

# INTELLIGENCE SURVEILLANCE CAMERA

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# Abstract

- The project presents the design and implementation of an intelligent surveillance camera system aimed at enhancing security and monitoring capabilities through realtimes analysis
- The surveillance systems typically employed are CCTV-based monitoring systems which require an administrator to scan the footage or manual monitoring of the area to detect .

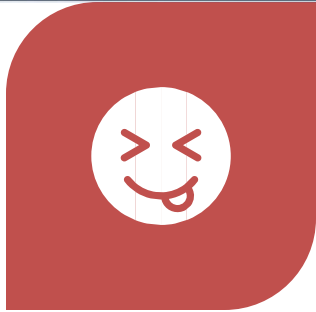


# Introduction

- The intelligent surveillance system uses advanced video analysis technologies, such as object detection, behavior recognition, face recognition, etc., to automatically identify and analyze targets in the surveillance scene.



# Problem Statement



PROBLEM FACED IN AI BIAS      DATA SECURITY RISK AND DATA COLLECTION AND AND  
MISIDENTIFICATION ? CYBER SECURITY RISK?    TRACKING?

# System Requirements

- Hardware: 4GB RAM, 2GHz CPU
- Software: Windows/Linux, Python 3.x, Jupyter Notebook
- Libraries: pandas, numpy, matplotlib, seaborn.





# DATA SOURCE



Fields include: Daily

Dataset downloaded Activity Trends, Peak from  
Kaggle. hour analysis, high risk  
activity, location wise  
security

### 📁 \*Data Source in the Notebook\*

The data source is explicitly defined in the following line:

```
python  
df = pd.read_csv('Intelligent_Surveillance_Dataset.csv')
```

#### ✅ Details:

Attribute	Description
*Format*	CSV (Comma-Separated Values)
*Filename*	Intelligent_Surveillance_Dataset.csv
*Access Method*	Loaded using pandas.read_csv()
*Location*	Expected to be in the same directory as notebook
*Tool Used*	Python pandas library

#### 📌 Summary:

\* The dataset is likely a structured table with columns such as:

- \* Timestamp
- \* People\_Count
- \* Motion\_Score
- \* Alert\_Flag
- \* Activity\_Type
- \* Location

# Architecture

1. DATA 2. DATA 3. DATA 4. QUERY COLLECTION CLEANING  
EXPLORATION ANALYSIS



5.  
VISUALIZATION



6. CONCLUSION  
& REPORTING.

# Data Cleaning

Handled missing  
values & duplicates.

Verified data types  
and ensured proper  
data formatting for  
analysis.

## #### 1. \*Date-Time Conversion\*

```
python
df['Timestamp'] = pd.to_datetime(df['Timestamp'])
```

\* Converts timestamp to a proper datetime format.

---

## #### 2. \*Feature Engineering from Timestamp\*

```
python
df['Hour'] = df['Timestamp'].dt.hour
df['Day'] = df['Timestamp'].dt.day
df['DayOfWeek'] = df['Timestamp'].dt.day_name()
df['Date'] = df['Timestamp'].dt.date
```

## #### 3. \*Categorization and Mapping\*

### \* \*Day/Night Classification\*

```
python
df['Time_Period'] = df['Hour'].apply(categorize_time)
```

### \* \*Weekend Identification\*

```
python
df['IsWeekend'] = df['DayOfWeek'].isin(['Saturday', 'Sunday'])
```

### \* \*Crowd Size Labeling\*

```
python
df['Crowd_Size'] = df['People_Count'].apply(categorize_crowd)
```

### \* \*Activity Type Mapping\*

```
python
df['Activity_Category'] = df['Activity_Type'].apply(lambda
x: ...)
```



- Peak Hours Analysis - When is the facility busiest?

- Location-wise Security Analysis

- High-Risk Activity Detection

- Temporal Pattern Analysis - Day vs Night Activity

- Crowding Events Analysis

- Motion Score Anomaly Detection

- Daily Activity Trends

- Alert Correlation Analysis

- Loitering Incidents Analysis

# Sample Queries (1-10)

- Running Activity Security Assessment

- Weekend vs Weekday Patterns

- Low-Occupancy High-Alert Events

- Activity Type Transition Patterns

- Motion Score vs People Count Relationship

- Location Security Risk Ranking

- Time-based Alert Clustering

- Standing vs Moving Activity Analysis

- Temporal Alert Prediction Indicators

- COMPREHENSIVE SECURITY DASHBOARD SUMMARY

## References



Kaggle Dataset



Python Libraries  
Documentation



Impact of intelligence  
surveillance camera on  
society.