
DETECTING PICKPOCKET SUSPECTS FROM LARGE SCALE PUBLIC TRANSIT RECORDS

V.Sarala¹, Borra Sivaram²,

¹Assistant professor , PG DEPT, Dantuluri Narayana Raju College, **Bhimavaram, Andharapradesh**
Email:- vedalasarala21@gmail.com

²PG Student of MCA, Dantuluri Narayana Raju College, **Bhimavaram, Andharapradesh**
Email:- bsivaram6@gmail.com

ABSTRACT

Massive data collected by automated fare collection (AFC) systems provide opportunities for studying both personal traveling behaviors and collective mobility patterns in urban areas. Existing studies on AFC data have primarily focused on identifying passengers' movement patterns. However, we creatively leveraged such data for identifying pickpocket suspects. Stopping pickpockets in the public transit system has been crucial for improving passenger satisfaction and public safety. Nonetheless, in practice, it is challenging to discern thieves from regular passengers. In this paper, we developed a suspect detection and surveillance system, which can identify pickpocket suspects based on their daily transit records. Specifically, we first extracted a number of useful features from each passenger's daily activities in the transit system. Then, we took a two-step approach that exploits the strengths of unsupervised outlier detection and supervised classification models to identify thieves, who typically exhibit abnormal traveling behaviors. Experimental results demonstrated the effectiveness of our method. We also developed a prototype system for potential uses by security personnel.

1 INTRODUCTION

Public transit passengers can easily become distracted in crowded environments, where they are often rushing from one location to another. Having their focus drift from their belongings, they often become common targets of pickpockets [1, 2]. During the first 9 months of 2014, it was reported that 350 pickpockets were apprehended in the subway system and 490 on buses in Beijing.¹ Many other big cities around the world, such as Barcelona, Rome, and Paris, also suffer from pickpocket problems.² Indeed, it is challenging to detect theft activities committed by cunning thieves who know how to escape without being discovered. It is critical to provide a smart surveillance and tracking tool for transit system security personnel.

Literature Survey

Detecting pickpocketing incidents from large-scale public transit records involves utilizing various data-driven techniques and methodologies. Here's a structured literature survey and approach outline:

1. Data Collection and Preprocessing

- **Sources:** Transit records include data from fare collection systems, surveillance cameras, passenger reports, etc.
- **Preprocessing:** Cleaning data, timestamp normalization, spatial-temporal data alignment.

2. Feature Extraction

- **Spatial Features:** Geographical location, station density, routes.
- **Temporal Features:** Time of day, day of the week, seasonality.
- **Behavioral Features:** Anomalous behavior patterns (e.g., frequent changes in direction or speed).
- **Contextual Features:** Weather conditions, special events, holidays.

3 IMPLEMENTATION STUDY

EXISTING SYSTEM:

The System of existing literature focuses on finding patterns in passenger activity records. Such knowledge can be useful in a variety of applications, and plays a vital role in effectively finding and satisfying passenger needs. Examples include assessing the performance of the transit network, identifying and optimizing problematic or flawed bus routes, improving the accuracy of passenger flow forecasted between two regions, and making service adjustments that accommodate variations in ridership on different days. In particular, [4] estimated the crowdedness of various stations in the transportation network using AFC data. [9] measured the variability of transit behaviors on different days of the week.

Disadvantages:

There is no Smart Card Based Travelling due to Only Manual Passengers Activity Patterns and Manual Transit Records.

Proposed System & algorithm

In the proposed system, the system adopted a comprehensive approach to the pickpocket detection problem. The overall framework of our solution is illustrated in this system. The system first partitioned the city area into regions with functional categories. Then, the

mobility characteristics of passengers were extracted from transit records dynamically over time.

4.1 Advantages:

Effective techniques for Abnormal Traveling Behavior Detection.

Architecture

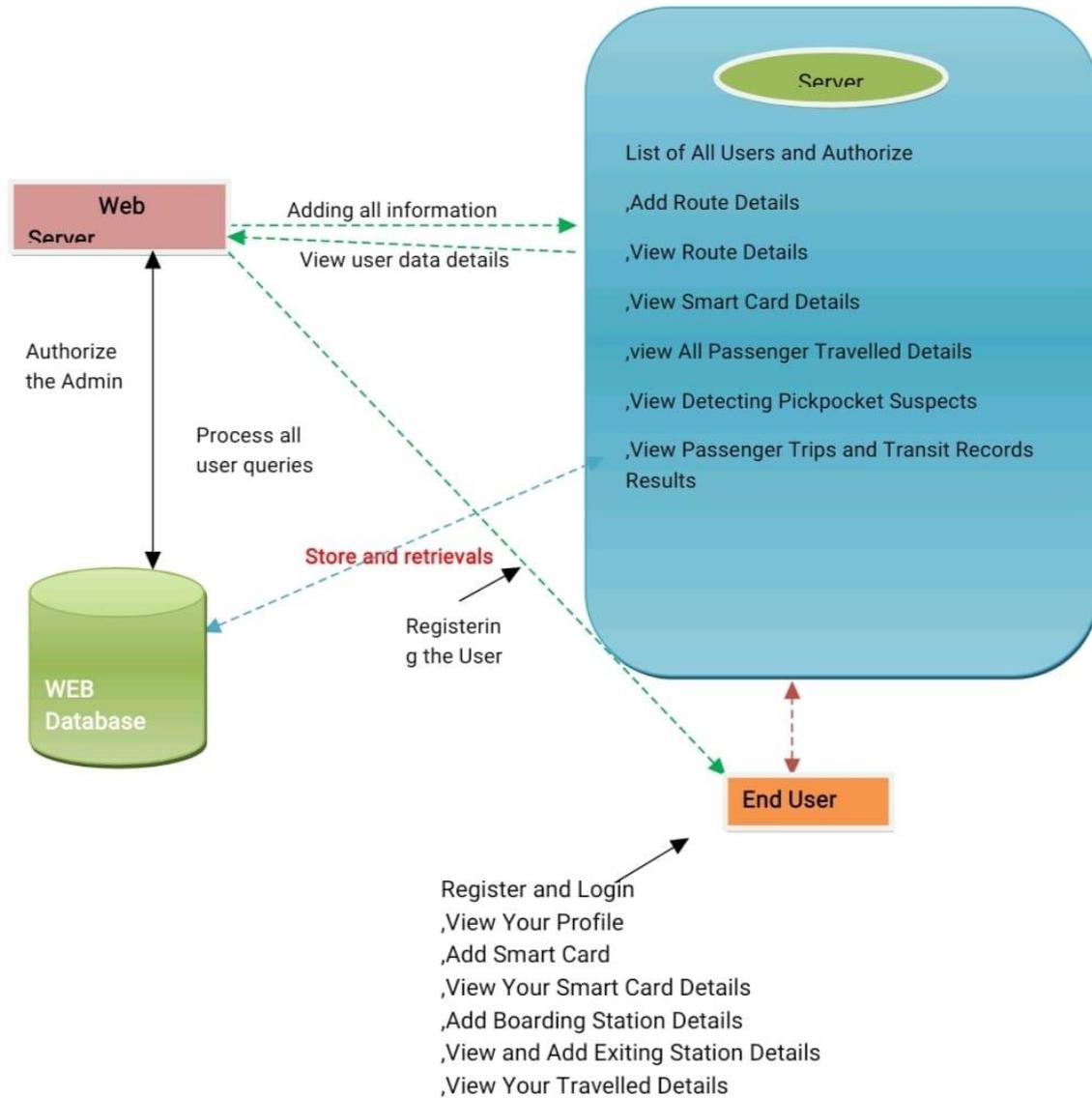


Fig:3.1 System Architecture

IMPLEMENTATION

- **Server**

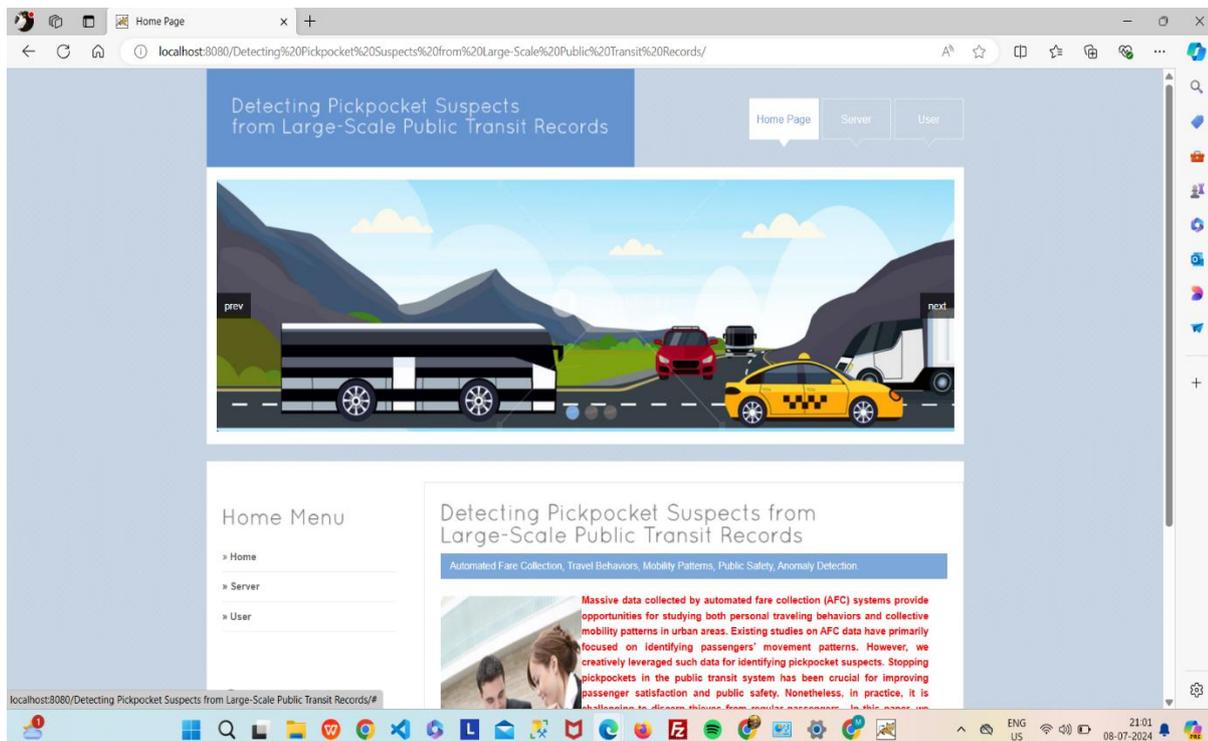
In this module, the Web Server has to login by using valid user name and password. After login successful he can do some operations such as List of All Users and Authorize, Add Route Details, View Route Details, View Smart Card Details ,view All Passenger Travelled Details ,View Detecting Pickpocket Suspects ,View Passenger Trips and Transit Records Results

- **User**

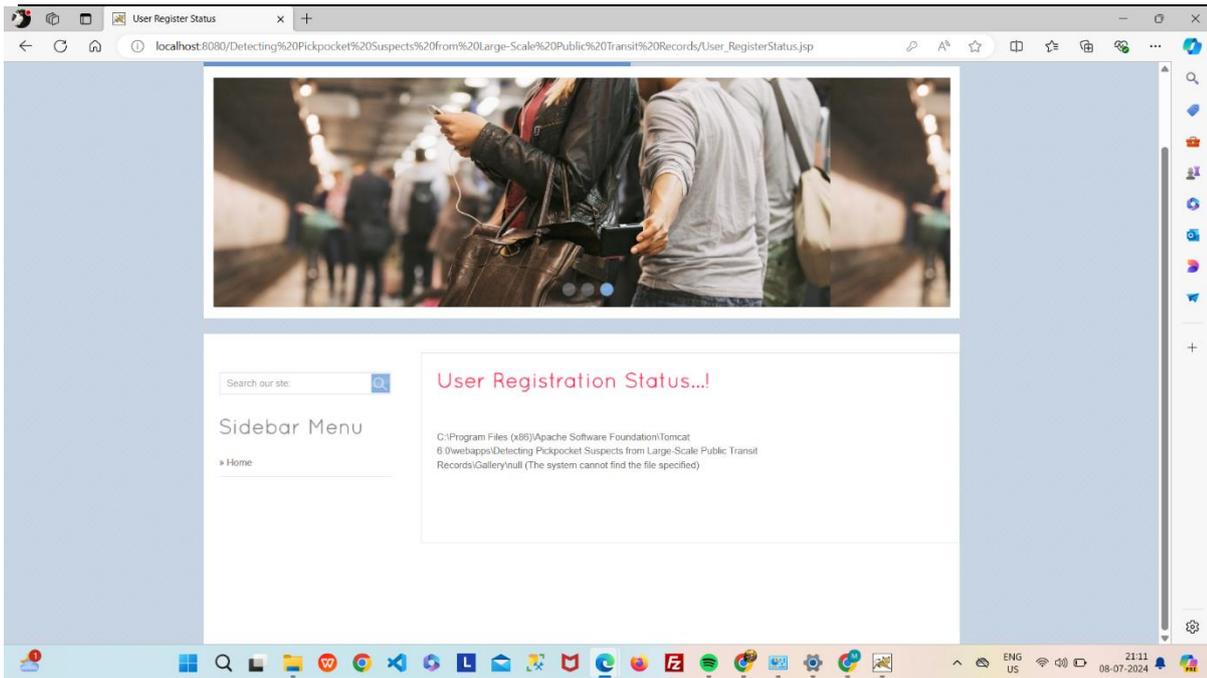
In this module, there are n numbers of users are present. User should register before doing some operations. After registration successful he has to login by using authorized user name and password. Login successful he will do some operations like View Your Profile, Add Smart Card, View Your Smart Card Details, Add Boarding Station Details, View and Add Exiting Station Details, View Your Travelled Details.

5 RESULTS AND DISCUSSION

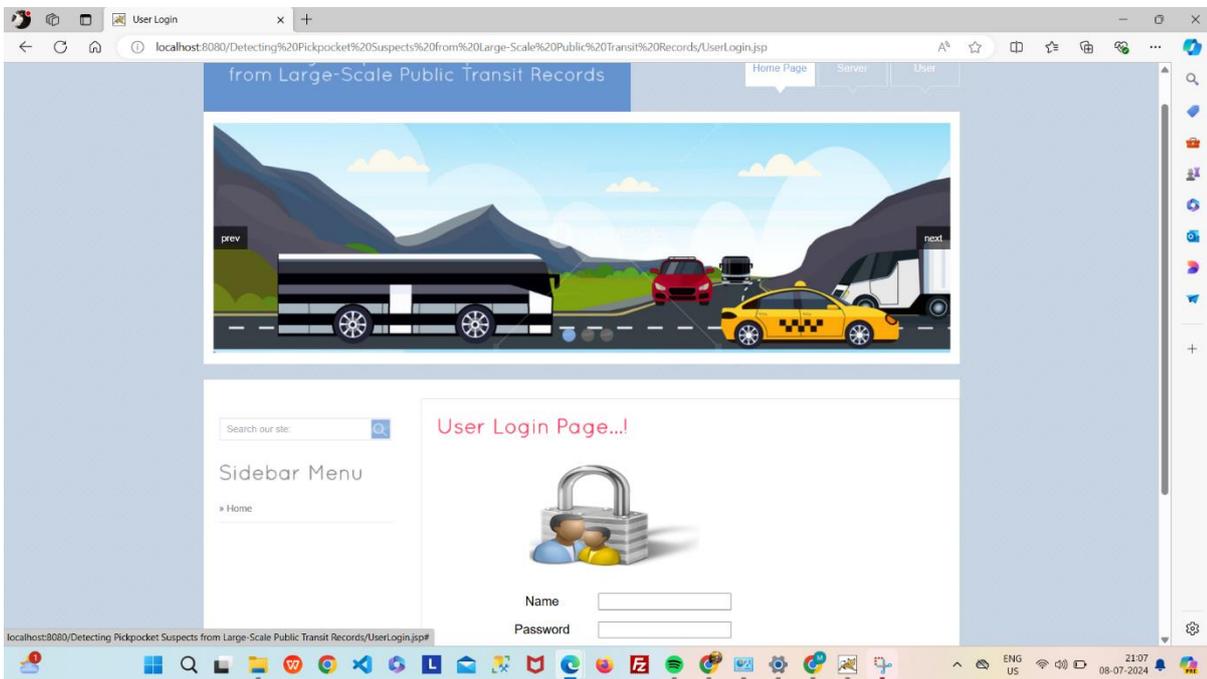
HOME PAGE:

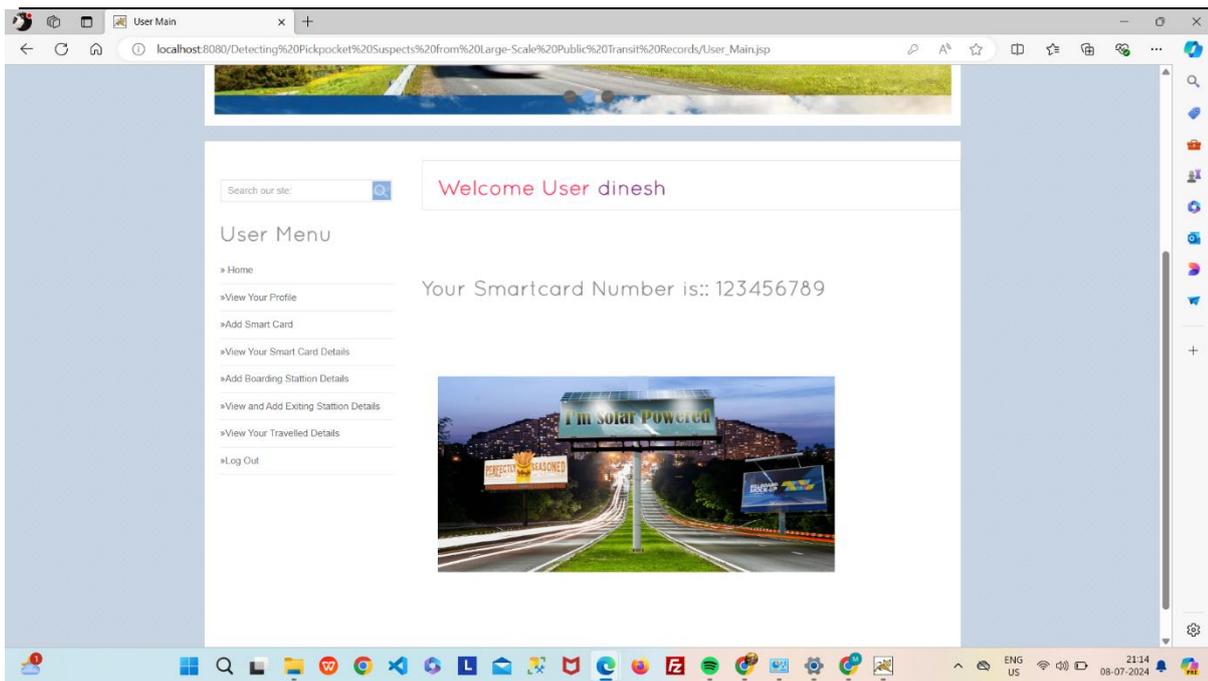


User Registration status:

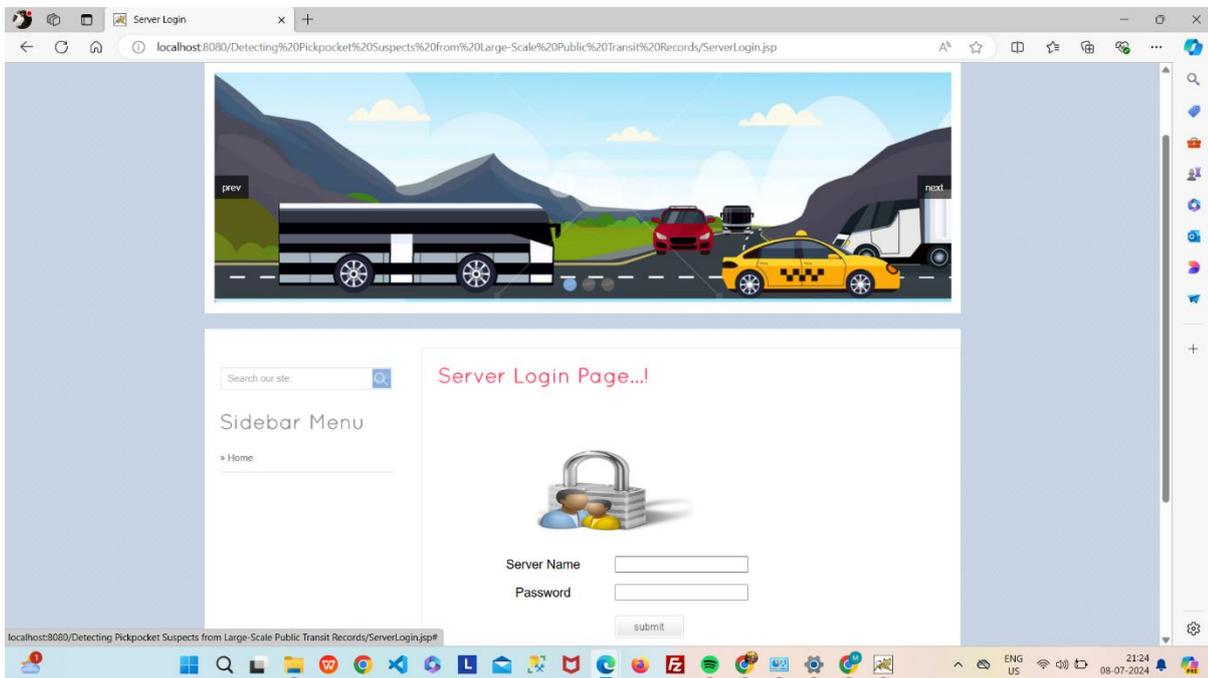


User Login Page:





Server Login Page :



Server Menu

- Home
- List of All Users and Authorize
- Add Route Details
- View Route Details
- View Smart Card Details
- View All Passenger Travelled Details
- View Detecting Pickpocket Suspects
- View Passenger Trips and Transit Records Results
- Logout

(a) Example Activities

(b) Trips

Station	Time
a	07:15
b	07:40
c	07:46
d	08:23
e	17:35
f	18:08
a	19:32
b	19:54
c	19:58
d	20:15

(c) Transit Records

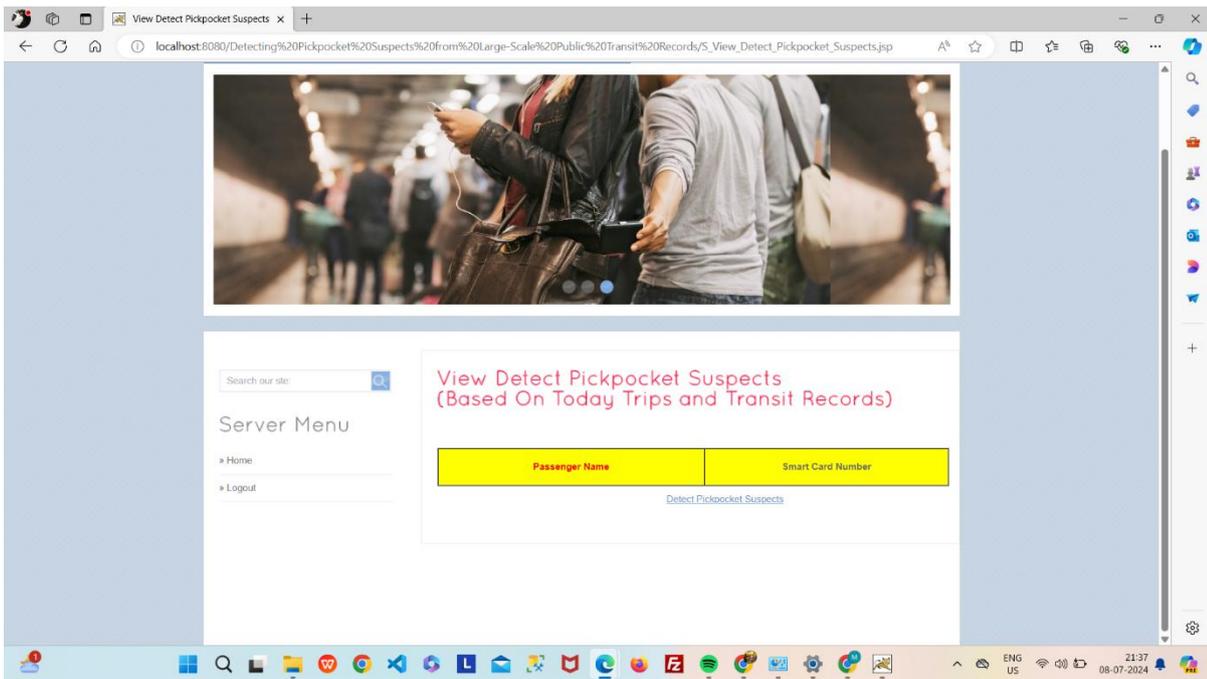
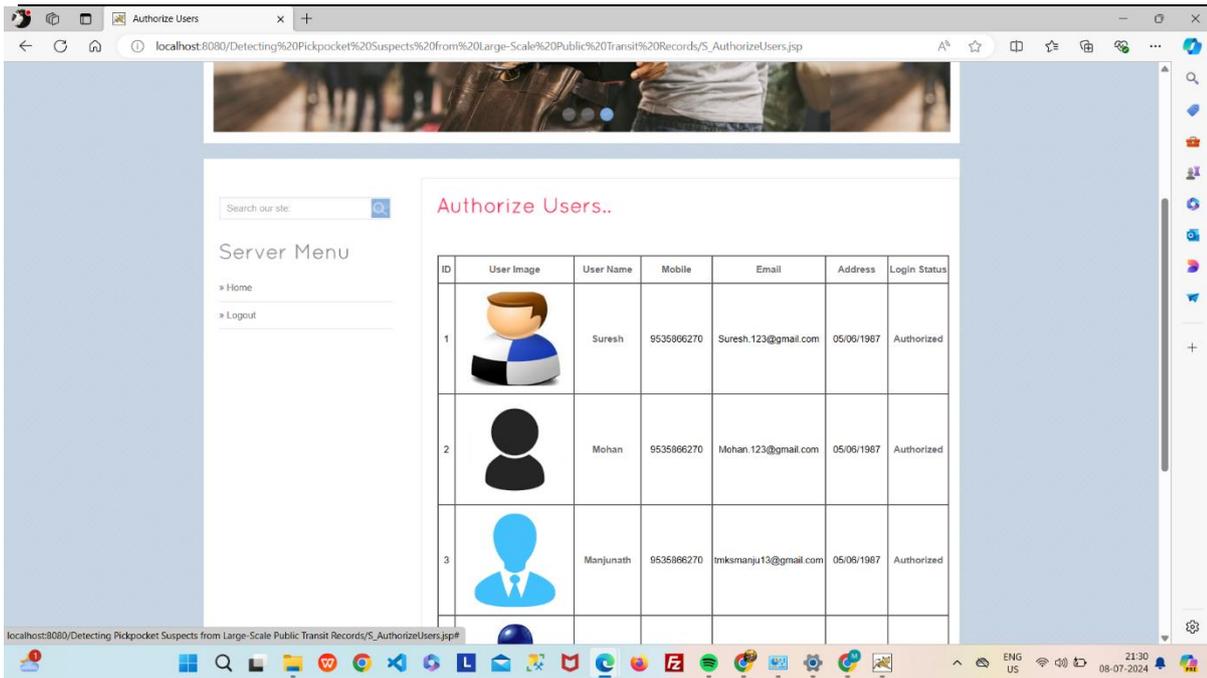
Smart Card ID	Route Number	Boarding Station	Boarding Time	Exiting Station	Exiting Time
4322	Route 52	a	07:15	b	07:40
4322	Route 26	c	07:46	d	08:23
4322	Route 11	d	17:35	e	18:08
4322	Route 11	e	19:32	f	19:54
4322	Route 16	f	19:58	a	20:15

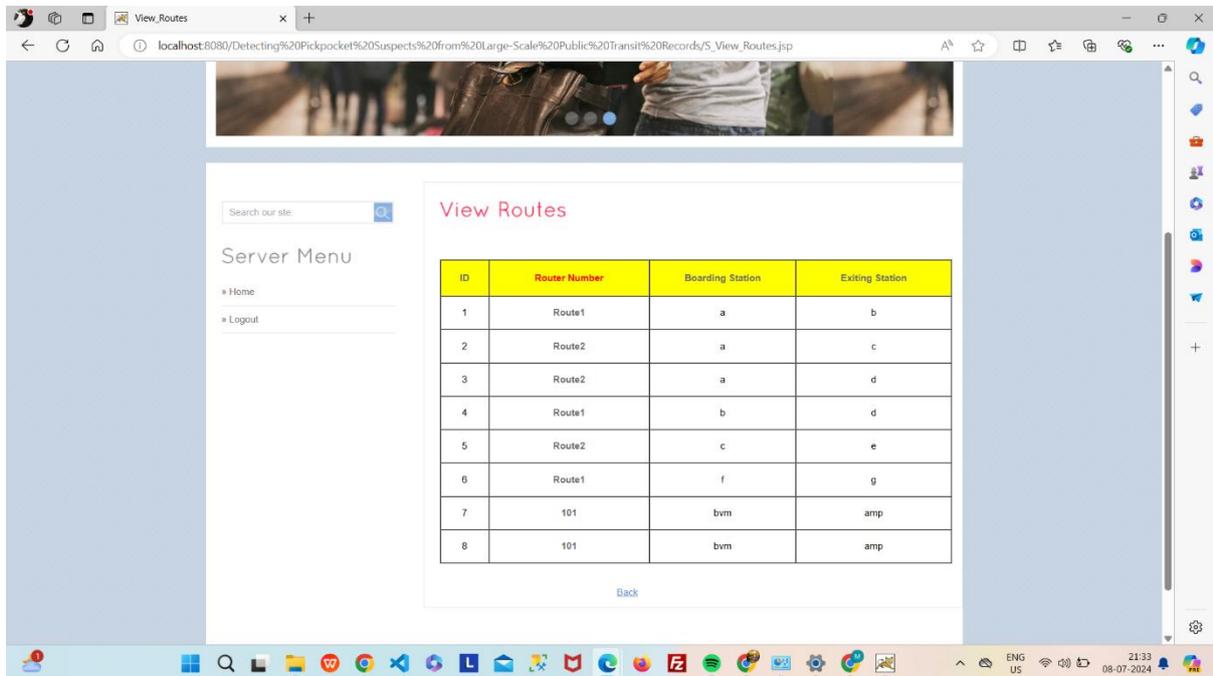
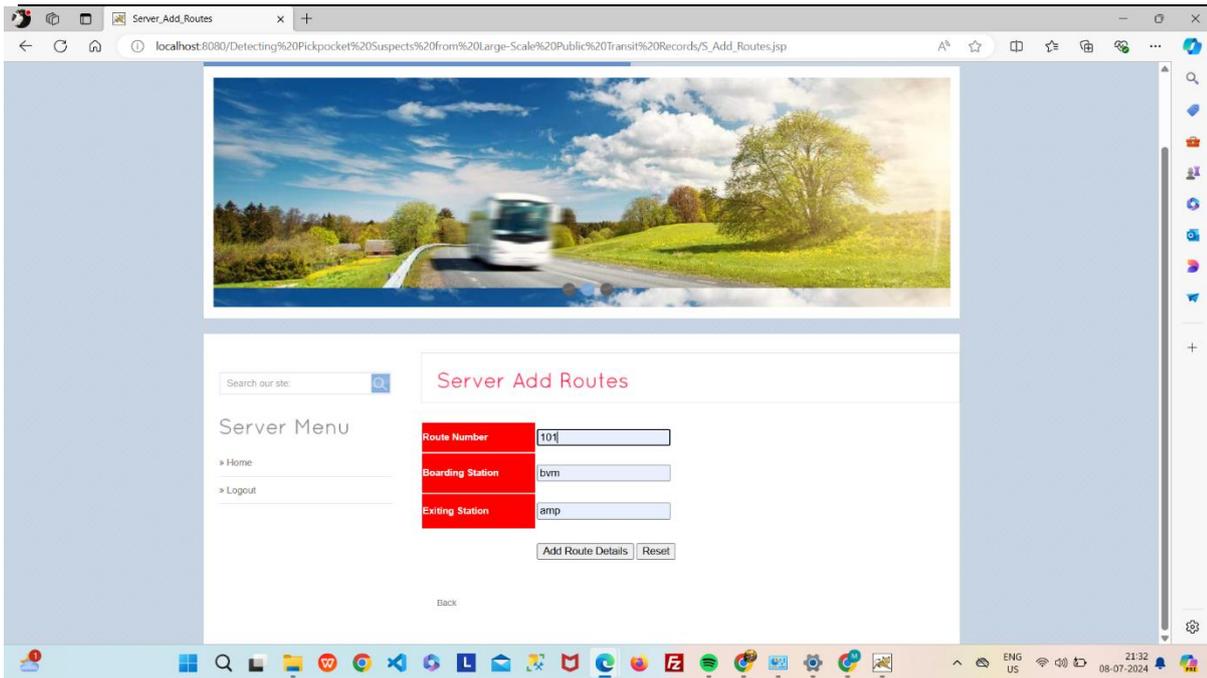
Server Menu

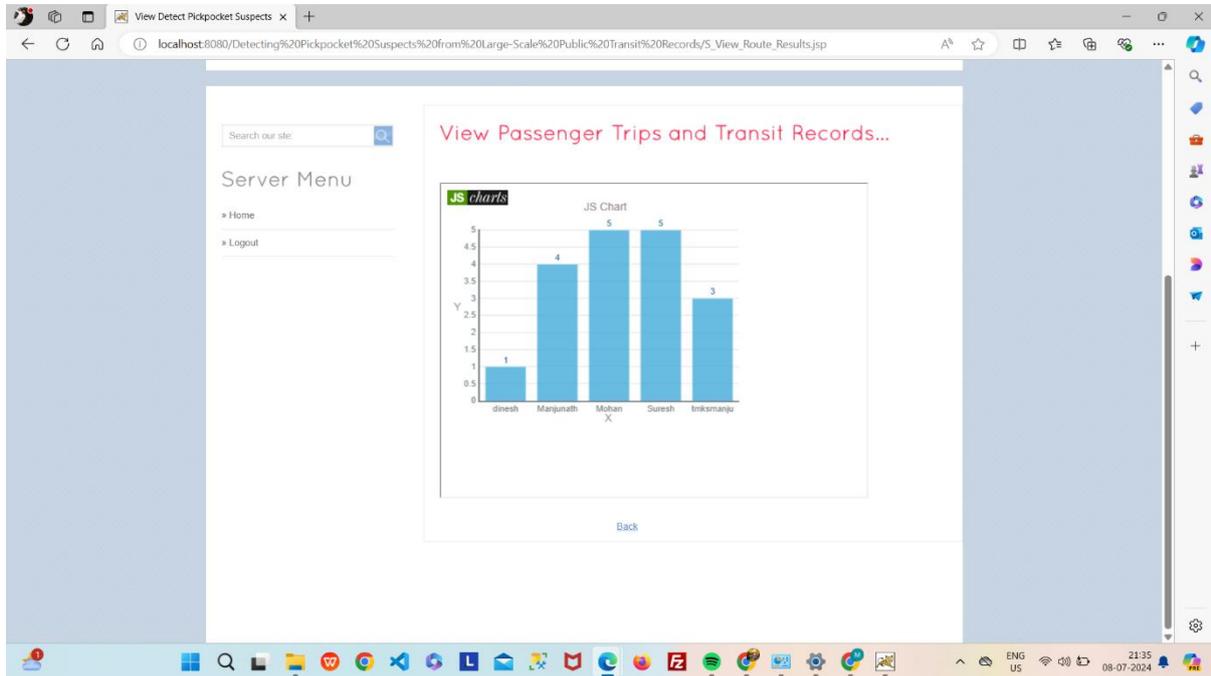
- Home
- Logout

View All Passenger Travelled Details

ID	Passenger Name	Card Number	Route Number	Boarding Station	Exiting Station	Boarding DT	Exiting DT
18	dinesh	123456789	101	bvm	amp	15/06/2024 15:54:05	15/06/2024 15:54:21
9	Manjunath	782972189	Route1	a	b	21/08/2019 17:31:45	21/08/2019 17:32:35
11	Manjunath	782972189	Route1	c	d	21/08/2019 17:33:20	21/08/2019 17:33:58
14	Manjunath	782972189	Route1	b	e	21/08/2019 18:48:20	21/08/2019 18:49:01
15	Manjunath	782972189	Route1	c	g	21/08/2019 18:49:31	21/08/2019 18:50:03
4	Mohan	7878753	Route1	a	c	21/08/2019 16:03:58	21/08/2019 16:06:59
7	Mohan	7878753	Route2	b	d	21/08/2019 16:36:51	21/08/2019 16:39:27
8	Mohan	7878753	Route1	b	c	21/08/2019 16:49:36	21/08/2019 16:50:22
10	Mohan	7878753	Route1	b	d	21/08/2019 17:32:13	21/08/2019 17:32:58
12	Mohan	7878753	Route1	b	No	21/08/2019 17:33:45	No
1	Suresh	34523456	Route1	a	c	21/08/2019 12:54:04	21/08/2019 13:42:42
2	Suresh	34523456	Route2	b	d	21/08/2019 14:28:59	21/08/2019 14:34:29
3	Suresh	34523456	Route2	c	d	21/08/2019 14:35:03	21/08/2019 15:02:54
5	Suresh	34523456	Route2	b	d	21/08/2019	21/08/2019







6. CONCLUSION AND FUTURE WORK

CONCLUSION

In this paper, we developed a suspect detection and tracking system by mining large-scale transit records. The system assists in identifying pickpocket suspects' and enables active surveillance in high-risk areas. Specifically, we first constructed a feature representation for profiling passengers. Then, we established a novel two-step framework to distinguish regular passengers from pickpocket suspects. Finally, we leveraged real-world datasets from multiple sources for model training and validation, and implemented a prototype system for end users. Experimental results on real-world data showed the effectiveness of our proposed approach.

7. REFERENCES

- [1] G. R. Newman and M. M. McNally, "Identity theft literature review," United States Department of Justice, Report 210459, July 2005.
- [2] M. Felson and R. V. Clarke, "Opportunity makes the thief: Practical theory for crime prevention," Policing and Reducing Crime Unit: Police Research Series, Report 98, 1998.
- [3] T. L. C. da Silva, J. A. F. de Macêdo, and M. A. Casanova, "Discovering frequent mobility patterns on moving object data," in *MobiGIS*, 2014, pp. 60–67.
- [4] I. Ceapa, C. Smith, and L. Capra, "Avoiding the crowds: understanding tube station congestion patterns from trip data," in *UrbComp*, 2012, pp. 134–141.
- [5] X. Ma, Y.-J. Wu, Y. Wang, F. Chen, and J. Liu, "Mining smart card data for transit riders travel patterns," *Transportation Research Part C*, vol. 36, pp. 1–12, 2013.
- [6] K. Zheng, Y. Zheng, N. J. Yuan, S. Shang, and X. Zhou, "Online discovery of gathering patterns over trajectories," *IEEE Transactions on Knowledge and Data Engineering*, vol. 26, no. 8, pp. 1974–1988, 2014.
- [7] Y. Liu, C. Liu, J. Yuan, L. Duan, Y. Fu, H. Xiong, S. Xu, and J. Wu, "Intelligent bus routing with heterogeneous human mobility patterns," *Knowledge and Information Systems*, 2016, forthcoming, DOI: 10.1007/s10115-016-0948-6.
- [8] B. Du, C. Liu, W. Zhou, Z. Hou, and H. Xiong, "Catch me if you can: Detecting pickpocket suspects from large-scale transit records," in *KDD*, 2016, pp. 87–96.
- [9] C. Morency, M. Trépanier, and B. Agard, "Analysing the variability of transit users behaviour with smart card data," in *ITSC*, 2006, pp. 44–49.

