period of **time**.



Transition

Change of presence between **two different** geographic **locations** in which user has a **stay**.



→ Spatio-temporal clustering

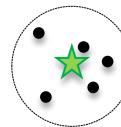
1. Introduction

Research identification

Research questions

- Which spatio-temporal **clustering approach** is the most adequate?
 - **Best algorithm** to detect user's visited places?
 - **Differences** between tested algorithms?
 - Best values for algorithms **parameters**?

- Which **approach** is adequate to **characterise** stays and transitions?
 - Relevant information for **stays** representation?
 - Relevant information for **transitions** representation?



2. Theory

- 1. Introduction
- **2. Theory**
- 3. Method
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- 6. Conclusions

Clustering algorithms

- **1. Incremental clustering**

TBC (*Kang et al. 2005*)

- **2. Incremental + density-based clustering**

TBC (*Ye et al. 2009*) + DBSCAN (*Ester et al. 1996*)

- **3. Density-based clustering**

DBSCAN (*Ester et al. 1996*)

Quality Evaluation

(*Salzburg Research 2015*)

- **4 quality measures**

- **Confusion matrix**

2. Theory

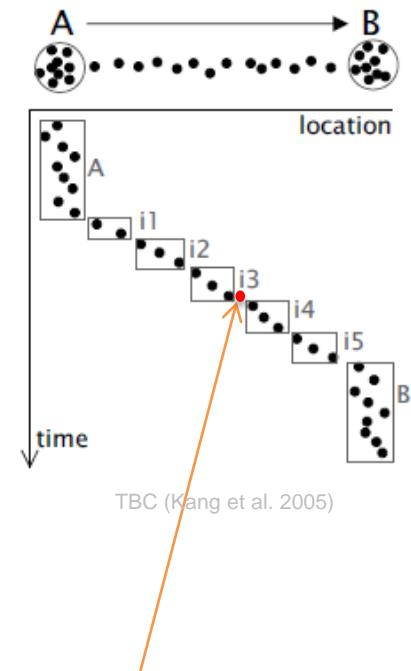
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Clustering algorithms

➤ 1. Incremental clustering (Kang et al. 2005)

- Clusters **computed incrementally** as new location estimates are generated.
- 2 main parameters: **d** and **t** determine **number** and **size** of extracted places.

- A stream of coordinates is clustered along time
- If stream moves away from current cluster and **distance > d** → **new cluster**
- Smaller clusters, where little time spent, dropped (i_1, \dots, i_5)
- If cluster duration: **time $\geq t$** → **detected place** (A, B)



- Third parameter “**L**” to determine if the user is really **moving away** from current cluster position

If **plocs** grows beyond **L seconds** the user is really moving away and starts a new cluster

2. Theory

Clustering algorithms

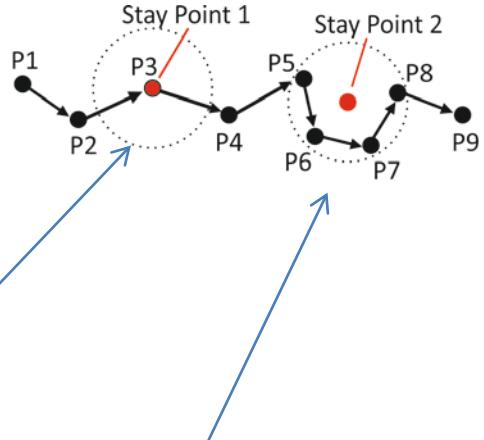
➤ 2. Incremental + density-based clustering (Ye et al. 2009)

1st Algorithm (Ye et al. 2009) + 2nd DBSCAN (Ester et al. 1996)

Stay point: a geographic region in which the user stays for a while

Two types considered:

- Individual maintains **stationary** at a point for over a time threshold ($> t$)
(e.g. enters a building)
- Individual **wanders around** within a spatial region ($\leq d$) for over a time threshold ($> t$)
(e.g. park, square)
 - Mean longitude and latitude of GPS points construct a stay point
 - Arrival time and leaving time respectively equals timestamp of first and last GPS point constructing stay point. (P5, P8)



2. Theory

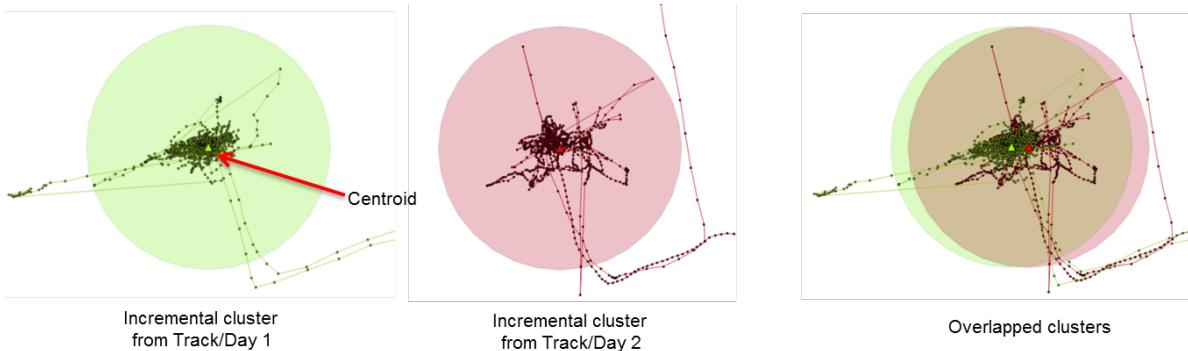
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Clustering algorithms

➤ 2. Incremental + density-based clustering (Ye et al. 2009)

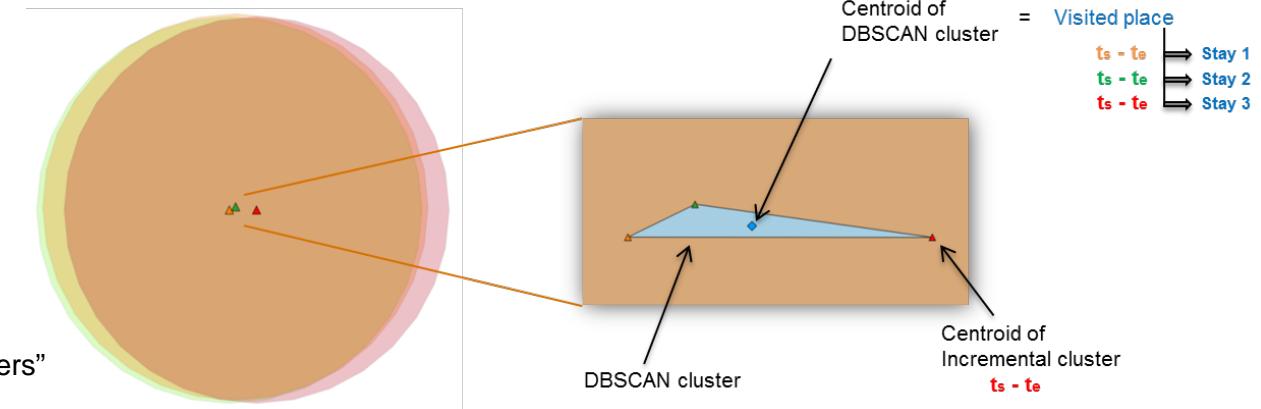
1st Algorithm (Ye et al. 2009) + 2nd DBSCAN (Ester et al. 1996)

Fuzziness of locations



Authors perform a second clustering of the initial **stay points** detected → **DBSCAN**

Final visited place will be the
“cluster of the incremental clusters”

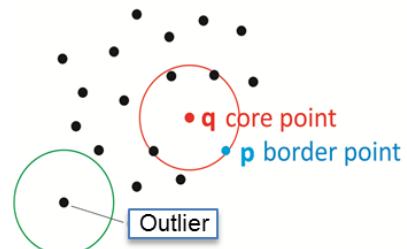


2. Theory

Clustering algorithms

➤ 3. Density-based clustering *DBSCAN (Ester et al. 1996)*

- **For each point of a cluster**, the cardinality of the neighbourhood of a given radius (**Eps**) has to exceed a threshold (**MinPts**).
i.e. Within an Eps radius, a minimum number of points should be present

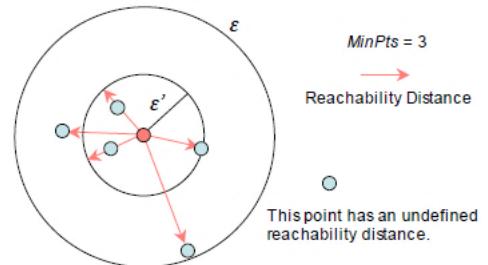


DBSCAN

To determine optimum **Eps** and **MinPts** → auxiliar algorithm

Generalizes DBSCAN creating an ordering of the points

- Allows extraction of clusters with **arbitrary values for ε (Eps)**
- A „maximum Eps“ (ϵ) to consider is selected as input
- Calculates **reachability-distance** of every point in dataset



OPTICS cluster (Ankerst et al. 1999)

(Ankerst et al. 1999)

OPTICS

- Does not generate a unique clustering, but
- Helps choosing optimal **Eps**

2. Theory

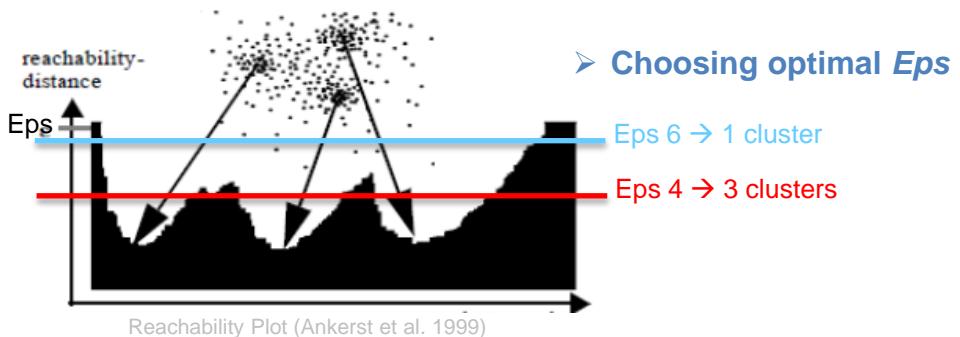
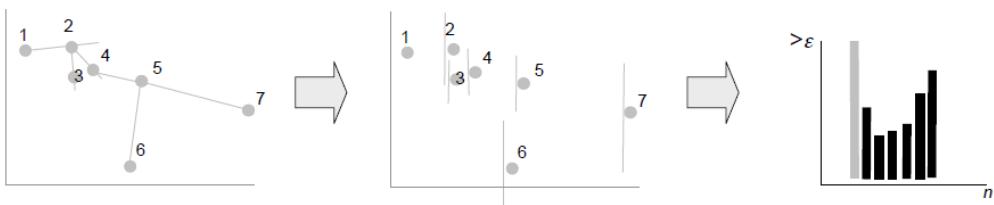
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Clustering algorithms

➤ 3. Density-based clustering DBSCAN (Ester et al. 1996)

Reachability plot

- Bar chart that shows each point's reachability distance in the order the object was processed
- Clearly show cluster structure of the data



2. Theory

Quality Evaluation (Salzburg Research 2015)

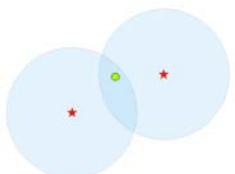
➤ Quality measures

SPATIAL

1. Spatial accuracy (Qsa)

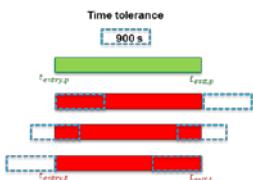


2. Spatial uniqueness (Qsu)



TEMPORAL

3. Temporal accuracy (Qta)



4. Amount of temporal incorrectness (Qt)

$$Q_{ti} := 1 - \frac{N_{incorr}}{D} = \frac{N_{corr}}{D}$$

Detected place = Clusters generated by algorithm
Tagged places = Real locations visited by user (GTD)

➤ Confusion matrix

		Predicted	Positive	Negative
Actual	Predicted	TP	FN	
	Positive	FP	TN	
Positive				
Negative				

CLASSES

1. True positive (TP). A tagged (real) place is detected.
2. False negative (FN). A tagged (real) place is not detected.
3. False positive (FP). A detection is obtained where there is no tagged (real) place.

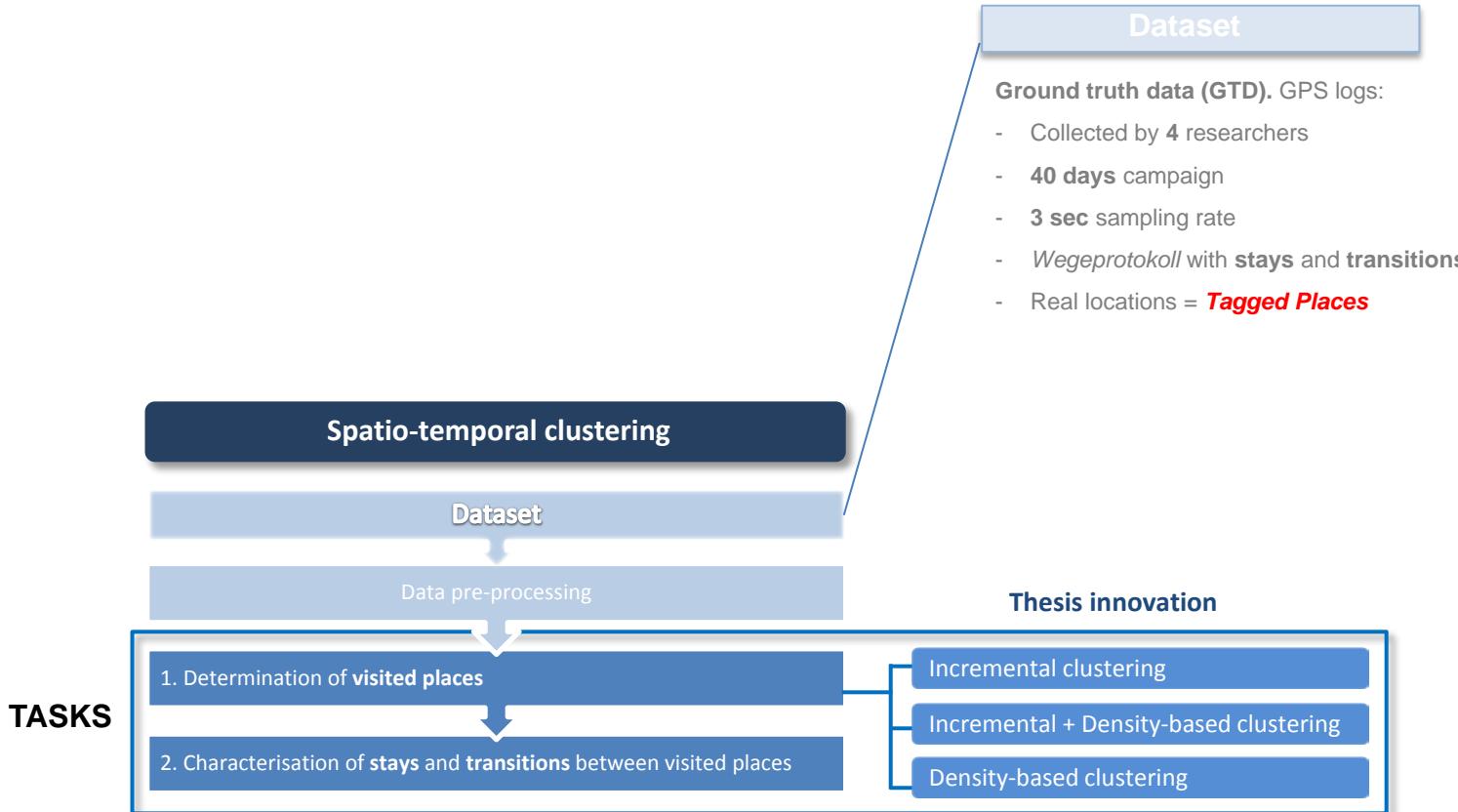
$$Precision = \frac{TP}{TP + FP} \quad \xrightarrow{\hspace{1cm}} \text{Exactness of the model}$$

$$Recall = \frac{TP}{TP + FN} \quad \xrightarrow{\hspace{1cm}} \text{Completeness of the model}$$

$$F \text{ measure} = \frac{2 * (Precision * Recall)}{Precision + Recall} \quad \xrightarrow{\hspace{1cm}} \text{Effectiveness of retrieval}$$

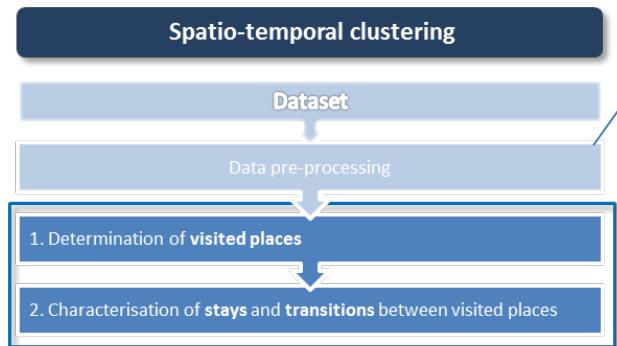
3. Method

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3. Method

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Data pre-processing

Dealing with **systematic** and **random** positioning errors

Filtering

- Removal of points with **equal timestamp**
- Removal of points with **equal geometry**
- Correction of **tunnels**
- Removal of **spikes**

Smoothing

- Kernel based approach



1. Determination of visited places

1.1. Incremental clustering

(Kang et al. 2005)

1.2. Incremental + Density-based clustering

(Ye et al. 2009) + DBSCAN (Ester et al. 1996)

1.3. Density-based clustering

DBSCAN (Ester et al. 1996)

1.4. Quality Evaluation

Implementation (Salzburg Research 2015)

2. Characterization of stays and transitions

2.1. Extraction of stays at visited places

- Process for stays extraction

2.2. Extraction of transitions between visited places

- Process for transitions extraction

2.3. Quality Evaluation of the extraction

- Confusion matrix
- Comparison of different parameter settings

2.4. Characterization of stays and transitions

- User profile and visualizations

3. Method

- 1. Introduction
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1. Determination of visited places

GOALS

- Determining **visited places** in a user's daily life
- **Evaluation** of clustering **algorithms**



RESULTS

-  - **Clusters** representing visited places
-  - **Comparison** of clustering algorithms **performance**
-  - **Assessment** of the algorithms and selection of the **best**

TARGETS

Quality of the clustering

- SPATIAL**
 - Precision > 66%**
 - Recall > 66%**
- TEMPORAL**
 - Times detected > 66%**



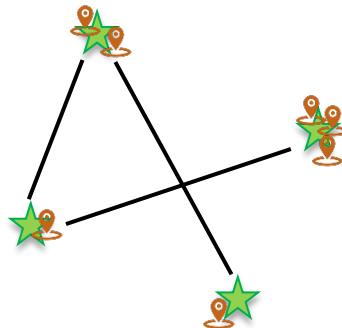
3. Method

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2. Characterization of stays and transitions

GOALS

- Representing **stays** at places with characteristic values
- Representing **transitions** with characteristic values
- Evaluating **time extraction** performed by spatially best algorithm



RESULTS

- User's dwell time at significant locations (**stays**)
- Transitions between significant locations (**transitions**)
- **Evaluation** of the stays and transitions **extraction**



TARGETS

Quality of the extraction

 **STAYS** detection

 **TRANSITIONS** detection

Precision > 66%

Recall > 66%

Precision > 66%

Recall > 66%

4. Implementation

- 1. Introduction
- 2. Theory
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1.1. Incremental clustering (Kang et al. 2005)

➤ Parameters

- Reference values

d = (30 / 50 m)

t = (300 / 1800 sec)

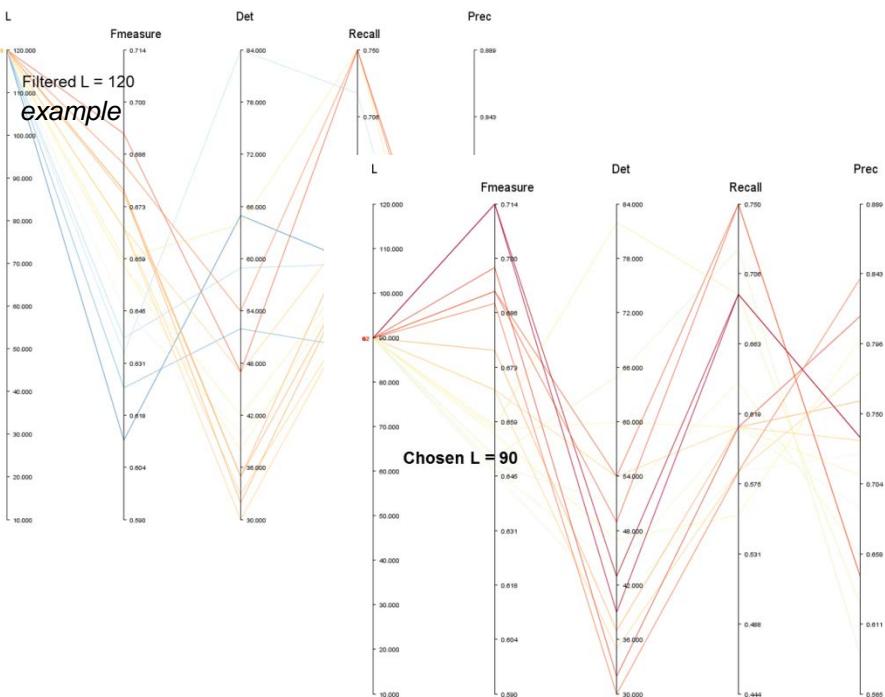
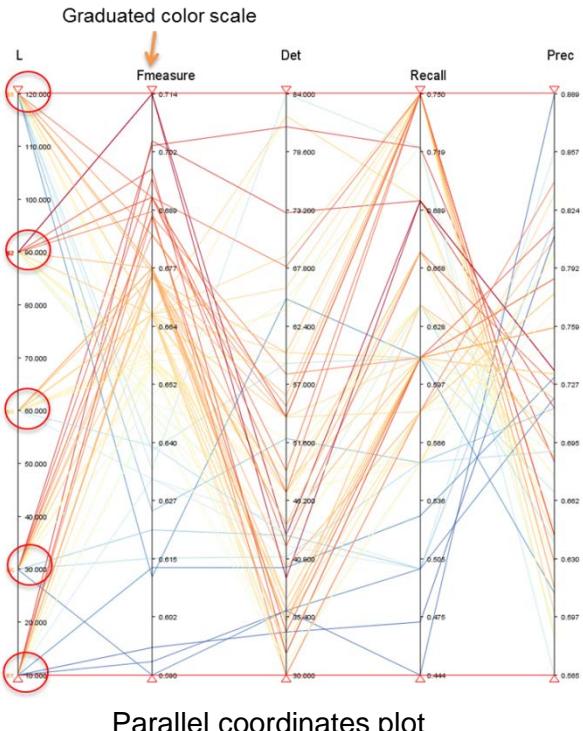
L = (no reference value)



Tested values:

d (m)	t (s)	L (s)
20	600	10
30	900	30
40	1200	60
50	1500	90
		120

80 combinations



4. Implementation

- 1. Introduction
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1.1. Incremental clustering (*Kang et al. 2005*)

➤ Parameters

- Tested values

L = 90 sec

d (m)	t (s)
20	300
26.5	600
30	900
40	1200
53	1500
60	1800
70	2100
80	
90	
100	
200	

11 x 7

77 combinations

- Implemented additional Java class for **batch processing**

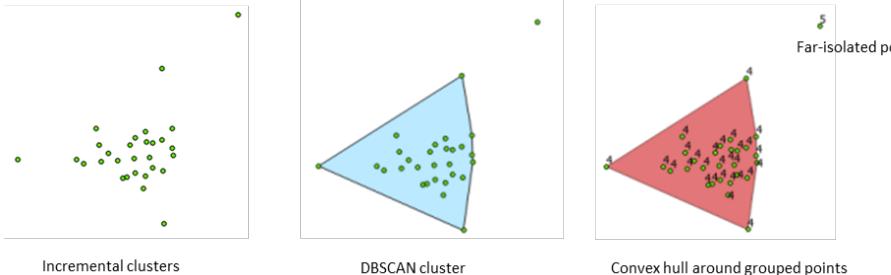
4. Implementation

- 1. Introduction
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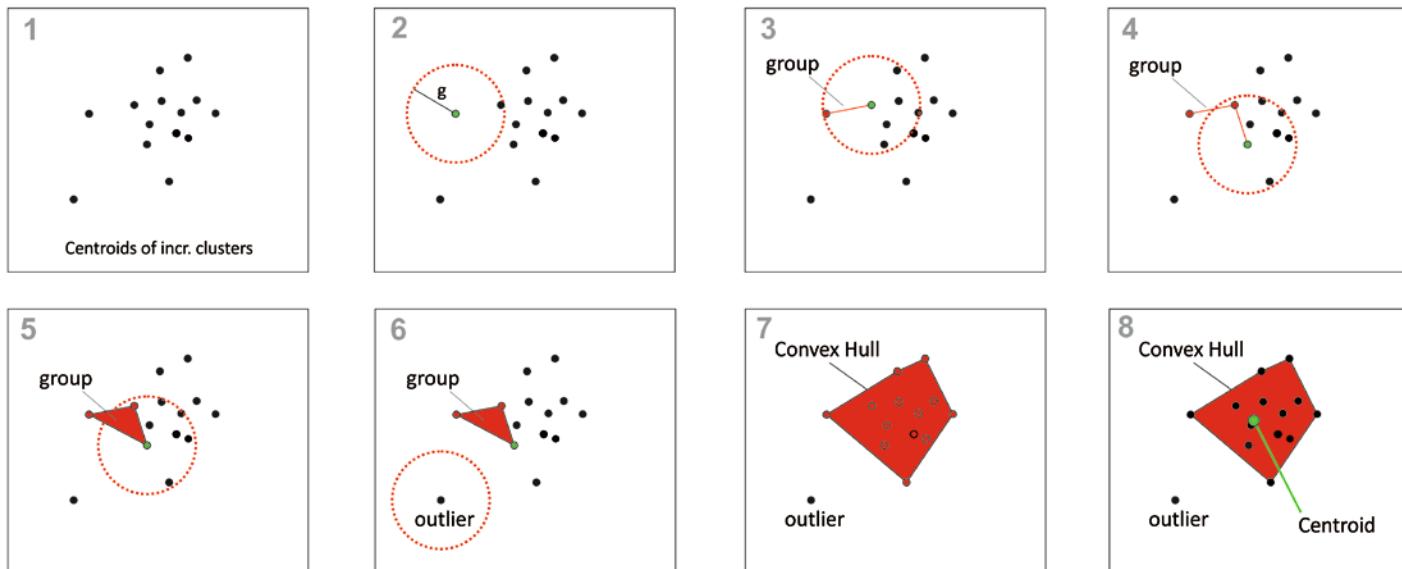
1.2. Incremental clustering + density-based clustering

1st Algorithm (Ye et al. 2009) + 2nd DBSCAN (Ester et al. 1996)
+ 2nd ConvexHull

- Alternative to DBSCAN
- Seamless integration within our process
- Almost same performance than DBSCAN
- Own solution



Grouping of points → Creation of convex hull → Centroid calculation



4. Implementation

- 1. Introduction
- 2. Theory
- 3. Method
- **4. Implementation**
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1.2. Incremental clustering + density-based clustering

1st Algorithm (Ye et al. 2009) + ~~2nd DBSCAN (Ester et al. 1996)~~
+ *2nd ConvexHull*

➤ Parameters

- Reference values

d = (200 meters)

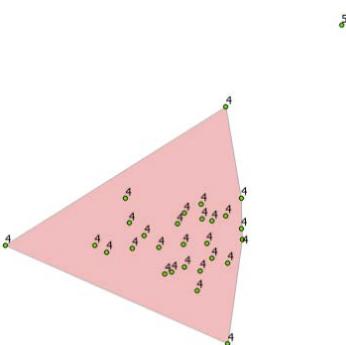
t = (1800 seconds)

g = (40 meters) ← ConvexHull ← DBSCAN ← OPTICS

- Tested values

d (m)	t (s)
26.5	300
53	600
100	900
200	1200
300	1800
400	2400
500	3000
600	
700	

g (m)
40



9 x 7

63 combinations

➤ Implemented additional Java class for **batch processing**

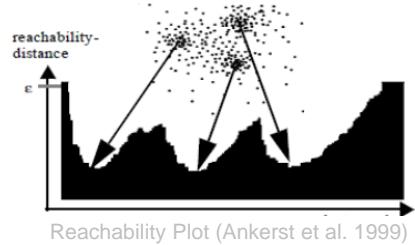
4. Implementation

1.3. Density-based clustering DBSCAN (Ester et al. 1996)

To determine optimal parameters: **Eps** and **MinPts** → OPTICS (Ankerst et al. 1999)

ELKI
 Environment for DeveLoping KDD-Applications
Supported by Index-Structures

Java software used for
OPTICS/DBSCAN clustering



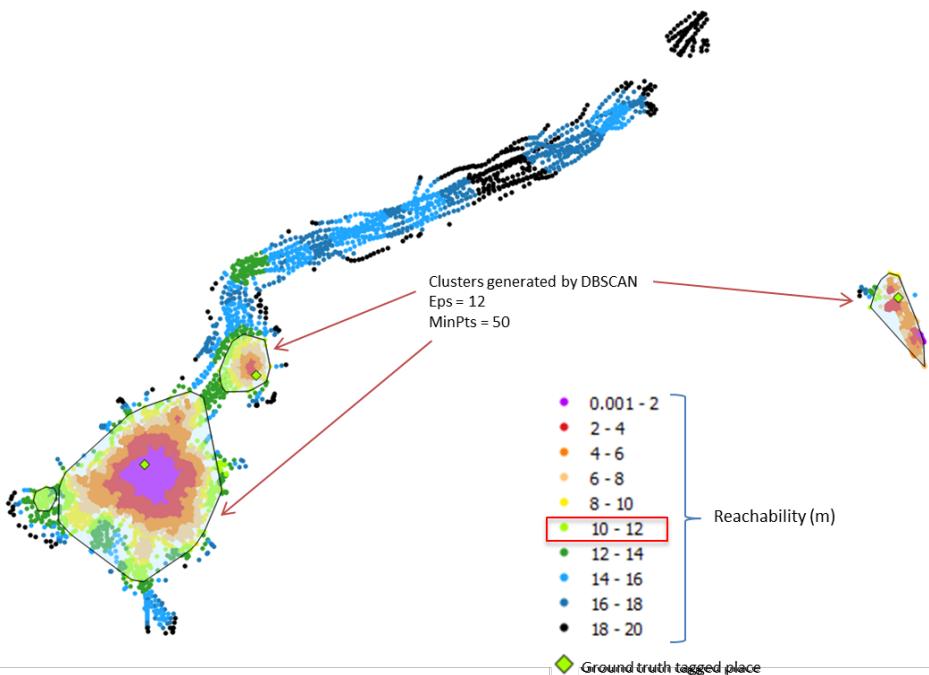
Reachability plot for 450.000 points → **not operative**

Alternative **visual approach** with QGIS



QGIS

- Represented **points** with their **reachability** values
- Represented **ground truth places**
 - Chosen reachability determines clusters to be obtained
 - Visual check of the whole dataset



DBSCAN clustering with determined **Eps**

4. Implementation

- 1. Introduction
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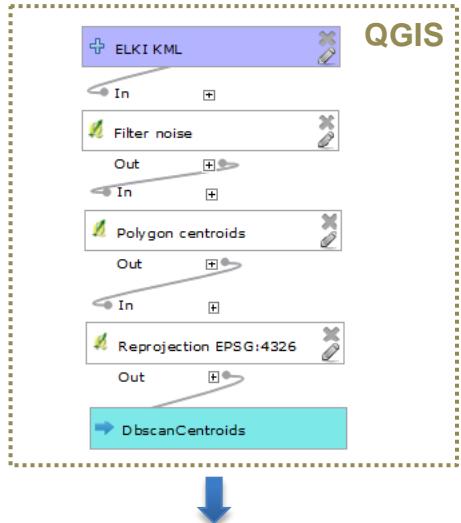
1.3. Density-based clustering DBSCAN (Ester et al. 1996)

- Tested values

MinPts	Eps (m)
20	2
30	3
40	6
50	9
60	12
70	15
80	18
90	
100	
110	
120	
11	x
11	7

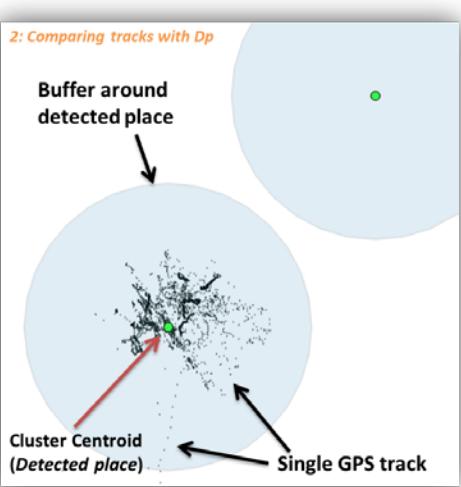
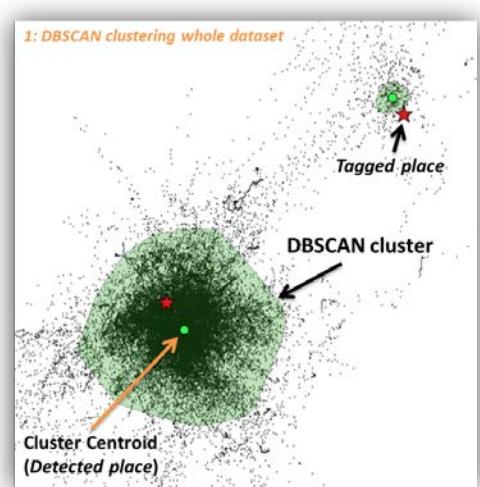
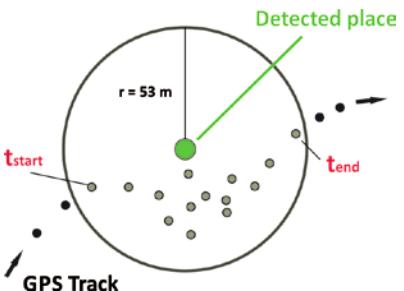


77 combinations



Java

Implemented class for dwell time extraction and batch processing



4. Implementation

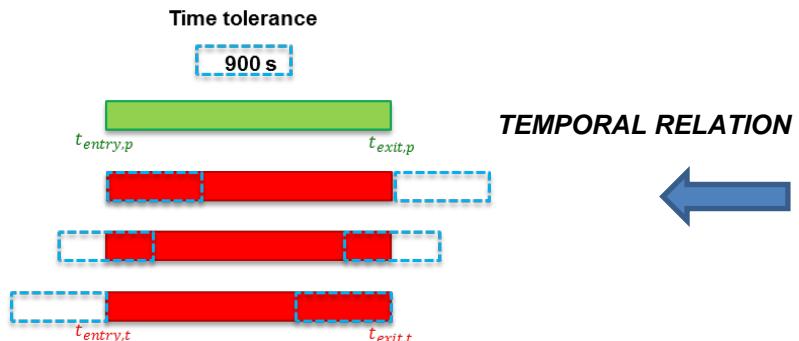
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1.4. Quality Evaluation

- Developed at Salzburg Research → Mostly implemented within this thesis (Java)
- Applied on 3 spatio-temporal clustering approaches
- Spatial and temporal component of detecting the GTD tagged places
- Measures to evaluate spatial and temporal accuracy of the estimations
- Performance compared in a confusion matrix

Detected place = Clusters generated by algorithm
Tagged places = Real locations visited by user (GTD)

- Detected places related to tagged places for each test user



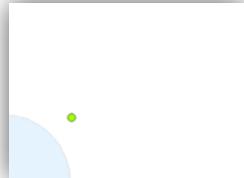
Mean time deviation calculated for each tP

$$\Delta t_{t,p} = \frac{|(t_{entry,p} - t_{entry,t})| + |(t_{exit,p} - t_{exit,t})|}{2}$$

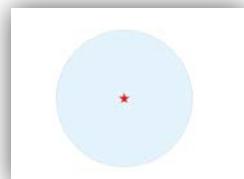
SPATIAL RELATION



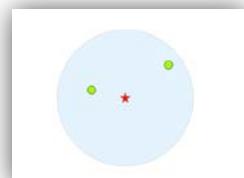
Circular buffer around
tagged places



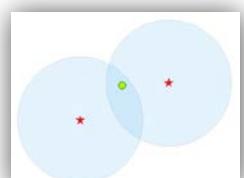
1. Det. without tagged place



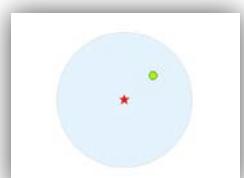
2. Tagged place without det.



3. Multiple detections



4. Mult. tagged places

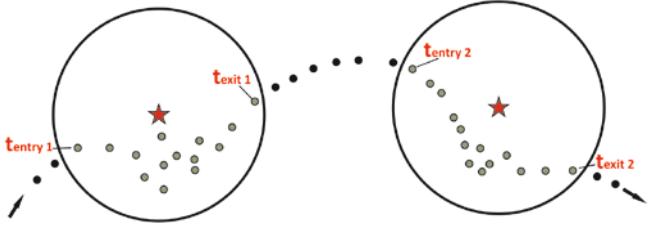


5. Det. with tagged place

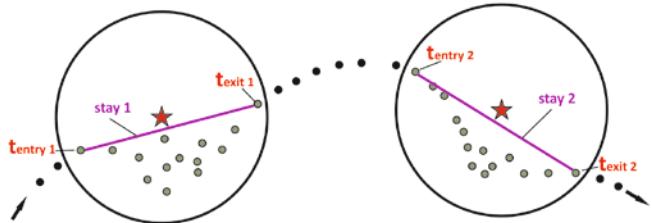
4. Implementation

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2.1. Extraction of stays // 2.2. Extraction of transitions



STAYS



- Stored in each detected place
- Duration = texit - tentry

```
// Cluster duration
long currentClusterDuration = calculateDurationOfList(currentCluster);

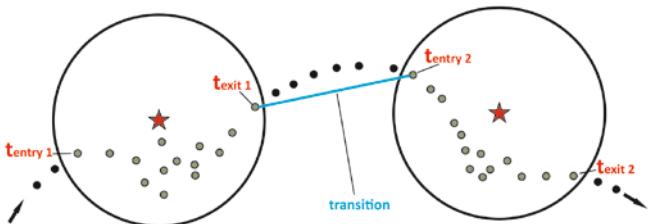
if (currentClusterDuration >= timeParam) {
    boolean withinInDistance = false;
    int lastInCluster = 0;
    lastInCluster = currentCluster.size() - 1;

    Date timeEntry = currentCluster.get(0).getTimestamp();
    Date timeExit = currentCluster.get(lastInCluster).getTimestamp();
    TimeElement timeElement = new TimeElement(currentClusterCentroid, timeEntry, timeExit);
    Transition transition = null;

    for (DetectedPlace dp : detectedPlaces) {
        // Calculate distance between currentCluster and each already detected DetectedPlace
        double clusterToDpDistance = calculateDistance(dp.getLocation().getCoordinate(), cu...
```

DpID	VisitID	TimeStart	TimeEnd	Duration(s)	Month	Day	WD
1	0	12/08/2014 16:27	12/08/2014 17:48	4904	8	12	Tue
1	1	13/08/2014 16:28	13/08/2014 17:10	2488	8	13	Wed
1	2	13/08/2014 17:45	13/08/2014 23:59	22450	8	13	Wed
1	2	14/08/2014 00:00	14/08/2014 08:16	29814	8	14	Thu
1	3	14/08/2014 16:14	14/08/2014 16:32	1061	8	14	Thu
1	4	17/08/2014 18:32	17/08/2014 23:59	19659	8	17	Sun
1	4	18/08/2014 00:00	18/08/2014 08:04	29042	8	18	Mon
1	5	18/08/2014 16:18	18/08/2014 16:53	2104	8	18	Mon
1	6	18/08/2014 17:41	18/08/2014 18:28	2819	8	18	Mon
1	7	19/08/2014 16:01	19/08/2014 16:37	2165	8	19	Tue
1	8	19/08/2014 18:09	19/08/2014 18:46	2231	8	19	Tue

TRANSITIONS



- Stored with associated data
- Duration = tentry 2 - texit 1

TriID	Orig.	Dest.	TimeDeparture	TimeArrival	Distance(m)	Duration(s)	Sp(km/h)	Mon	Day	WD	SameDay
1	1	2	12/08/2014 17:48	12/08/2014 17:55	441.0	405	3.920	8	12	Tue	TRUE
2	2	3	12/08/2014 20:12	13/08/2014 08:17	490.8	43477	0.041	8	12	Tue	FALSE
3	3	1	13/08/2014 16:01	13/08/2014 16:28	320.4	1623	0.711	8	13	Wed	TRUE
4	1	1	13/08/2014 17:10	13/08/2014 17:45	0.0	2131	0.000	8	13	Wed	TRUE
5	1	3	14/08/2014 08:16	14/08/2014 08:23	320.4	423	2.726	8	14	Thu	TRUE
6	3	1	14/08/2014 16:10	14/08/2014 16:14	320.4	264	4.368	8	14	Thu	TRUE
7	1	4	14/08/2014 16:32	14/08/2014 18:47	45377.3	8107	20.150	8	14	Thu	TRUE
8	4	5	14/08/2014 19:28	14/08/2014 19:29	113.6	96	4.260	8	14	Thu	TRUE
9	5	6	15/08/2014 09:59	15/08/2014 10:03	860.0	192	16.125	8	15	Fri	TRUE
10	6	5	15/08/2014 11:59	15/08/2014 12:02	860.0	195	15.877	8	15	Fri	TRUE

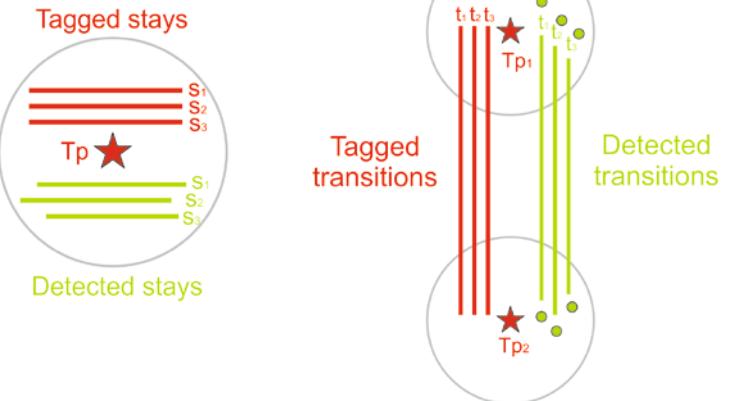
4. Implementation

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2.3. Quality Evaluation of stays and transitions extraction

$$\bar{t}_{t,d} = \frac{|(t_{entry,d} - t_{entry,t})| + |(t_{exit,d} - t_{exit,t})|}{2}$$

if $\bar{t}_{t,d} \leq 900 \text{ sec} \rightarrow \text{Matches}$



➤ Proportion of time extracted

$$St_{ext} := \frac{St_d}{St_t} \quad \begin{array}{l} \text{Total time in stays detected} \\ \text{Total time in stays tagged (GTD)} \end{array}$$

$$Tt_{ext} := \frac{Tt_d}{Tt_t} \quad \begin{array}{l} \text{Total time in transitions detected} \\ \text{Total time in transitions tagged (GTD)} \end{array}$$

➤ Confusion matrix classes

- | | |
|------------------------------|---|
| - True positive (TP) | A tagged stay/transition is detected |
| - False negative (FN) | A tagged stay/transition is not detected |
| - False positive (FP) | A stay/transition is obtained when there is no tagged stay/transition |

2.4. Characterization of stays and transitions

➤ Selection of best clustering algorithm

- Selection according to maximum F measure
 - Selection of best parameters settings for the best algorithm

➤ Data mining of algorithm outputs

- Creation of SQLite database
 - Design of SQL statements
 - Tests with 3 algorithms
 - Processing of data for 3D visualizations

➤ INDICATORS

- **Stays**

1. Number of stays at each tagged place per weekday

- 1.1. Number of detected occurrences
 - 1.2. Number of GTD occurrences (tagged)
 - 1.3. Proportion of GTD stays detected

- **Transitions**

1. Number of transitions between tagged places per weekday

- 1.1. Number of detected occurrences
 - 1.2. Number of GTD occurrences (tagged)
 - 1.3. Proportion of GTD transitions detected

2. Duration of the stays at each tagged place per weekday

- 2.1. Duration of detected stays
 - 2.2. Duration of GTD stays
 - 2.3. Proportion of GTD stays duration detected

2. Duration of the transitions between tagged places per weekday

- 2.1. Duration of detected transitions
 - 2.2. Duration of GTD transitions
 - 2.3. Proportion of GTD transitions duration detected

Stays duration at each tagged place for each weekday																											
Detected total duration (h)							GTD total duration (h)							Proportion of GTD total duration detected													
TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	PROP	
1	23.48	22.02	31.85	13.48	9.10	0.00	0.00	99.93	1	36.73	22.52	31.85	29.77	15.47	0.00	0.00	136.3	2	1	0.64	0.98	1.00	0.45	0.59		0.73	
2	0.00	2.27	0.00	0.00	0.00	0.00	0.00	2.27	2	0.50	2.27	0.00	0.00	0.00	0.00	0.00	2.77	3	2	0.00	1.00					0.82	
3	23.48	17.97	40.55	42.20	28.68	11.78	30.35	195.0	3	76.23	65.90	67.03	60.38	36.58	12.28	47.70	366.1	6	3	0.31	0.27	0.60	0.70	0.78	0.96	0.64	
4	0.00	0.00	0.00	4.52	37.43	86.80	61.32	190.1	4	0.00	0.00	0.00	5.22	42.98	96.15	61.45	205.8	4	4			0.87	0.87	0.90	1.00	0.92	
5	0.00	0.00	0.00	0.00	0.93	0.00	0.00	0.93	5	0.00	0.00	0.00	0.10	0.93	3.72	1.60	6.35	5	5			0.00	1.00	0.00	0.00	0.15	
6	0.00	0.00	0.00	0.00	0.65	0.00	0.00	0.65	6	0.00	0.00	0.00	0.65	0.00	0.00	0.00	0.65	6	6			1.00				1.00	
7	0.00	0.00	0.00	0.00	5.07	3.13	5.80	14.00	7	0.00	0.00	0.00	0.00	5.07	5.85	5.80	16.72	7	7			1.00	0.54	1.00		0.84	
8	0.00	0.00	0.00	0.00	0.00	0.00	0.83	0.83	8	0.00	0.00	0.00	0.00	0.00	0.00	0.83	0.83	8	8					1.00		1.00	
9	0.00	0.00	0.00	0.00	0.00	0.00	1.52	1.52	9	0.00	0.00	0.00	0.00	0.00	0.00	1.52	1.52	9	9					1.00		1.00	
10	0.00	0.00	0.00	0.00	1.93	0.00	0.00	1.93	10	0.00	0.00	0.00	0.00	1.93	0.00	0.00	1.93	10	10			1.00				1.00	

5. Results

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Output tables (example)

Qsa		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.471	0.499	0.493	0.506	0.507	0.501	0.490	0.486	0.473	0.461	0.313		
600	0.411	0.438	0.432	0.469	0.475	0.470	0.455	0.450	0.445	0.421	0.325		
900	0.380	0.393	0.396	0.439	0.443	0.444	0.437	0.429	0.432	0.407	0.313		
1200	0.355	0.357	0.360	0.386	0.395	0.399	0.398	0.394	0.397	0.376	0.286		
1500	0.335	0.328	0.338	0.357	0.366	0.372	0.384	0.378	0.379	0.367	0.305		
1800	0.306	0.326	0.336	0.351	0.353	0.362	0.381	0.373	0.369	0.353	0.302		
2100	0.295	0.303	0.317	0.344	0.339	0.358	0.368	0.361	0.367	0.352	0.295		

Qsu		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.992	0.996	0.996	1.000	0.996	1.000	0.996	0.996	0.996	0.998	0.994		
600	0.987	0.995	0.995	1.000	1.000	1.000	0.994	0.994	0.994	0.991	0.995		
900	1.000	1.000	0.993	1.000	1.000	1.000	0.993	0.993	0.993	0.995	0.993		
1200	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.994	0.991	
1500	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.994	1.000	
1800	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.993	1.000	
2100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.993	1.000	

Qta		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.556	0.512	0.522	0.542	0.574	0.590	0.611	0.606	0.577	0.572	0.386		
600	0.518	0.512	0.506	0.579	0.624	0.648	0.652	0.643	0.614	0.615	0.411		
900	0.531	0.516	0.503	0.543	0.570	0.612	0.624	0.620	0.589	0.580	0.423		
1200	0.529	0.514	0.493	0.566	0.593	0.619	0.629	0.621	0.593	0.586	0.385		
1500	0.521	0.516	0.523	0.599	0.583	0.622	0.641	0.616	0.594	0.598	0.419		
1800	0.525	0.564	0.556	0.615	0.570	0.621	0.656	0.633	0.600	0.597	0.424		
2100	0.526	0.596	0.558	0.616	0.582	0.633	0.659	0.640	0.602	0.603	0.416		

Qti		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.452	0.551	0.612	0.664	0.685	0.719	0.788	0.785	0.792	0.817	0.831		
600	0.492	0.585	0.669	0.667	0.728	0.748	0.784	0.789	0.792	0.837			
900	0.549	0.608	0.649	0.705	0.713	0.751	0.777	0.782	0.775	0.808	0.838		
1200	0.581	0.640	0.651	0.739	0.739	0.770	0.778	0.791	0.791	0.813	0.825		
1500	0.570	0.636	0.645	0.737	0.745	0.763	0.774	0.774	0.784	0.798	0.804		
1800	0.602	0.636	0.651	0.729	0.745	0.759	0.770	0.774	0.803	0.809	0.805		
2100	0.602	0.655	0.671	0.743	0.776	0.772	0.775	0.773	0.783	0.792	0.834		

Recall		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.697	0.737	0.745	0.731	0.728	0.728	0.719	0.720	0.698	0.658	0.503		
600	0.586	0.617	0.629	0.648	0.649	0.648	0.644	0.654	0.645	0.603	0.488		
900	0.514	0.522	0.548	0.577	0.581	0.581	0.576	0.597	0.589	0.561	0.453		
1200	0.471	0.491	0.499	0.513	0.516	0.518	0.527	0.541	0.531	0.505	0.427		
1500	0.445	0.459	0.475	0.480	0.484	0.489	0.507	0.504	0.500	0.448			
1800	0.414	0.450	0.461	0.462	0.464	0.485	0.489	0.497	0.484	0.473	0.437		
2100	0.399	0.416	0.435	0.449	0.437	0.467	0.472	0.484	0.463	0.460	0.425		

Precision		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.441	0.459	0.470	0.428	0.416	0.413	0.407	0.392	0.372	0.351	0.285		
600	0.624	0.591	0.610	0.594	0.579	0.585	0.552	0.559	0.556	0.517	0.416		
900	0.728	0.699	0.722	0.695	0.673	0.661	0.641	0.671	0.664	0.636	0.489		
1200	0.750	0.727	0.731	0.736	0.724	0.725	0.698	0.706	0.695	0.673	0.590		
1500	0.803	0.763	0.783	0.776	0.745	0.762	0.754	0.749	0.736	0.704	0.657		
1800	0.828	0.785	0.810	0.790	0.775	0.786	0.757	0.777	0.760	0.744	0.674		
2100	0.846	0.772	0.824	0.802	0.787	0.795	0.764	0.786	0.778	0.757	0.695		

Fmeasure		Distance d (m)											
Time t (s)		20	26.5	30	40	53	60	70	80	90	100	200	
300	0.539	0.563	0.574	0.538	0.527	0.525	0.518	0.506	0.484	0.457	0.360		
600	0.600	0.601	0.618	0.619	0.611	0.614	0.594	0.602	0.596	0.554	0.446		
900	0.600	0.593	0.620	0.628	0.621	0.617	0.604	0.629	0.621	0.593	0.467		
1200	0.574	0.582	0.588	0.602	0.600	0.601	0.598	0.609	0.598	0.574	0.493		
1500	0.568	0.568	0.587	0.591	0.583	0.592	0.596	0.601	0.593	0.580	0.528		
1800	0.548	0.564	0.583	0.581	0.578	0.596	0.590	0.602	0.588	0.574	0.524		
2100	0.538	0.534	0.564	0.572	0.557	0.584	0.578	0.593	0.574	0.567	0.520		

- 1. Averaged results from 4 users → **Algorithms assessment**
- 2. Selection of **best individual dataset** (*User1*)
- 3. Selection of **parameter settings** generating highest **F measure**
- 4. **Characterization of stays and transitions** from *User1*

5. Results

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1. Determination of visited places

Algorithms assessment

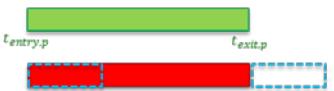
Algorithm	KANG		YE		DBSCAN	
	Parameters	Value	Parameters	Value	Parameters	Value
Recall	30 / 300	0.745	100 / 300	0.718	6 / 20	0.770
Precision	20 / 2100	0.846	26.5 / 3000	0.781	2 / 120	0.663
Fmeasure	80 / 900	0.629	100 / 1200	0.575	18 / 110	0.561
Qsa	53 / 300	0.507	53 / 300	0.490	9 / 20	0.493
Qsu	Multi	1.000	Multi	1.000	Multi	1.000
Qta	70 / 2100	0.659	26.5 / 3000	0.691	2 / 80	0.852
Qt	200 / 900	0.838	200 / 3000	0.796	15 / 120	0.589

- **Absolute temporal** performance highly dependant on **spatial** performance
- Without spatial assignment there is no possible temporal assignment

➤ Temporal performance

- Maximum Qta: **DBSCAN**

Accuracy of times detected



- Maximum Qt: **Kang**

83.8% of times detected are correct

➤ Spatial performance

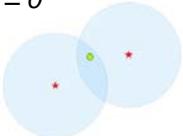
- Maximum Qsa: **Kang**

Spatial accuracy of detections (distance/number)



- Maximum Qsu: **ALL**

Cases of detections within overlapping areas = 0



Performance of the clustering



Confusion matrix metrics



F measure

5. Results

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1. Determination of visited places

➤ Clustering performance

Best F measure and associated measures			
Algorithm	KANG	YE	DBSCAN
Parameters (d/t)	80 / 900	100 / 1200	18 / 110
	Value	Value	Value
Recall	0.597	0.525	0.630
Precision	0.671	0.642	0.508
Fmeasure	0.629	0.575	0.561
Detections	47	45	72
Qsa	0.432	0.391	0.459
Qsu	0.993	1.000	0.994
Qta	0.589	0.537	0.621
Qt _i	0.775	0.671	0.515

- Best F measure: Kang
- Best individual results: Kang (User1)

$$F \text{ measure} = \frac{2 * (\text{Precision} * \text{Recall})}{\text{Precision} + \text{Recall}}$$

➤ Best individual performance (User 1)

Best F measure and associated measures			
Algorithm	KANG	YE	DBSCAN
Parameters	53 / 900	100 / 1200	6 / 120
	Value	Value	Value
Recall	0.722	0.639	0.667
Precision	0.765	0.719	0.800
Fmeasure	0.735	0.676	0.727
Detections	36	32	38
Qsa	0.554	0.505	0.530
Qsu	1.000	1.000	0.974
Qta	0.483	0.375	0.486
Qt _i	0.815	0.913	0.567



5. Results

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2. Characterization of stays and transitions

➤ Comparison of the extraction (User 1)

Results for the best param. setting for each algorithm				
Algorithm	KANG	YE	DBSCAN	
Parameters	53 / 900	100 / 1200	6 / 120	
	Value	Value	Value	
Recall	0.722	0.639	0.667	
Precision	0.765	0.719	0.800	
Fmeasure	0.735	0.676	0.727	
Detections	36	32	38	
Qsa	0.554	0.505	0.530	
Qsu	1.000	1.000	0.974	
Qta	0.483	0.375	0.486	
Qt _i	0.815	0.913	0.567	
StayT_ext	0.619	0.175	0.468	Stay and Transition time extracted
TranT_ext	0.493	0.295	0.373	
StRecall	0.600	0.317	0.450	Stay detection quality
StPrecision	0.684	0.471	0.600	
StFmeas	0.639	0.379	0.514	
TrRecall	0.568	0.331	0.374	Transition detection quality
TrPrecision	0.594	0.387	0.452	
TrFmeas	0.581	0.357	0.409	

Kang
Best performance

5. Results

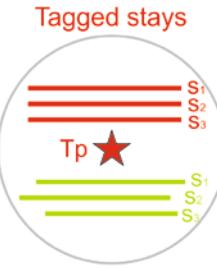
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2. Characterization of stays and transitions

Kang (User1) Example of movement behaviour profiling

Number of stays at each tP for each week day																												
Detected Occurrences													GTD Occurrences							Proportion of GTD Occurrences detected								
TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	PROP		
1	3	4	4	2	2	0	0	15	1	8	5	4	6	6	0	0	29	1	0.38	0.80	1.00	0.33	0.33	0.52	Work			
2	0	1	0	0	0	0	0	1	2	1	1	0	0	0	0	0	2	2	0.00	1.00	0.50							
3	4	3	6	8	7	1	4	33	3	11	10	11	12	8	2	7	61	3	0.36	0.30	0.55	0.67	0.88	0.50	0.57	0.54	Home1	
4	0	0	0	1	10	14	12	37	4	0	0	0	0	3	11	16	13	43	4	0.33	0.91	0.88	0.92	0.86	0.52	0.54	0.50	Home2
5	0	0	0	0	1	0	0	1	5	0	0	0	0	1	1	1	1	4	5	0.00	1.00	0.00	0.00	0.00	0.25	0.25	0.25	
6	0	0	0	1	0	0	0	1	6	0	0	0	0	1	0	0	0	1	6	1.00	0.00	1.00	0.00	0.00	1.00	1.00	1.00	
7	0	0	0	0	0	2	1	2	7	0	0	0	0	0	2	2	2	6	7	0.00	1.00	0.50	1.00	0.83	0.83	0.83	0.83	
8	0	0	0	0	0	0	1	1	8	0	0	0	0	0	0	0	1	1	8	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	
9	0	0	0	0	0	0	0	1	9	0	0	0	0	0	0	0	1	9	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
10	0	0	0	1	0	0	0	1	10	0	0	0	1	0	0	0	1	10	0.00	1.00	0.00	1.00	1.00	1.00	1.00	1.00		
11	0	1	0	0	0	0	0	1	11	0	1	1	1	0	0	0	3	11	0.00	0.00	0.00	0.00	0.00	0.33	0.33	0.33		
12	0	0	1	0	0	0	0	1	12	0	0	1	0	0	0	0	1	12	1.00	0.00	1.00	0.00	0.00	1.00	1.00	1.00		
13	0	0	0	0	1	0	0	1	13	0	0	0	0	0	1	0	0	13	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
14	0	0	0	0	0	0	0	0	14	0	0	0	0	0	1	0	0	14	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
15	0	0	0	0	0	1	0	1	15	0	0	0	0	0	0	1	0	15	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
16	0	0	0	0	0	1	0	1	16	0	0	0	0	0	0	1	0	16	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
17	0	0	0	0	0	1	0	0	17	0	0	0	0	0	0	1	0	17	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
18	0	0	0	0	0	0	1	1	18	0	0	0	0	0	0	0	1	18	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
19	0	0	0	0	0	0	1	1	19	0	0	0	0	0	0	0	1	19	0.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00		
20	1	0	0	0	0	0	0	1	20	1	0	0	0	0	0	0	0	20	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
21	0	0	0	0	0	0	0	0	21	1	0	0	0	0	0	0	0	21	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
22	0	0	0	0	0	0	0	0	22	1	0	0	0	0	0	0	0	22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
23	0	0	0	0	0	0	0	0	23	1	3	2	2	0	0	0	8	23	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
24	0	0	0	0	0	0	0	0	24	0	1	0	0	0	0	0	1	24	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
25	0	0	0	0	0	0	0	0	25	0	1	1	1	0	0	0	3	25	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
26	0	0	0	0	0	0	0	0	26	0	0	1	0	0	0	0	1	26	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
27	0	0	0	0	0	0	1	1	27	0	0	0	0	0	0	1	1	27	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
28	0	0	0	0	0	0	1	1	28	0	0	0	0	0	0	1	1	28	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
29	0	0	0	0	0	1	0	1	29	0	0	0	0	0	1	0	1	29	1.00	0.00	1.00	0.00	0.00	0.00	0.00	0.00		
30	0	0	0	0	0	0	0	0	30	0	0	0	0	0	0	1	0	30	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
SUM	8	9	11	13	23	20	24	108	SUM	24	22	21	28	30	26	29	180	PROP	0.33	0.41	0.52	0.46	0.77	0.77	0.83	0.60		

3 most stayed places
>83% time



Mondays worst detection
Weekends best detection

5. Results

- 1. Introduction
- 2. Theory
- 3. Method
- 4. Implementation
- 5. Results
- 6. Conclusions

2. Characterization of stays and transitions

Kang (User1) Example of movement behaviour profiling

Stays duration at each tagged place for each weekday																											
Detected total duration (h)								GTD total duration (h)				Proportion of GTD total duration detected															
TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	SUM	TpID	Mon	Tue	Wed	Thu	Fri	Sat	Sun	PROP	
1	23.48	22.02	31.85	13.48	9.10	0.00	0.00	99.93	1	36.73	22.52	31.85	29.77	15.47	0.00	0.00	136.3	1	0.64	0.98	1.00	0.45	0.59	0.73	0.73	Work	
2	0.00	2.27	0.00	0.00	0.00	0.00	0.00	2.27	2	0.50	2.27	0.00	0.00	0.00	0.00	0.00	0.00	2.77	2	0.00	1.00	0.00	0.00	0.00	0.00	0.82	
3	23.48	17.97	40.55	42.20	28.68	11.78	30.35	195.0	3	76.23	65.90	67.03	60.38	36.58	12.28	47.70	366.1	3	0.31	0.27	0.60	0.70	0.78	0.96	0.64	0.53	Home1
4	0.00	0.00	0.00	4.52	37.43	86.80	61.32	190.1	4	0.00	0.00	0.00	5.22	42.98	96.15	61.45	205.8	4	0.00	0.00	0.00	0.87	0.87	0.90	1.00	0.92	Home2
5	0.00	0.00	0.00	0.00	0.93	0.00	0.00	0.93	5	0.00	0.00	0.00	0.10	0.93	3.72	1.60	6.35	5	0.00	1.00	0.00	0.00	0.00	0.00	0.15		
6	0.00	0.00	0.00	0.65	0.00	0.00	0.00	0.65	6	0.00	0.00	0.00	0.65	0.00	0.00	0.00	0.65	6	0.00	1.00	0.00	0.00	0.00	0.00	1.00		
7	0.00	0.00	0.00	0.00	5.07	3.13	5.80	14.00	7	0.00	0.00	0.00	0.00	5.07	5.85	5.80	16.72	7	0.00	0.54	1.00	0.84	0.00	0.00	0.00		
8	0.00	0.00	0.00	0.00	0.00	0.83	0.83	0.83	8	0.00	0.00	0.00	0.00	0.00	0.00	0.83	0.83	8	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
9	0.00	0.00	0.00	0.00	0.00	1.52	1.52	1.52	9	0.00	0.00	0.00	0.00	0.00	0.00	1.52	1.52	9	0.00	0.00	0.00	0.00	1.00	0.00	1.00		
10	0.00	0.00	0.00	1.93	0.00	0.00	0.00	1.93	10	0.00	0.00	0.00	1.93	0.00	0.00	0.00	1.93	10	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
11	0.00	8.70	0.00	0.00	0.00	0.00	0.00	8.70	11	0.00	8.70	9.20	9.38	0.00	0.00	0.00	27.28	11	0.00	0.00	0.00	0.00	0.00	0.00	0.32		
12	0.00	0.00	3.42	0.00	0.00	0.00	0.00	3.42	12	0.00	0.00	3.42	0.00	0.00	0.00	0.00	3.42	12	0.00	1.00	0.00	0.00	0.00	0.00	1.00		
13	0.00	0.00	0.00	0.00	1.40	0.00	0.00	1.40	13	0.00	0.00	0.00	0.00	1.40	0.00	0.00	1.40	13	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
14	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	14	0.00	0.00	0.00	0.00	0.50	0.00	0.00	0.50	14	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
15	0.00	0.00	0.00	0.00	1.00	0.00	0.00	1.00	15	0.00	0.00	0.00	0.00	0.00	1.00	0.00	1.00	15	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
16	0.00	0.00	0.00	0.00	0.70	0.00	0.00	0.70	16	0.00	0.00	0.00	0.00	0.00	0.70	0.00	0.70	16	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
17	0.00	0.00	0.00	0.00	2.48	0.00	0.00	2.48	17	0.00	0.00	0.00	0.00	0.00	2.48	0.00	2.48	17	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
18	0.00	0.00	0.00	0.00	0.00	0.17	0.17	0.17	18	0.00	0.00	0.00	0.00	0.00	0.17	0.00	0.17	18	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
19	0.00	0.00	0.00	0.00	0.32	0.00	0.32	0.32	19	0.00	0.00	0.00	0.00	0.00	0.32	0.00	0.32	19	0.00	0.00	0.00	1.00	0.00	0.00	1.00		
20	1.27	0.00	0.00	0.00	0.00	0.00	0.00	1.27	20	1.27	0.00	0.00	0.00	0.00	0.00	0.00	1.27	20	1.00	0.00	0.00	0.00	0.00	0.00	1.00		
21	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	21	0.80	0.00	0.00	0.00	0.00	0.00	0.00	0.80	21	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	22	0.22	0.00	0.00	0.00	0.00	0.00	0.00	0.22	22	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
23	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	23	9.87	14.37	10.88	10.58	0.00	0.00	0.00	45.70	23	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
24	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	24	0.00	4.78	0.00	0.00	0.00	0.00	0.00	4.78	24	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
25	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	25	0.00	4.28	8.32	7.55	0.00	0.00	0.00	20.15	25	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
26	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	26	0.00	0.00	3.13	0.00	0.00	0.00	0.00	3.13	26	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
27	0.00	0.00	0.00	0.00	0.00	1.83	1.83	1.83	27	0.00	0.00	0.00	0.00	0.00	1.83	1.83	1.83	27	0.00	0.00	0.00	1.00	1.00	1.00	1.00		
28	0.00	0.00	0.00	0.00	0.00	1.22	1.22	1.22	28	0.00	0.00	0.00	0.00	0.00	1.22	1.22	1.22	28	0.00	0.00	0.00	1.00	1.00	1.00	1.00		
29	0.00	0.00	0.00	0.00	0.73	0.00	0.73	0.73	29	0.00	0.00	0.00	0.00	0.73	0.00	0.73	0.73	29	0.00	0.00	0.00	1.00	1.00	1.00	1.00		
30	0.00	0.00	0.00	0.00	0.00	0.73	0.00	0.00	30	0.00	0.00	0.00	0.00	0.35	0.00	0.00	0.35	30	0.00	0.00	0.00	0.00	0.00	0.00	0.00		
	SUM	48.2	50.9	75.8	62.8	82.6	106.6	103.3	SUM	125.6	122.8	133.8	125.6	102.9	123.3	122.4	856.5	PROF	0.38	0.41	0.57	0.50	0.80	0.87	0.84	0.62	

3 most stayed places

>83% time

Tagged stays

Mondays worst detection

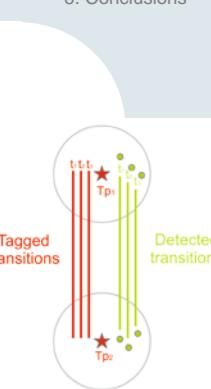
Weekends best detection

Kang (*User1*)

Transitions between tagged places for each weekday

Detected total duration (h)										GTD total duration (h)								Proportion of GTD total duration detected									
Tran	Mon	Tue	Wed	Thu	Fri	Sat	Sun		Tran	Mon	Tue	Wed	Thu	Fri	Sat	Sun		Tran	Mon	Tue	Wed	Thu	Fri	Sat	Sun	PROP	
01-01	0.00	0.15	0.00	0.00	0.00	0.00	0.00	0.15	01-01	0.25	0.15	0.00	0.28	0.18	0.00	0.00	0.87	01-01	0.00	1.00		0.00	0.00			0.17	
01-02	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	01-02	0.27	1.92	0.00	0.00	0.00	0.00	0.00	0.00	2.18	01-02	0.00	0.00						0.00
01-03	0.38	0.35	0.80	0.30	0.50	0.00	0.00	2.33	01-03	0.50	0.35	0.92	0.60	0.50	0.00	0.00	2.87	01-03	0.77	1.00	0.87	0.50	1.00			0.81	
01-31	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	01-31	0.00	0.00	0.00	0.10	0.00	0.00	0.00	0.10	01-31	0.00	0.00		0.00				0.00	
02-01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	02-01	2.60	0.00	0.00	0.00	0.00	0.00	0.00	2.60	02-01	0.00							0.00	
02-03	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	02-03	0.00	0.15	0.00	0.00	0.00	0.00	0.00	0.15	02-03	0.00							0.00	
03-01	0.42	0.12	0.13	0.38	0.52	0.00	0.00	1.57	03-01	0.72	0.35	0.47	0.53	0.52	0.00	0.00	2.58	03-01	0.58	0.33	0.29	0.72	1.00			0.61	
03-03	0.82	1.45	0.33	0.30	0.00	0.00	0.00	2.90	03-03	0.82	1.45	0.33	0.30	0.00	0.72	0.00	3.62	03-03	1.00	1.00	1.00	1.00	0.00			0.80	
03-04	0.00	0.00	0.00	0.00	2.82	0.00	0.00	2.82	03-04	0.00	0.00	0.00	1.25	4.95	0.00	0.00	6.20	03-04				0.00	0.57			0.45	
03-10	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	03-10	0.00	0.00	0.00	0.62	0.00	0.00	0.00	0.62	03-10				0.00				0.00	
03-11	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	03-11	0.00	0.32	0.12	0.28	0.00	0.00	0.00	0.72	03-11	0.00	0.00	0.00	0.00				0.00	
03-12	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	03-12	0.00	0.00	0.55	0.00	0.00	0.00	0.00	0.55	03-12				0.00				0.00	
03-13	0.00	0.00	0.00	0.00	1.53	0.00	0.00	1.53	03-13	0.00	0.00	0.00	0.00	1.53	0.00	0.00	1.53	03-13				1.00				1.00	
03-19	0.57	0.00	0.00	0.00	0.00	0.00	0.00	0.57	03-19	0.57	0.00	0.00	0.00	0.00	0.00	0.00	0.57	03-19	1.00							1.00	
03-26	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1.70	03-26	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1.70	03-26				1.00				1.00	
04-03	0.00	0.00	0.00	0.00	2.37	2.37	0.00	0.00	04-03	0.00	0.00	0.00	0.00	0.00	5.88	0.00	5.88	04-03				0.40	0.40			Home2 → Home1	
04-04	0.00	0.00	0.00	0.00	2.70	4.13	1.87	8.70	04-04	0.00	0.00	0.00	0.00	0.00	2.70	4.28	1.87	04-04	1.00	0.96	1.00	1.00	0.98			Home2 → Home2	
04-05	0.00	0.00	0.00	0.00	0.55	0.10	0.10	0.75	04-05	0.00	0.00	0.00	0.10	0.55	0.10	0.10	0.85	04-05				0.00	1.00	1.00	1.00	0.88	
04-06	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	04-06	0.00	0.00	0.00	0.05	0.00	0.00	0.00	0.05	04-06				0.00				0.00	
04-07	0.00	0.00	0.00	0.00	0.20	0.12	0.12	0.43	04-07	0.00	0.00	0.00	0.00	0.20	0.12	0.12	0.43	04-07				1.00	1.00	1.00	1.00	1.00	
04-08	0.00	0.00	0.00	0.00	0.00	0.12	0.12	0.00	04-08	0.00	0.00	0.00	0.00	0.00	0.12	0.12	0.00	04-08				1.00	1.00	1.00	1.00	1.00	
04-09	0.00	0.00	0.00	0.00	0.00	0.83	0.83	0.00	04-09	0.00	0.00	0.00	0.00	0.00	0.83	0.00	0.83	04-09				1.00	1.00	1.00	1.00	1.00	
04-14	0.00	0.00	0.00	0.00	0.67	0.00	0.00	0.67	04-14	0.00	0.00	0.00	0.00	0.67	0.00	0.00	0.67	04-14				1.00	1.00	1.00	1.00	1.00	
04-16	0.00	0.00	0.00	0.00	0.13	0.00	0.13	0.00	04-16	0.00	0.00	0.00	0.00	0.13	0.00	0.00	0.13	04-16				1.00	1.00	1.00	1.00	1.00	
04-17	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	04-17	0.00	0.00	0.00	0.00	0.00	3.87	3.87	0.00	04-17				0.00	0.00			0.00	
04-28	0.00	0.00	0.00	0.00	0.27	0.00	0.00	0.27	04-28	0.00	0.00	0.00	0.00	0.27	0.00	0.00	0.27	04-28				1.00	1.00	1.00	1.00	1.00	
05-04	0.00	0.00	0.00	0.10	0.00	0.00	0.10	0.10	05-04	0.00	0.00	0.00	0.10	0.10	0.12	0.17	0.48	05-04				0.00	1.00	0.00	0.00	0.21	
06-04	0.00	0.00	0.00	0.03	0.00	0.00	0.00	0.03	06-04	0.00	0.00	0.00	0.03	0.00	0.00	0.00	0.03	06-04				0.00				1.00	
07-04	0.00	0.00	0.00	0.17	0.10	0.17	0.17	0.43	07-04	0.00	0.00	0.00	0.00	0.17	0.13	0.17	0.47	07-04				1.00	0.75	1.00	1.00	0.93	
08-04	0.00	0.00	0.00	0.00	0.00	2.15	2.15	0.00	08-04	0.00	0.00	0.00	0.00	0.00	2.15	2.15	0.00	08-04				1.00	1.00	1.00	1.00	1.00	
09-04	0.00	0.00	0.00	0.00	0.00	0.62	0.62	0.00	09-04	0.00	0.00	0.00	0.00	0.00	0.62	0.62	0.00	09-04				1.00	1.00	1.00	1.00	1.00	
10-03	0.00	0.00	0.00	0.27	0.00	0.00	0.00	0.27	10-03	0.00	0.00	0.00	0.27	0.00	0.00	0.00	0.27	10-03				1.00	1.00	1.00	1.00	1.00	
11-03	0.00	0.43	0.00	0.00	0.00	0.00	0.00	0.43	11-03	0.00	0.43	0.00	0.27	0.00	0.00	0.00	0.70	11-03				1.00	0.62			0.62	
12-03	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	12-03	0.00	0.00	0.15	0.00	0.00	0.00	0.00	0.15	12-03				0.00				0.00	
13-04	0.00	0.00	0.00	0.15	0.00	0.00	0.00	0.15	13-04	0.00	0.00	0.00	0.15	0.00	0.00	0.00	0.15	13-04				1.00				1.00	
14-15	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	14-15	0.00	0.00	0.00	0.00	0.00	0.88	0.00	0.88	14-15				0.00	0.00			0.00	
15-04	0.00	0.00	0.00	0.00	1.55	0.00	0.00	1.55	15-04	0.00	0.00	0.00	0.00	1.55	0.00	0.00	1.55	15-04				1.00	1.00	1.00	1.00	1.00	
16-04	0.00	0.00	0.00	0.00	0.17	0.00	0.17	0.00	16-04	0.00	0.00	0.00	0.00	0.17	0.00	0.00	0.17	16-04				1.00	1.00	1.00	1.00	1.00	
17-18	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	17-18	0.00	0.00	0.00	0.00	0.00	1.48	1.48	0.00	17-18				0.00	0.00			0.00	
18-04	0.00	0.00	0.00	0.00	1.33	0.00	0.00	1.33	18-04	0.00	0.00	0.00	0.00	0.00	1.33	1.33	0.00	18-04				1.00	1.00	1.00	1.00	1.00	
19-20	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	19-20	0.93	0.00	0.00	0.00	0.00	0.00	0.00	0.93	19-20	0.00							0.00	
20-21	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	20-21	3.10	0.00	0.00	0.00	0.00	0.00	0.00	3.10	20-21	0.00							0.00	
21-22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	21-22	0.42	0.00	0.00	0.00	0.00	0.00	0.00	0.42	21-22	0.00							0.00	
22-23	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	22-23	0.00	0.10	0.00	0.00	0.00	0.00	0.00	0.10	22-23	0.00							0.00	
22-24	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	22-24	0.00	0.13	0.17	0.20	0.00	0.00	0.00	0.50	22-24	0.00	0.00	0.00	0.00	0.00			0.00	
23-22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	23-22	0.00	0.15	0.00	0.00	0.00	0.00	0.00	0.15	23-22	0.00							0.00	
24-22	0.00	0.00	0.00	0.00	0.18	0.00	0.17	0.00	24-22	0.00	0.00	0.67	0.00	0.00	0.00	0.00	0.35	24-22	0.00							0.00	
24-25	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	24-25	0.00	0.00	0.67	0.00	0.00	0.00	0.00	0.67	24-25								0.00	
25-22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	25-22	0.00	0.83	0.00	0.00	0.00	0.00	0.00	0.83	25-22								0.00	
26-27	0.00	0.00	0.00	0.00	0.12	0.12	0.12	0.00	26-27	0.00	0.00	0.00	0.00	0.12	0.12	0.12	0.00	26-27								0.00	
27-03	0.00	0.00	0.00	0.00	1.05	1.05	1.05	0.00	27-03	0.00	0.00	0.00															

- 1. Introduction
 - 2. Theory
 - 3. Method
 - 4. Implementation
 - **5. Results**
 - 6. Conclusions



i.: Long time spent moving

Home1 → Home2

sun: Long time spent moving

Home2 → Home1

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Wednesdays worst detection

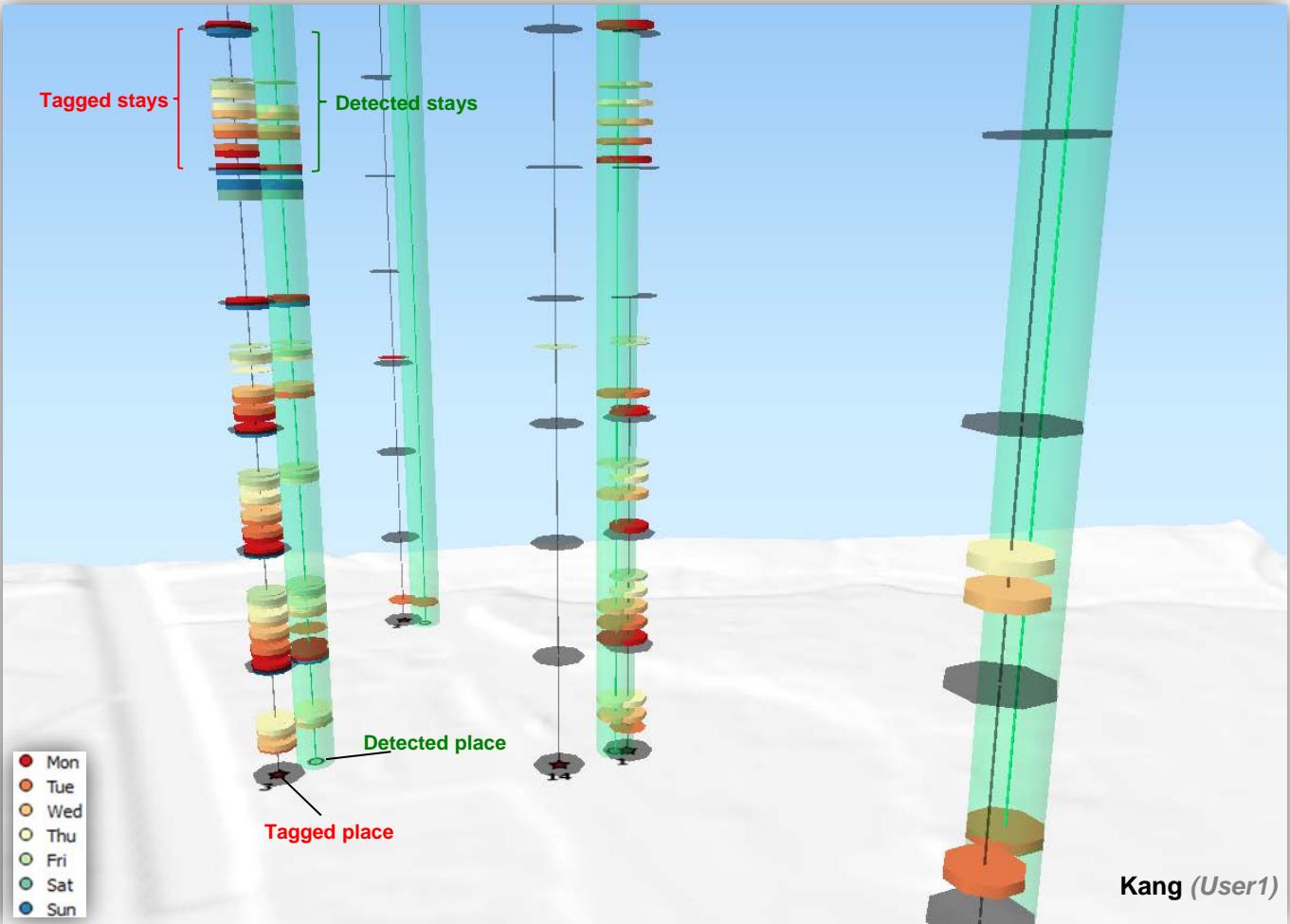
Weekends best detection

5. Results

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2. Characterization of stays and transitions

➤ Visualization of movement behaviour



6. Conclusions

- 1. Introduction
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- **Incremental approach** best performance: **detection of visited places** and **extraction of stays/transitions**:
 - User's typical **stay length** influences optimum **time** parameter
 - Algorithm **runtime** → implementation on mobile applications
- **Incremental + density-based poor** results → GTD dataset structure
- **Density-based** approach **intermediate** results:
 - Parameter settings highly dependant on dataset structure.
 - Clustering requires whole dataset and demands higher computing resources.
- User's **movement behaviour** profiling.
- **Visit duration / weekday** basic info for **stays** and **transition** characterization.
- **Transitions** require also *origin / destination. Euclidean distance / speed* → detection **anomalous situations** and **errors**

Future work

- **Presence probabilities** with **Markov Models**
- **Means of transport** → dynamic adaptation of Kang's *distance*
- Additional **pre-processing** of input → improve **combined approach** performance

THANK YOU VERY MUCH FOR YOUR ATTENTION !

Muchas gracias por su atención !

And, thank you very much:



References

- Ankerst, M. et al., 1999. Optics: Ordering points to identify the clustering structure. In *ACM Sigmod Record*. pp. 49–60. Available at: <http://dl.acm.org/citation.cfm?id=304187>.
- Ester, M. et al., 1996. A Density-Based Algorithm for Discovering Clusters in Large Spatial Databases with Noise. In *Second International Conference on Knowledge Discovery and Data Mining*. pp. 226–231. Available at:
<http://citeseerx.ist.psu.edu/viewdoc/summary?doi=10.1.1.20.2930>.
- Kang, J.H. et al., 2005. Extracting places from traces of locations. *ACM SIGMOBILE Mobile Computing and Communications Review*, 9(3), p.58. Available at:
<http://portal.acm.org/citation.cfm?doid=1024733.1024748> [Accessed August 13, 2014].
- Montoliu, R., Blom, J. & Gatica-Perez, D., 2013. Discovering places of interest in everyday life from smartphone data. *Multimedia Tools and Applications*, 62, pp.179–207.
- Shekhar, S., Zhang, P. & Huang, Y., 2003. Trends in Spatial Data Mining. *Science*, 7, pp.357–379. Available at:
<http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.77.1454&rep=rep1&type=pdf>.
- Ye, Y. et al., 2009. Mining individual life pattern based on location history. *Proceedings - IEEE International Conference on Mobile Data Management*, pp.1–10.
- Salzburg Research Forschungsgesellschaft mbH, 2015. (*Internal reports*).