

# Project Title:

## Smart Traffic Management System + Solar Powered Street Lighting

### Presenter's Name:

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## Project Idea:

### Problem Description:

- Modern cities suffer from uncontrolled traffic congestion and high energy waste from street lighting.
- Lack of real-time monitoring causes delays, accidents, and unnecessary power consumption.

### Proposed Solution:

- A fully automated real-time data pipeline:
- IoT Sensors → MQTT → Kafka → Spark Streaming → ML Models → Smart Dashboard.
- Predict traffic, detect congestion, generate alerts, and optimize lighting consumption.

### Unique Value Proposition:

- Real-time streaming data processing.
- Machine-learning-based traffic prediction (Milestone 4).
- Automated energy-saving lighting control.
- End-to-end scalable pipeline using Big Data technologies.

## Project Wireframe:

### User Interfaces & Workflow:

- Dashboard Home: live monitoring of traffic speed, count, congestion, and energy status.
- Alerts Panel: shows real-time warnings (High congestion / High lighting load).
- Analytics View: visualization of processed windowed results and historical trends.
- Admin Tools: view predicted next-minute traffic + predicted lighting levels.

### User Journey

1. System ingests city sensor data through MQTT.
2. Kafka stores and streams messages.
3. Spark processes data and updates live dashboards.
4. ML model predicts future traffic + lighting demand.
5. Dashboard displays analytics + actuator triggers lighting adjustments.

## End Users + Features:

### Primary End Users:

- City Traffic Management Authority
- Smart City Energy Department
- Emergency Control Units
- Urban Planners & Data Analysts

### Key Features:

- Live traffic metrics (count, speed, congestion level).
- Real-time alerts (traffic congestion / energy overuse).
- Predictive analytics for upcoming traffic flow.
- Automated lighting load estimation & optimization.
- Data lake and SQL storage for historical analysis.

### How Features Solve User Problems:

- Reduces traffic jams using predictions + alerts.
- Reduces electricity waste using dynamic lighting control.
- Improves decision-making with accurate data and trends.
- Enables long-term planning based on historical data.

## Data Structure:

### Database Architecture

- **Hybrid Structure:**

- SQL Server for structured cleaned data (3 main fact tables).
- Parquet Data Lake for streaming results + raw analytics.

### Main Entities (SQL):

1. traffic\_sensors\_data
2. traffic\_weather\_conditions
3. traffic\_energy\_analysis

### ERD Relationships:

- All three tables linked by:
- timestamp + street\_name (composite natural key).

### Data Flow:

MQTT → Kafka → Spark Structured Streaming →  
→ Cleaned tables (SQL Server) + Parquet data lake  
→ ML Model (Milestone 4) → Dashboard Insights

## Programming Languages + Frameworks:

### Languages:

- Python (Spark, ML, automation scripts)
- SQL (ETL, transformations, SSIS)
- YAML (configurations)

### Frameworks & Tools:

- Apache Kafka (real-time messaging)
- Apache Spark Structured Streaming (processing)
- MQTT (Paho client) (IoT ingestion)
- Docker Compose (orchestration)
- SSIS (ETL to SQL Server)
- Joblib + Scikit-Learn + LightGBM (ML models)
- Jupyter notebook (development)

### Supporting Technologies:

- HiveMQ Public Broker
- Python Kafka-Producer
- Docker Volumes for persistence
- Parquet storage format

```
[{"street_name": "El-Tahrir",  
 "vehicles_per_window": 92,  
 "avg_speed": 18.4,  
 "alert_level": "HIGH",  
 "is_alert": true}
```

## Live Application + Test:

### Current Live System State:

Fully running pipeline on Docker & Spark.  
Real-time streaming from MQTT → Kafka → Spark → SQL/Parquet.  
ML prediction pipeline works on incoming live data.

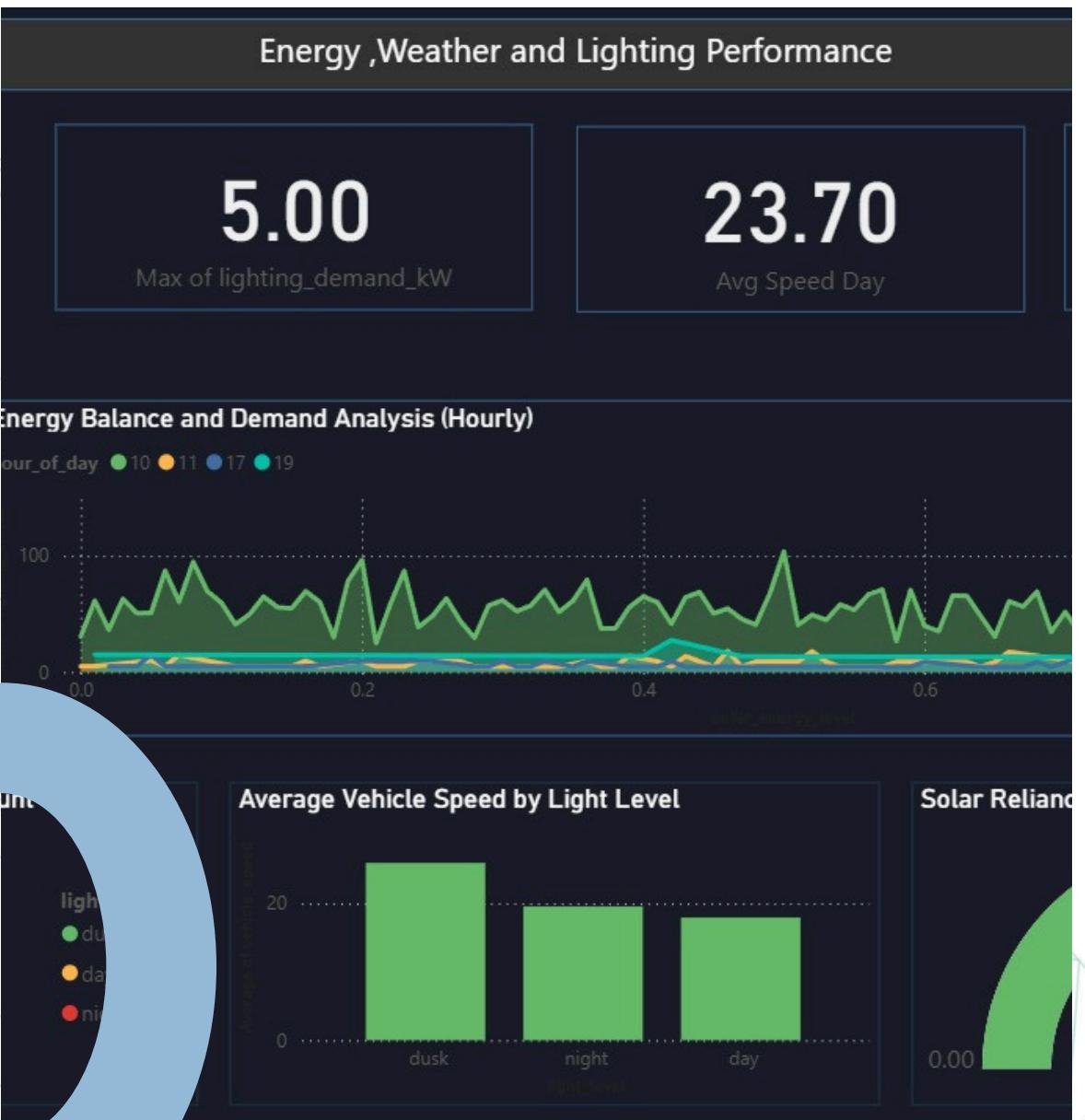
### Testing Phases:

Unit Testing: message parsing, schema validation, ML inference.  
Integration Testing: MQTT → Kafka → Spark end-to-end.  
User Testing: dashboard validation & alert monitoring.  
Load Testing: Kafka throughput; Spark micro-batch size.

### Feedback

Stable under continuous streaming.  
Low latency (<1.5 sec).  
Prediction accuracy validated across test dataset.

## Deliverables



### Provided Deliverables:

System Documentation (architecture, ERD, data flow).  
Source Code (full project ZIP).  
Live Running Application (dashboard + streaming).  
SQL Database Backup + SSIS ETL Package.  
ML Models + Evaluation Reports.  
Parquet Data Lake outputs.  
PPT Presentation (this one).

### Timeline (Milestones):

Data Ingestion & Kafka setup  
Spark Transformations  
Data Lake + SQL ETL  
Predictive Models  
Dashboard + Final Integration

## Team Members + Roles:

### Team Lead

**Hassan Gamal Ghanem** —(millstone 1 , millstone 3)



### Members

- **Maya Yaser Amin**— (millstone 2, millstone 3)
- **Mohammed Mohammed Sobhy** —(millstone 2, dashboard)
- **Amr Mohammed Youssef**—(millstone 1, Testing & documentation)
- **Radwa Hany Sobhy**—(millstone 2, millstone 4 )
- **Habiba Ashraf Elboghdady** —(millstone 2)

### Collaboration Methods

GitHub

Trello for task management

Agile method (Weekly Sprints)

WhatsApp coordination

# Thank You Slide

## Thank You!



Feel free to ask questions.

For inquiries or further details:

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