



CNeRG IIT Kharagpur

UDAT: User Discrimination using Activity-Time Information

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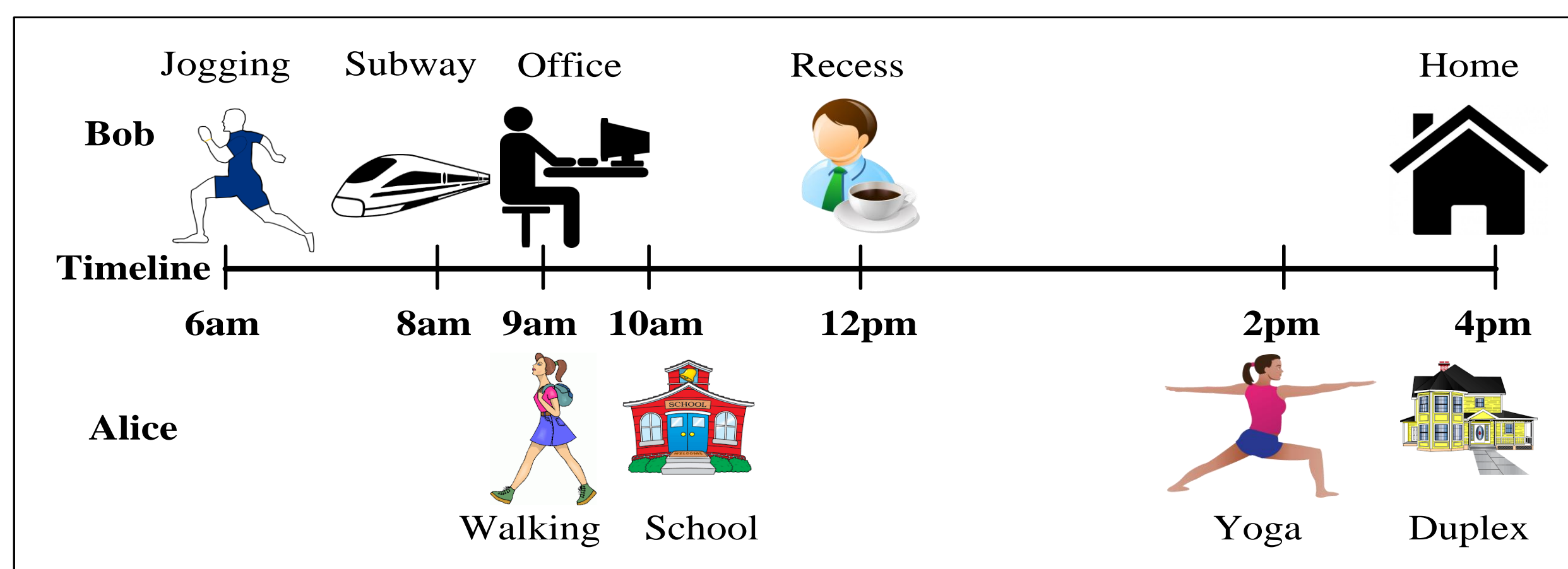
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Objective

- ✓ Developing *UDAT*, an user identifier cum discriminator model
 - Leveraging on the daily activity patterns and
 - Temporal information of the users
 - In a privacy preserving manner

Key Idea

- ✓ Activity pattern followed by an individual carries a signature of that person
- ✓ **Example:** Alice and Bob's daily routine in the weekdays



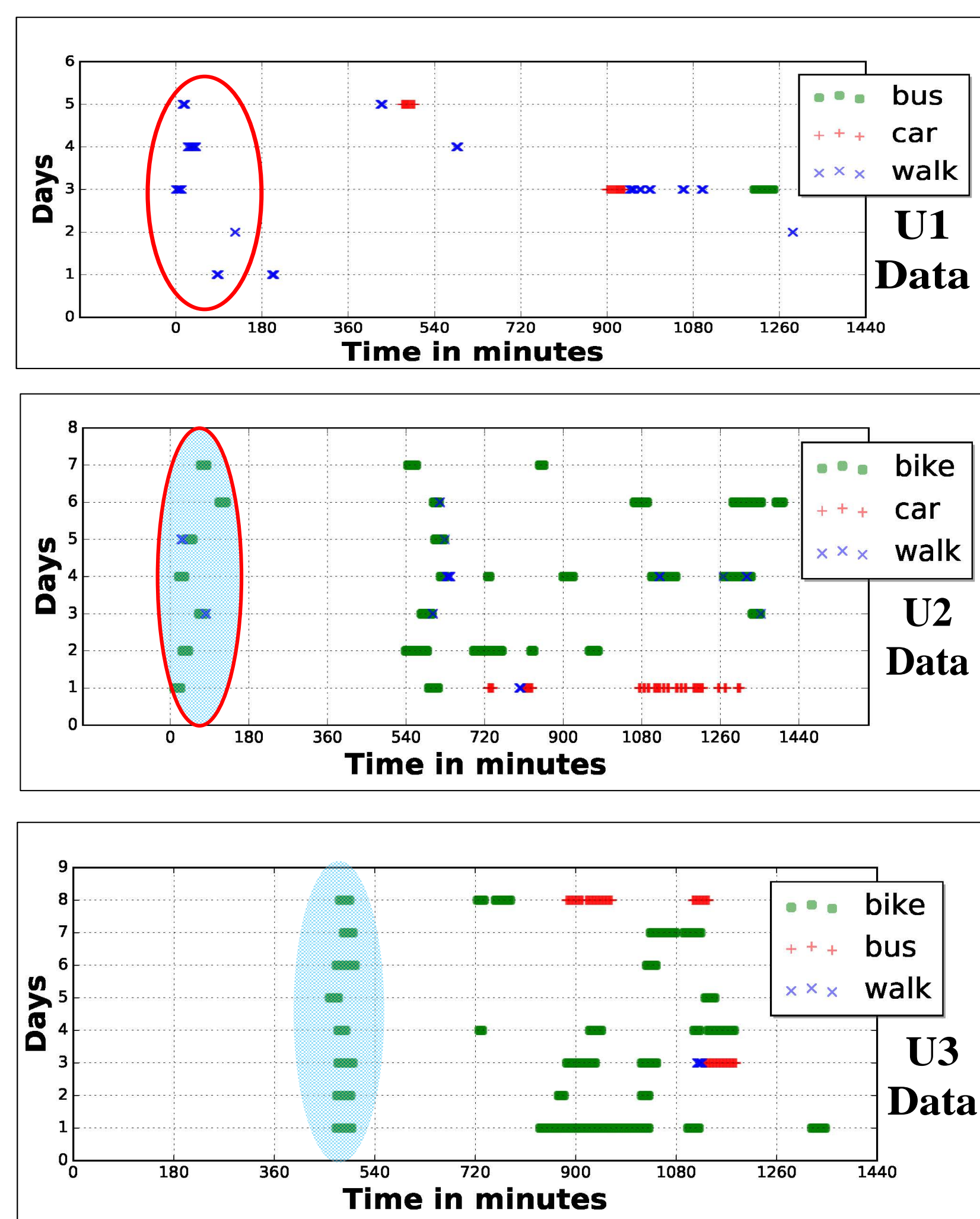
- ✓ Daily activity patterns of Alice & Bob are **substantially different** in terms of activities and temporal information

Motivation

- **Personalized recommendation** is necessary for the development of the future generation services in the domain of “**Smart City**”
- Smart devices should **automatically recognize** individual users
- **Registering** individual users each time may appear **restrictive**
- System should automatically identify the users in a **privacy preserved manner** without accessing any personally identifiable information (like IMEI) and perform service differentiation

Does daily routine discriminate the users?

- ✓ Daily routine of three users (Geolife¹ Dataset)
- ✓ **Major activity of U1:** walking
- ✓ **Major activity of U2:** biking
 - Performed activities simply **differentiate** the users U1 and U2
- ✓ **Biking time of U2:** 0 - 180 (12am - 3am) minutes of the day
- ✓ **Biking time of U3:** 360 – 540 (6am - 9am) minutes of the day
 - Routine of U2 and U3 **varies significantly over** the period of time



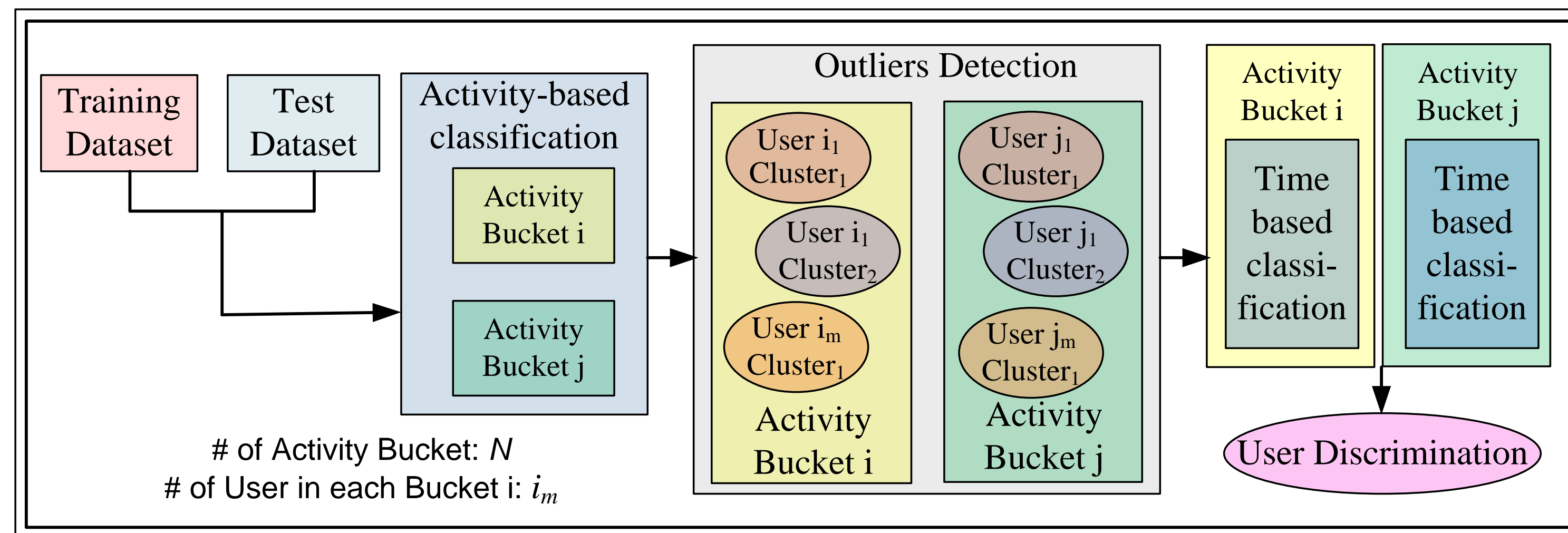
Contributions

- ✓ Developing *UDAT*, a user discrimination model, which involves three major modules – (a) Activity based classification, (b) Outliers detections, and (c) Temporal classification
- ✓ Developing smartphone based **data collection framework**, named as **UDAT Dataset**
- ✓ *UDAT* model discriminates users with **73.3%** and **80.68% accuracy**, for Microsoft Geolife and UDAT datasets respectively

Dataset

- ✓ **Geolife¹ Dataset:**
 - Users: 24 out of 178 labelled
 - Time: 4 years
 - Transportation mode: Bike, Bus, Car, Subway, and Walk
- ✓ **UDAT Dataset:** Launch experiment at IIT Kharagpur
 - Users: 15
 - Time: 3 months
 - Activity: InVehicle, OnFoot, OnBicycle, Running, Walking, Still, Tilting, and Unknown

UDAT Model



- ✓ **Activity based Classification:**
 - The module classifies users based on the activities performed by them
 - It construct one bucket for each activity
 - Each activity bucket contains information of several users
- ✓ **Outliers Detection:**
 - Activity bucket holds normal activity pattern along with few exception
 - This module eliminates those exception using **DBSCAN** model
 - It identifies the users' multiple patterns of activity
- ✓ **Temporal Classification:**
 - The module discriminates the users inside each activity bucket
 - **Models:** kNN, Random Forest, Logistic Regression, SVC
 - **Temporal features:**
 - First start time of the activity
 - Total duration for that activity
 - Maximum activity duration
 - Activity trip count

Experimental Results

GEOLIFE DATASET: ACCURACY (%) COMPARISON OF DIFFERENT MODELS (KM: Kwapisz's MODEL)					
Activity	UDAT	KM	Activity	UDAT	KM
Bike	85.47	69.47	Bus	82.92	65.03
Car	83.20	79.46	Subway	70.50	68.68
Walk	44.41	32.44	Average	73.30	63.01

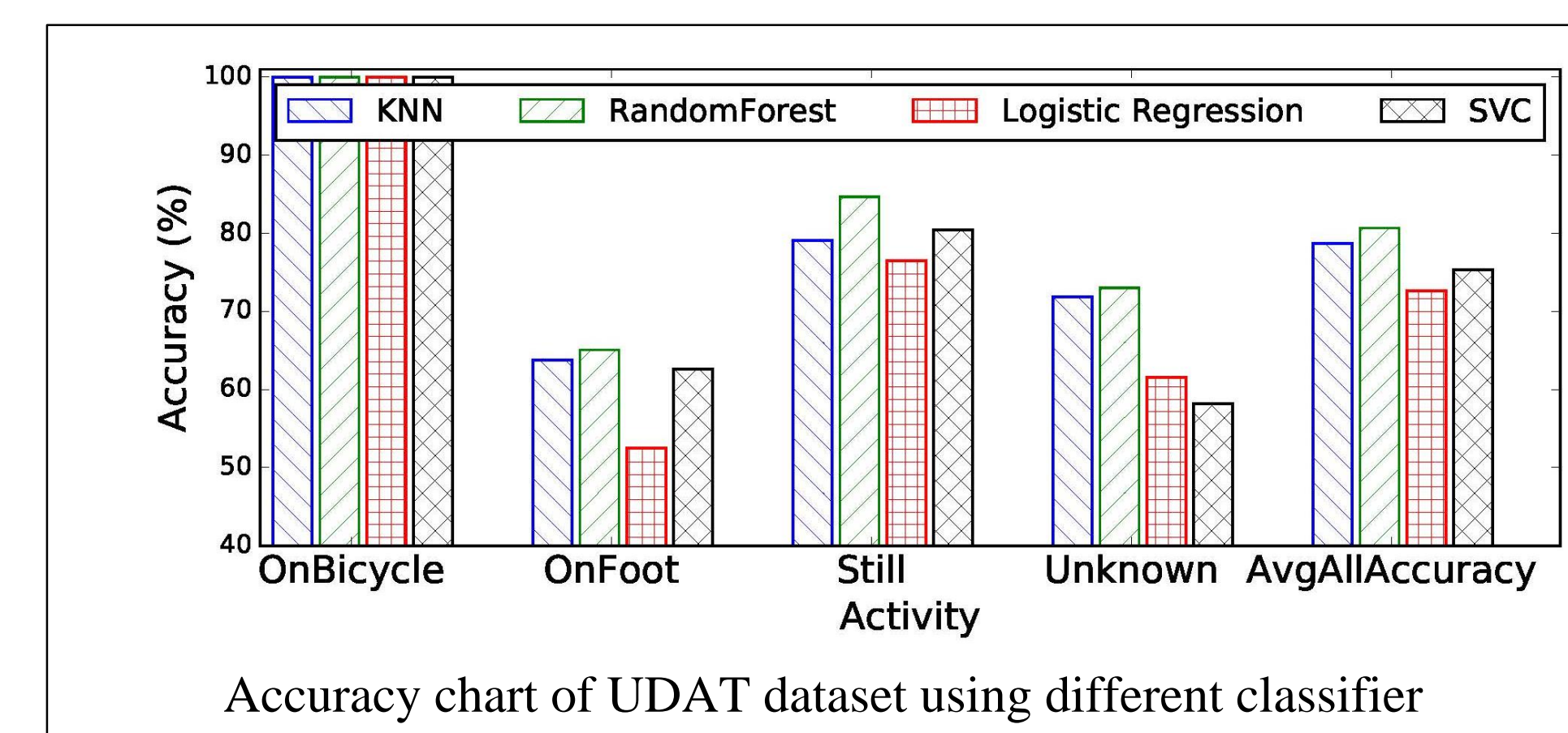
✓ Geolife¹ DataSet

- *UDAT* Model accuracy – **73.3%**
- *Kwapisz's²* Model accuracy – 63%

UDAT DATASET: ACCURACY (%) COMPARISON OF DIFFERENT MODELS			
Activity	UDAT	Kwapisz's Model	UDAT_Outlier
Tilting	NA	39.29	31.13
InVehicle	NA	32.65	33.82
OnBicycle	100.00	22.36	48.03
OnFoot	65.07	32.92	40.78
Running	NA	32.77	NA
Still	84.64	33.16	46.75
Unknown	73.01	24.56	41.61
Walking	NA	56.86	NA
Average	80.68	34.32	40.35

✓ UDAT DataSet

- *UDAT* Model accuracy – **80.68%**
- *Kwapisz's²* Model accuracy – 34.32%
- *UDAT_Outlier* Model accuracy – 40.35%



✓ Different ML Algorithms

- Average accuracy for all activities > **70%**
- **Random Forest** surpasses others

Conclusion

- ✓ Demonstrate **activity signatures** as a **valid alternative** to sensor-driven user identification paradigm
- ✓ Developed *UDAT* model **discriminates** the users using the daily activity patterns
- ✓ Experimental results state that *UDAT* model **outperformed** the sensor-driven competing algorithms for both Microsoft *Geolife¹* and UDAT dataset

References

1. Y. Zheng, L. Wang, R. Zhang, X. Xie, and W.-Y. Ma, "Geolife: Managing and understanding your past life over maps," IEEE Mobile Data Management, 2008, pp. 211–212.
2. J. R. Kwapisz, G. M. Weiss, and S. A. Moore, "Cell phone-based biometric identification," IEEE Biometrics: Theory Applications & Systems, 2010, pp. 1–7.