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# UrbanFlow360: A Cloud-Native Architecture for Real-Time Urban Traffic Congestion Prediction

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**Abstract:** UrbanFlow360 is a cloud-native Software-as-a-Service (SaaS) platform for real-time traffic congestion prediction and urban traffic analytics. Developed using a modular microservice architecture and containerized with Docker, the system integrates predictive machine learning pipelines and offers live analytics through an interactive Streamlit dashboard. This paper details the end-to-end development lifecycle of UrbanFlow360, including dataset preprocessing, model training, application development, containerization, and deployment on AWS Elastic Container Registry (ECR) and Amazon EC2. The architecture demonstrates best practices in reproducibility, portability, and scalability of intelligent transportation systems.

**Index Terms:** Traffic prediction, cloud-native, AWS, machine learning, Streamlit, ECR, EC2, Docker, containerization, urban analytics, smart cities

## I. INTRODUCTION

Urban areas face mobility challenges due to rising vehicle volumes, infrastructure limitations, and unpredictable congestion. Traditional systems fall short in providing predictive insights.

UrbanFlow360 addresses this gap using cloud-native technologies. It enables real-time traffic congestion prediction via a browser-based dashboard powered by Streamlit and deployed through AWS infrastructure.

## II. TEAM AND CONTRIBUTIONS

- 1) Akash Tiwari: Cloud and DevOps lead — EC2, Docker, ECR, IAM roles.
- 2) Aashish Dewangan: Data preprocessing and ingestion — handled wide CSVs, JSONs, and schema design.
- 3) Kishan Kumar Bouri: ML pipeline development and training.
- 4) Yash Mathur: Streamlit frontend integration and UI/UX refinement.

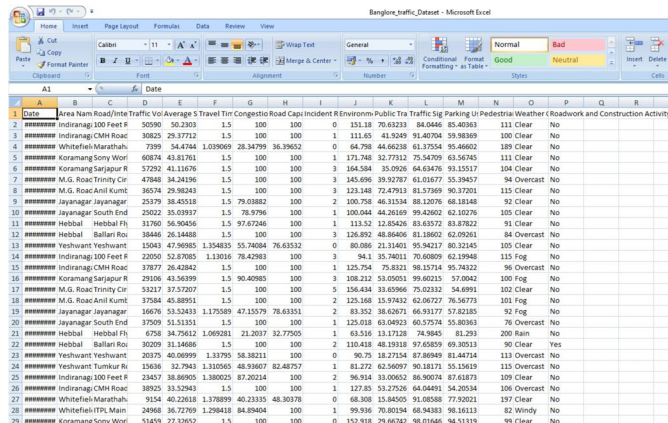
## III. DATASET DESCRIPTION

Data was collected from traffic logs across Delhi and Bangalore. A sample of the uploaded dataset is shown in Figure 1.

## IV. PREPROCESSING PIPELINE

We designed a consistent preprocessing flow to normalize column names, fill missing values, and encode categorical features.

- 1) Missing value imputation



A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S
1	Date	Area Nam Road/Indragiri	Average 5 Travel Time	Congestion Road	Cap Incident	Environment	Public Traffic	Sig Parking	Up Pedestrian	Weather	Roadwork	Construction	Activity					
2	Indragiri	100 Feet	50090	50.2303	1.5	100	100	0	151.18	70.6323	84.0446	85.40363	111	Clear	No			
3	Indragiri	CMH Road	30825	29.37712	1.5	100	100	1	111.85	41.5249	91.40704	99.98369	100	Clear	No			
4	Whitefield	Marathahalli	7399	54.4744	1.070609	28.3479	38.19652	0	64.788	44.6528	61.37504	95.46602	189	Clear	No			
5	Koramangla	Sony Worli	60874	43.81761	1.5	100	100	1	171.748	32.77312	75.54709	63.56745	111	Clear	No			
6	Koramangla	Sarajpur R	57292	41.14676	1.5	100	100	3	164.504	35.0926	64.63476	91.15517	104	Clear	No			
7	M.G. Road	Trinity Cir	47448	34.24196	1.5	100	100	3	145.690	39.92787	61.65877	55.39437	94	Overcast	No			
8	M.G. Road	Anil Kumar	36574	29.98243	1.5	100	100	3	123.148	72.47913	81.57869	90.17201	115	Clear	No			
9	Jayanagar	Jayanagar	25379	38.45318	1.5	79.03882	100	2	100.758	46.31534	88.12076	68.18148	92	Clear	No			
10	Jayanagar	South End	29032	31.03817	1.5	78.9796	100	1	100.044	44.26149	99.42602	61.10276	105	Clear	No			
11	Hebbal	Ballari Rd	31760	56.90456	1.5	100	100	1	113.52	12.85426	83.63572	81.87822	91	Clear	No			
12	Hebbal	Ballari Rd	38446	26.14448	1.5	100	100	3	126.892	48.86406	81.18052	62.09261	84	Overcast	No			
13	Yeswanth	Yeswanth	15943	47.90505	1.358215	55.7038	76.63312	0	80.086	21.1401	95.94217	60.32445	105	Clear	No			
14	Indragiri	100 Feet	22050	52.87085	1.13016	78.42983	100	3	94.1	35.74011	70.60809	62.19948	115	Fog	No			
15	Indragiri	CMH Road	37877	26.42842	1.5	100	100	1	125.754	79.8321	98.15794	95.74322	96	Overcast	No			
16	Koramangla	Sarajpur R	29036	43.30399	1.5	90.40953	100	3	108.212	53.05051	99.90215	97.00462	100	Fog	No			
17	M.G. Road	Trinity Cir	53217	37.57207	1.5	100	100	5	156.434	33.69666	75.02332	54.6991	102	Clear	No			
18	M.G. Road	Anil Kumar	37884	45.88051	1.5	100	100	2	125.168	15.57432	62.06727	70.56773	101	Fog	No			
19	Jayanagar	Jayanagar	14676	55.53433	1.17589	47.15579	76.63312	2	81.532	38.62671	68.81777	57.82185	92	Fog	No			
20	Jayanagar	South End	37059	51.53151	1.5	100	100	1	125.018	63.94923	60.37574	55.80363	76	Overcast	No			
21	Hebbal	Ballari Rd	6758	34.75612	1.060281	21.037	32.77505	1	63.516	13.17128	74.9845	81.293	200	Rain	No			
22	Hebbal	Ballari Rd	30209	51.14468	1.5	100	100	2	110.418	48.19118	97.63059	69.30513	90	Clear	Yes			
23	Yeswanth	Yeswanth	20175	40.98999	1.313705	58.38211	100	0	90.75	18.27154	87.86491	81.44714	113	Overcast	No			
24	Yeswanth	Tumkur Rd	15436	32.7943	1.310565	48.93607	82.48757	1	81.272	62.56097	90.18171	55.15619	115	Overcast	No			
25	Indragiri	100 Feet	23457	38.86905	1.380025	87.20214	100	2	96.914	13.05052	86.90074	87.61873	109	Clear	No			
26	Indragiri	CMH Road	38925	33.52943	1.5	100	100	1	127.63	53.27526	64.04491	54.20514	106	Overcast	No			
27	Whitefield	Marathahalli	9154	40.22618	1.378899	40.23335	48.30378	0	68.308	15.84005	91.08548	77.92021	197	Clear	No			
28	Whitefield	ITP Main	24968	36.72769	1.298418	84.89404	100	1	99.916	70.80194	68.94383	86.16113	82	Windy	No			
29	Koramangla	Sony Worli	51459	27.32052	1.5	100	100	0	152.918	29.66742	88.02646	94.53219	99	Clear	No			

Fig. 1. Uploaded CSV preview showing traffic metrics

- 2) Time and weekday feature extraction
- 3) Encoding and scaling

Figure 2 visualizes this pipeline.

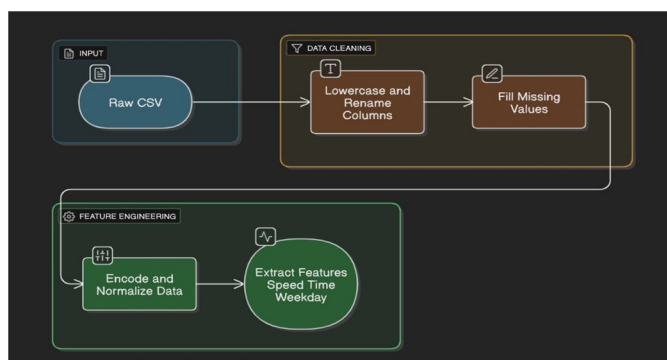


Fig. 2. Preprocessing stages for input data

## V. ARCHITECTURE OVERVIEW

UrbanFlow360's modular architecture includes preprocess- ing, model inference, UI, and cloud deployment.

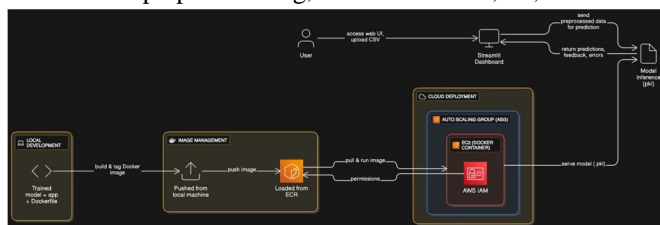


Fig. 3. UrbanFlow360 system architecture

## VI. MODEL DEVELOPMENT

### A. Algorithm

We used Random Forest Regressor for its interpretability and robustness.

### B. Training

```

model = RandomForestRegressor()
model.fit(X_train, y_train)
joblib.dump(model, "model.pkl")
  
```

### C. Evaluation

We measured accuracy using:

- Mean Absolute Error (MAE)
- Root Mean Square Error (RMSE)
- R2 score

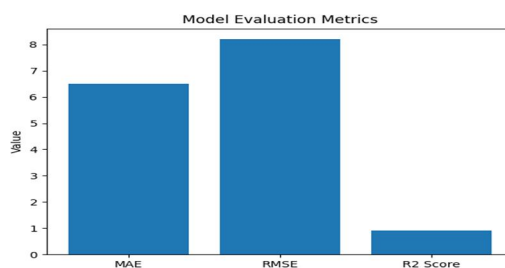


Fig. 4. Model performance metrics

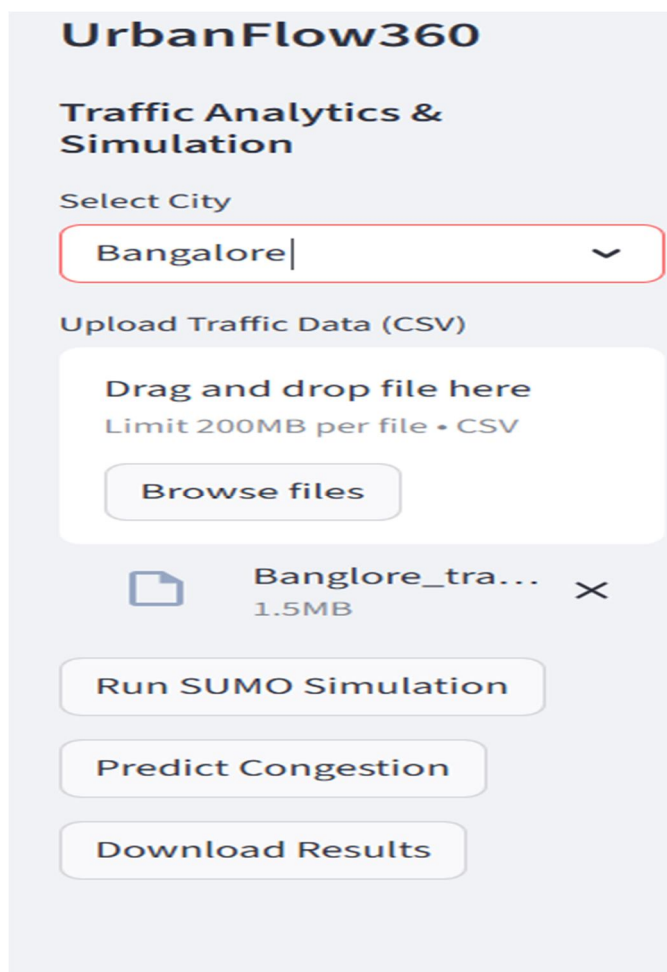
## VII. FRONTEND AND STREAMLIT INTEGRATION

Users interact via a Streamlit-based dashboard.

```
if predict_button:
    result_df = predict_csv_congestion(df, city)
    st.dataframe(result_df)
    st.bar_chart(result_df['Congestion']).
        value_counts())
```

Figure 5 shows file upload and city selection.

Figure 6 shows the prediction output with congestion chart.



**UrbanFlow360**

**Traffic Analytics & Simulation**

Select City

Bangalore

Upload Traffic Data (CSV)

Drag and drop file here  
Limit 200MB per file • CSV

Browse files

Bangalore\_tra...  
1.5MB

Run SUMO Simulation

Predict Congestion

Download Results

Fig. 5. Streamlit sidebar for file upload and city selection

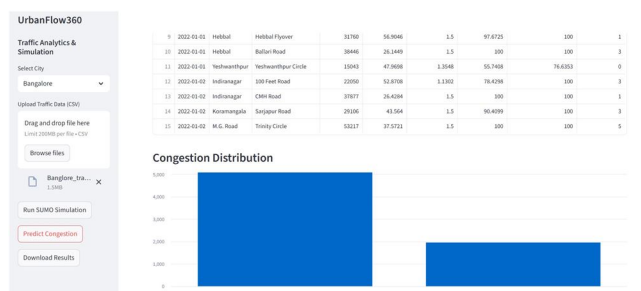


Fig. 6. Predicted congestion and bar chart output



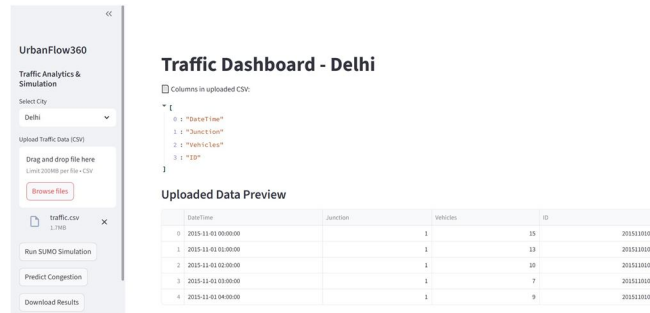


Fig. 7. Complete UI view after running predictions

## VIII. DOCKER CONTAINERIZATION

UrbanFlow360 is packaged in Docker for reproducibility.

### A. Dockerfile

```
FROM python:3.10-slim
WORKDIR /app
COPY . .
RUN pip install -r requirements.txt
CMD ["streamlit", "run", "frontend/app.py"]
```

### B. Push to ECR

```
docker build -t urbanflow360 .
docker tag urbanflow360:latest \
    631158448576.dkr.ecr.ap-south-1.amazonaws.com/urbanflow360-backend:latest
docker push 631158448576.dkr.ecr.ap-south-1.amazonaws.com/urbanflow360-backend:latest
```

## IX. AWS DEPLOYMENT (EC2 + ECR)

We deployed the model as a Docker container on EC2 after pushing it to ECR.

- 1) Create IAM role and ECR repo
- 2) Push Docker image to ECR
- 3) Launch EC2, SSH, pull image
- 4) Run app: docker run -p 8501:8501 urbanflow360

## X. RESULTS

The app performed well on city-level congestion predictions.

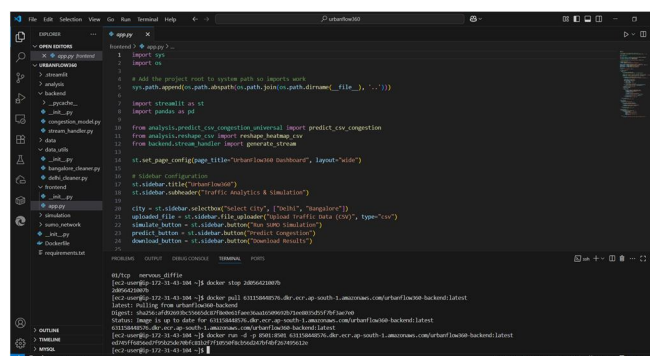
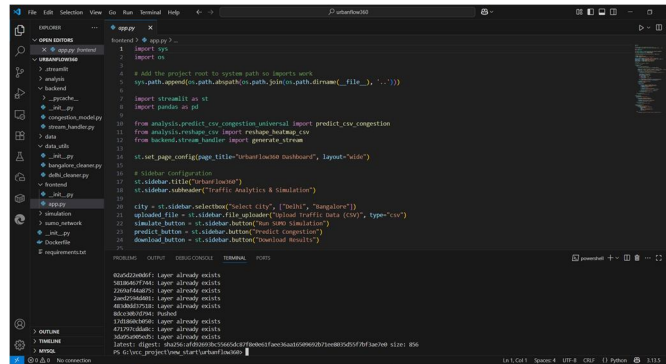


Fig. 8. Docker image pushed to AWS ECR



```

1 # Import libraries
2 import os
3
4 # Add the project root to system path so imports work
5 sys.path.append(os.path.abspath(os.path.join(os.path.dirname(__file__), '..')))
6
7 import streamlit as st
8 import pandas as pd
9
10 from analysis.predict_csv_congestion_universal import predict_csv_congestion
11 from analysis.reshape_csv import reshape_heatmap_csv
12 from backend.stream_handler import generate_stream
13
14 st.set_page_config(page_title="UrbanFlow Dashboard", layout="wide")
15
16 # Sidebar configuration
17 st.sidebar.title("UrbanFlow")
18 st.sidebar.subheader("Traffic Analytics & Simulation")
19
20 city = st.sidebar.selectbox("Select City", ["Bangalore", "Mumbai"])
21 uploaded_file = st.sidebar.file_uploader("Upload Traffic Data (CSV)", type="csv")
22 st.sidebar.button("Predict Congestion")
23 st.sidebar.button("Download Results")
24
25 # Main content area
26 if uploaded_file:
27     df = pd.read_csv(uploaded_file)
28     # Predict congestion
29     congestion = predict_csv_congestion(df, city)
30     # Reshape for heatmap
31     heatmap_data = reshape_heatmap_csv(congestion, city)
32     # Generate stream
33     stream = generate_stream(heatmap_data, city)
34     # Display stream
35     st.image(stream, use_column_width=True)
36
37 # Footer
38 st.write("UrbanFlow360 - Traffic Analytics & Simulation")
  
```

Fig. 9. SSH terminal on EC2 running Docker container

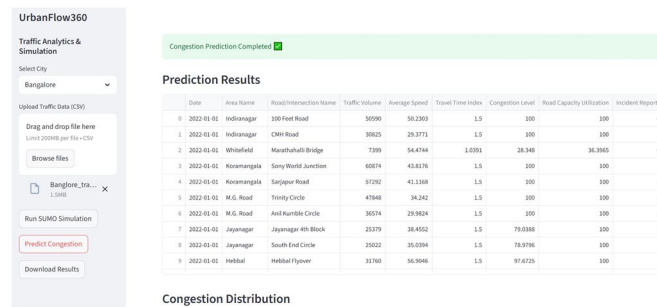


Fig. 10. Deployed dashboard accessed via EC2 public IP

Area	Avg Speed (km/h)	Predicted Congestion (%)
Koramangala	43.8	92.1
M.G. Road	29.9	85.5
Whitefield	54.5	21.2

TABLE I

SAMPLE CONGESTION PREDICTIONS FOR BANGALORE



Fig. 11. Bar chart visualization of congestion levels

## XI. COMPARISON WITH EXISTING TOOLS

- 1) Dashboards like TomTom offer historic data, not predic- tions.
- 2) Our tool allows CSV uploads + real-time ML predictions.
- 3) Dockerized, portable, and open-source.



## XII. CONCLUSION

UrbanFlow360 bridges cloud computing and urban mobility analytics. It is portable, scalable, and accessible to both cities and researchers. Future work includes multi-city support, live APIs, and SUMO integration for simulation.

## REFERENCES

- [1] DLR SUMO - <https://sumo.dlr.de/docs/>
- [2] AWS Docs - <https://docs.aws.amazon.com/>
- [3] Streamlit - <https://docs.streamlit.io/>
- [4] Scikit-learn - <https://scikit-learn.org>
- [5] Docker Docs - <https://docs.docker.com/>
- [6] GitHub Repo - <https://github.com/KishanBouri/urbanflow360>



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