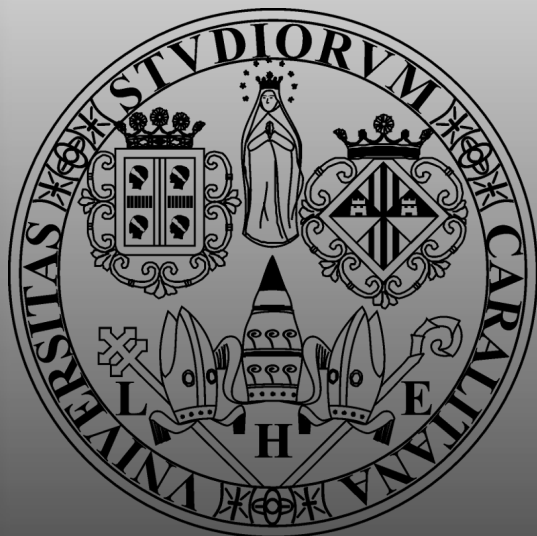


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Roadwatch: An Integrated Architecture for AI-Powered Surveillance and Anomaly Detection in Traffic Areas

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Outline

- **Overview**

Introduction ▪ Related Work ▪ Main Challenges

- **Intuition**

Idea ▪ Key Features

- **Formalization**

Architecture ▪ System Description

- **Evaluation**

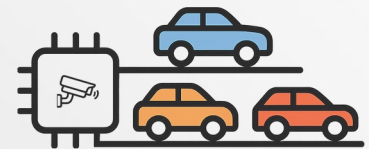
Configuration ▪ Experiments ▪ Conclusions and Future Work

OVERVIEW

Introduction



- Urban traffic complexity demands advanced **monitoring solutions**
- Traditional human surveillance appears **inadequate**
- *Roadwatch*: integrates **Artificial Intelligence** and **Computer Vision** for **real-time** monitoring
- It leverages **existing** camera infrastructure

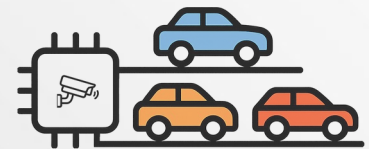


Introduction



The main *Roadwatch* **objectives** are:

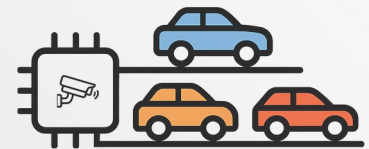
- **Enhance** surveillance in traffic areas
- **Detect** anomalies involving vehicles and pedestrians
- **Provide** real-time alerts and support decision making
- **Optimize** urban traffic management



State of the Art



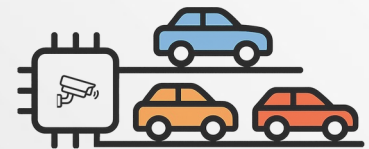
- *Smart Cities* **increasingly** rely on data-driven systems to address challenges in urban mobility, safety, and resource optimization
- **Artificial Intelligence (AI) and Machine Learning (ML)** are widely applied to enable:
 - *Vehicle and pedestrian detection* using computer vision techniques (e.g., YOLO)
 - *Anomaly and event recognition*, such as:
 - Illegal parking
 - Traffic congestion
 - Dangerous driving behaviors
 - Pedestrian intrusion in restricted areas



State of the Art

These **AI-powered systems** enable:

- **Real-time analysis** of video streams from urban surveillance networks
- **Predictive insights** to anticipate traffic conditions
- **Proactive interventions** to improve safety and efficiency



Main Challenges

Most state-of-the-art solutions **rely on custom** camera hardware, which introduces key limitations:

- Specific **technical requirements**, such as *resolution, frame rate, and sensor type*
- **Constrained positioning** to ensure an *optimal field of view* and accurate detection



INTUITION

Idea



To **overcome the limitations** of state-of-the-art solutions requiring custom camera hardware, We introduce a flexible, hardware-agnostic approach:

- **Leverages existing video** surveillance infrastructure (e.g., standard IP cameras)
- **No need for specialized** devices or **fixed camera placement**
- **Compatible with various resolutions** and orientations via configurable video preprocessing



Key Features



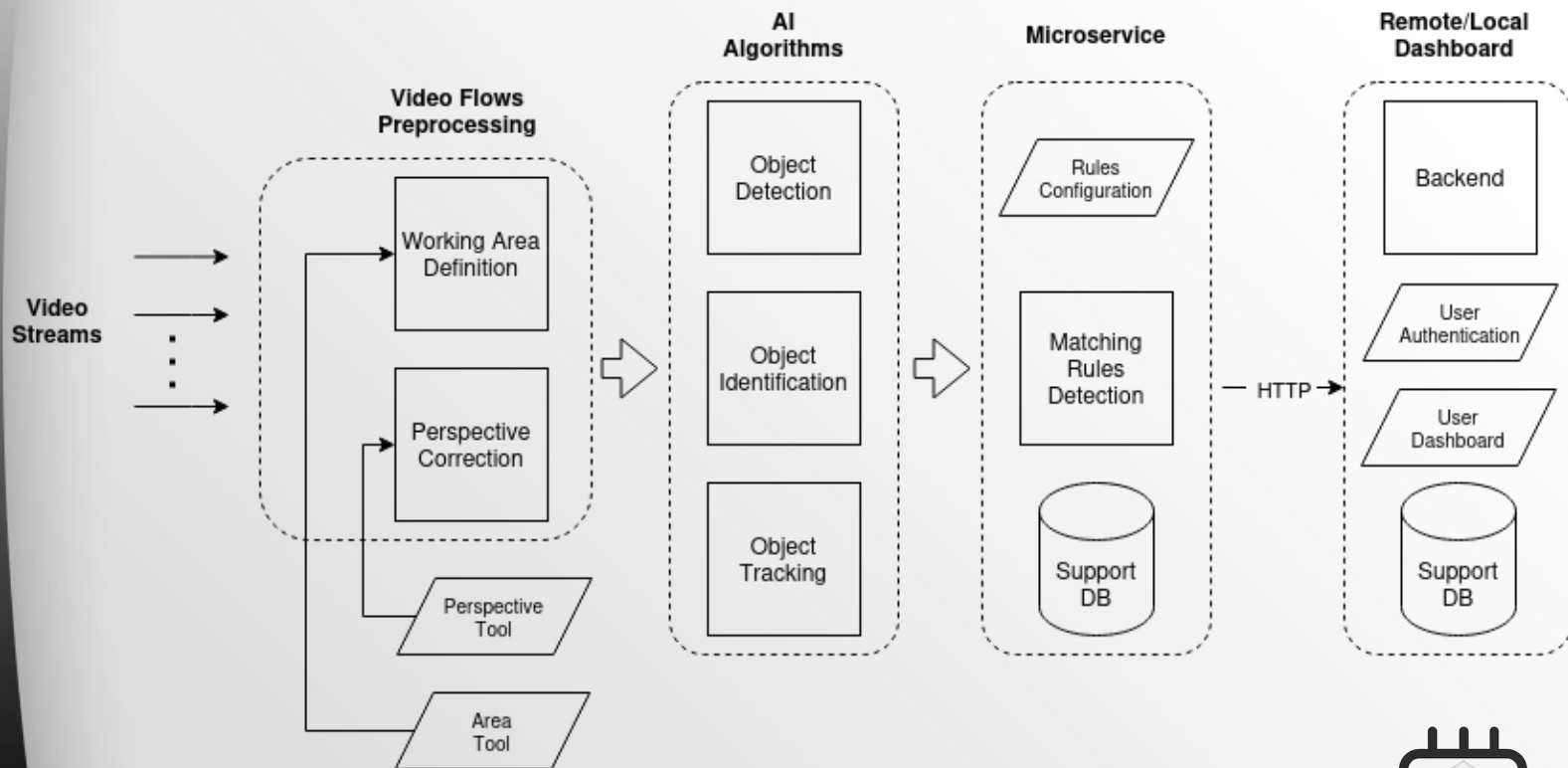
Key features:

- **Perspective correction** and region-of-interest selection via **web tools**
- **Modular pipeline** built on NVIDIA DeepStream and YOLOv4
- **Real-time detection and tracking** of traffic anomalies across diverse environments
- Enables **scalable deployment** in real-world Smart City contexts



FORMALIZATION

High-level System Architecture



Traffic Surveillance Problem Definition



Goal: Detect anomalous events involving vehicles or pedestrians

- Surveillance area represented as a **working area** segmented into **zones**
- Each **zone** associated with a semantic role (e.g., *crosswalk*, *road lane*, *restricted area*)
- Time is modeled discretely, with observations **processed frame by frame**



Object Representation and Tracking



Each **detected object** is represented by:

- A unique ID
- Position and bounding box
- Semantic class (e.g., car, person)

The **state** of an object is **tracked** over time as it moves across zones

Tracks are **sequences of object observations** with time annotations



Zone Interaction Model



Events are **detected based on interactions** between tracked objects and zones

Examples of **zone interaction**:

- *Person entering restricted area*
- *Vehicle stopping in no-parking zone*

System evaluates object transitions across zones to **identify rule violations**



Event Modeling and Detection Rules



Events are formalized as predicates on **sequences of object states**

Each **event type** defined by:

- **Preconditions** (e.g., object enters specific zone)
- **Temporal constraints** (e.g., object remains in zone for N seconds)

This allows for **flexible rule definition** without the need to retrain AI models



Benefits of the Proposed Approach



- **Decouples** *detection logic* from raw *AI inference*
- Enables **precise rule-based** reasoning on tracked object behavior
- Facilitates **scalability** and **reconfigurability** for different environments
- Foundation for **modular event detectors** within Roadwatch pipeline

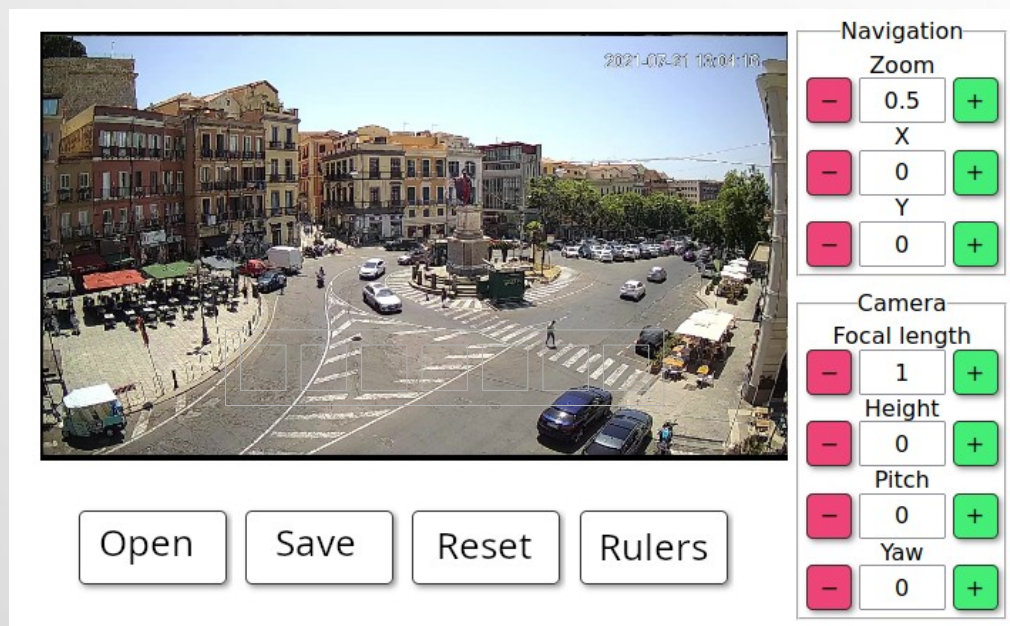


Evaluation



Configuration: One-Time Perspective Correction

- The system works with **pre-installed cameras**
- A **one-time perspective correction** is required to align the camera view with the detection model
- **No additional hardware** or camera repositioning needed





Configuration: One-time Defining Detection Zones

- Mark **areas of interest**
- **Rules** are applied only within these defined zones
- This setup enables **accurate and customized** detection

The interface displays a street scene with various detection zones marked by colored polygons and numbers. On the left, there are buttons for 'IDS' and an upward arrow. On the right, there are buttons for 'ADD', 'SELECT', 'DIRECTIONS', 'TYPES', 'SELECT VIDEO', 'LOAD CSV', and 'DOWNLOAD'. Below the video feed is a table with configuration details for each zone.

	ID	luX	luY	ruX	ruY	rbX	rbY	lbX	lbY	CarWalk	Left	Up	Right	Down	Parking	Stop
	35	550	795	0	685	516	601	592	716	☐	☐	☐	☐	☐	☐	None
	36	550	795	422	922	0	1008	0	685	☐	☐	☐	☐	☐	☐	None
	37	1466	1080	1502	998	1920	1080	1615	1080	☒	☐	☒	☐	☐	☒	None
	38	1381	1080	1501	922	1502	998	1466	1080	☒	☐	☒	☐	☐	☒	None



Preliminary Experiments

We taken into account **three video scenarios**:

- *Urban street*
- *Roundabout*
- *Parking area*

Events tested:

- *Congested Traffic*
- *Illegal Lane Change*
- *Person on Road*
- *Area Invasion*
- *Illegal Parking*
- *Reckless Driving*

We used real-time **AI analysis + manual validation**



Experiments

Evaluation Results:

- Total events: **634**
- Correctly detected: **578**
- Accuracy: **~91%**

Observations:

- **Strong** performance on Area Invasion, Reckless Driving
- **Lower** on occluded or distant objects (e.g. Person on Road)





Experiments

Validation Results for Event Detection by Camera:

Event Type	Camera #1		Camera #2		Camera #3		Total Accuracy
	Correct	Missing	Correct	Missing	Correct	Missing	
Congested Traffic	40	3	35	4	30	5	89.74%
Illegal Lane Change	28	5	35	2	35	1	92.45%
Person on Road	25	7	30	4	35	4	85.71%
Area Invasion	30	2	35	0	35	0	98.03%
Illegal Parking	33	3	31	5	26	4	88.23%
Reckless Driving	35	1	30	3	30	3	93.13%
All Events	191	21	196	18	191	17	91.16%





Conclusions

- Roadwatch demonstrates the **effectiveness** of a **modular AI architecture** for urban surveillance, operating on existing video infrastructure with **no need for specialized** camera hardware
- The system integrates video preprocessing, object detection and tracking, and rule-based anomaly recognition, enabling real-time detection of events such as:
 - Area intrusions
 - Illegal parking
 - Traffic congestion
 - Risky driving behavior
- Preliminary results show an overall **accuracy of ~91%**, with strong performance on zone-based and behavioral anomalies, confirming the robustness of the approach.





Future Work

The Roadwatch system requires **further optimization** to enhance performance and scalability.

In addition, our next **development directions** include:

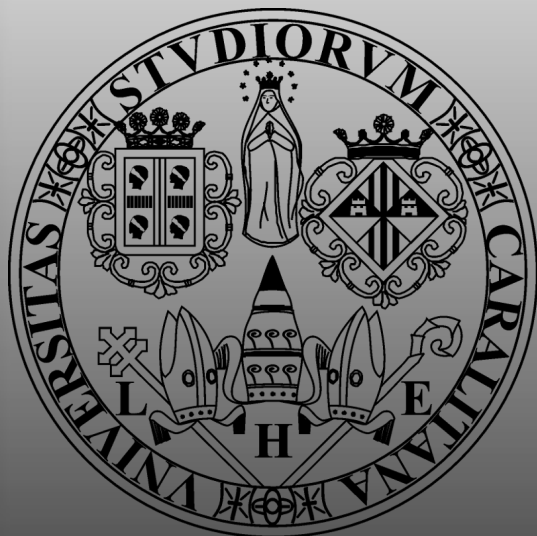
- AI algorithm refinement to **improve resilience** in complex environments, including occlusions and varying lighting or weather conditions
- Dataset expansion and real-world testing to **increase generalization** capabilities and validate performance across diverse urban scenarios
- **Integration** into broader Smart City ecosystems, involving:
 - **Edge computing** platforms
 - **Operator-facing** dashboards
 - Interoperability with **traffic control** and **alerting systems**



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