

# Advanced Analytics for Traffic Safety: Key Metrics, Predictive Insights, and High-Risk Junction Identification

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## Introduction

**Road safety** remains a critical global issue, with over **1.19 million deaths** annually. In Europe, progress is uneven, despite initiatives such as the EU Vision Zero, which aims to halve road deaths by 2030.

**Traditional crash-based analyses** are retrospective and limited by **underreporting** and low **data granularity**, requiring crashes to occur before action is taken.

Current road safety measures show **slow progress**, necessitating **new** approaches for **crash prediction and prevention**.

**Unsafe traffic events**, such as harsh accelerations and braking, occur **more frequently** and can be easily **identified** using smartphone app data.

**Leveraging real-time data** from smartphone sensors offers a **proactive approach to traffic safety analysis and intervention**.

## Objectives

- 1. Identify** high-risk urban junctions using smartphone driving data and traffic metrics.
- 2. Evaluate** the impact of speed variability, braking, and traffic volume on crash risk.
- 3. Develop** a predictive framework to guide proactive traffic safety interventions.

## Methodology

### 1. Study Area

- Location: Central Athens, Greece



Figure 1: Study network of Athens

### 2. Data Sources

- Telematics Data:** Collected via **OSeven smartphone app** (unsafe driving behaviors).
- Traffic Management Center & Google Maps** (traffic volume, road characteristics).
- Spatial mapping via **QGIS**.

### 3. Data Processing

- Statistical Modelling:** GLM & GLMM for crash–event association.
- Clustering:** K-Means, DBSCAN to group high-risk junctions.
- Predictive Modelling:** Random Forest for crash predictors (feature importance).
- Spatial Analysis:** Moran's I, Geary's C for unsafe event hotspots.
- Dimensionality Reduction:** PCA to optimize performance.

## Results

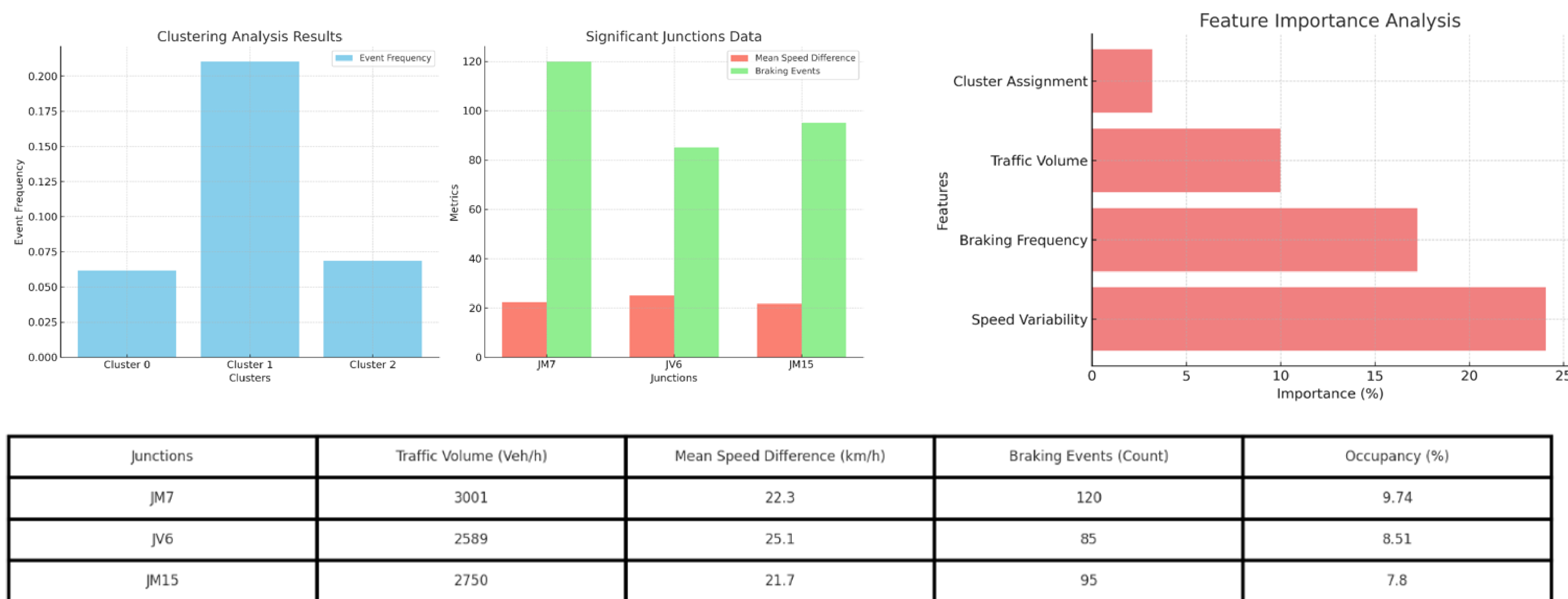


Figure 2: Results of analysis

- 1. Clustering Analysis:** 3 groups identified → Cluster 1 = highest risk (high speed variability, frequent braking).
- 2. Significant Junctions:**
  - JM7, JV6, JM15 → high traffic volume, speed differences, and braking events (see figure).
- 3. Feature Importance:**
  - Top predictors:** Speed variability (24%), braking frequency (18%), traffic volume (10%).
  - Spatial Hotspots:** Unsafe driving clusters overlapped with crash-prone areas.

## Conclusions

- 1. Unsafe driving behaviours** (especially harsh braking & speed fluctuations) are **strong predictors of crash risk**.
- 2. Smartphone** app data combined with ML provides a **proactive safety monitoring tool**.
- 3. Recommendations:**
  - Speed management enforcement.
  - Driver education via feedback apps.
  - Infrastructure improvements (clearer signage, optimized junction designs).
- 4. Future work:** Extend to broader regions, integrate weather and real-time traffic updates.

