

# Real-Time PPE Detection System

## Hackathon Team 2

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# Problem statement

- **PPE non-compliance** leads to accidents, financial losses, and legal penalties.
- **Manual PPE checks** are slow, error-prone, and hard to scale in large industries
- **Existing AI solutions** often lack scalability, efficiency, and real-time alerts.

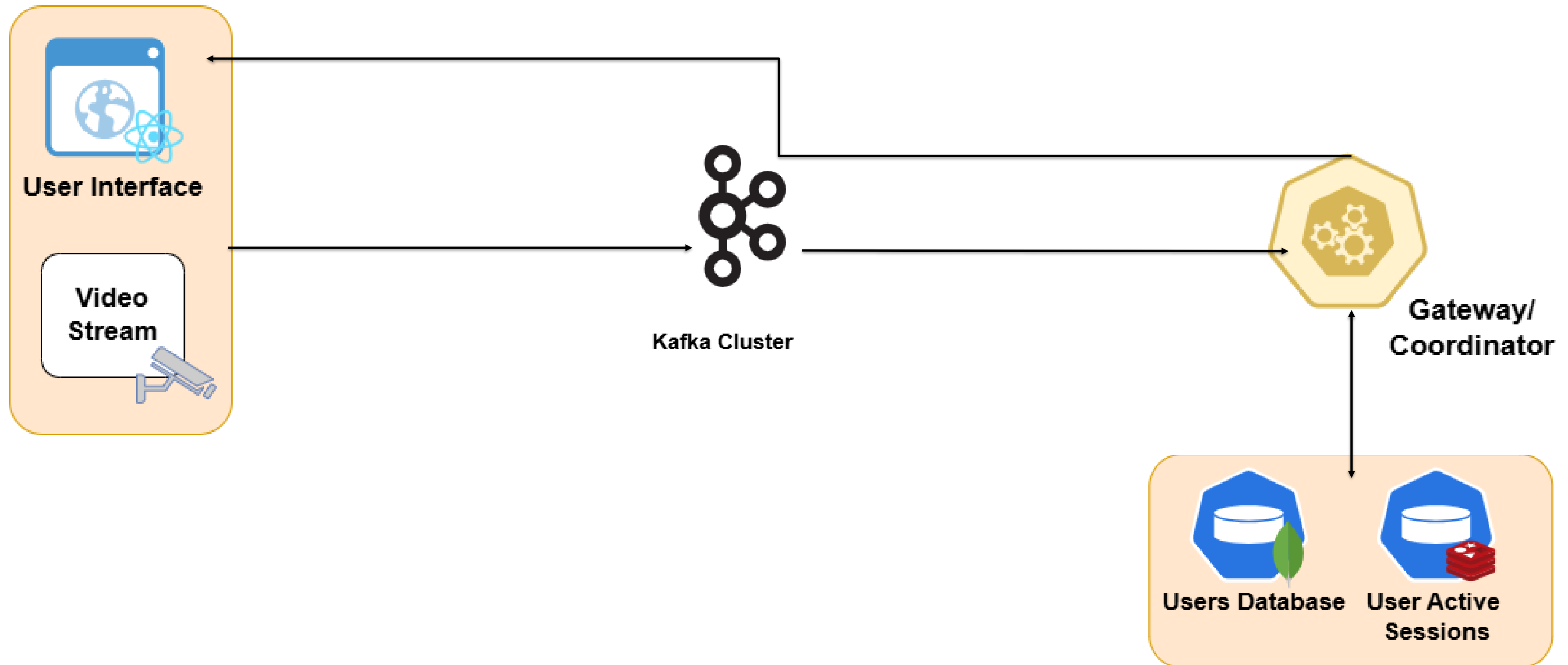


# Solution

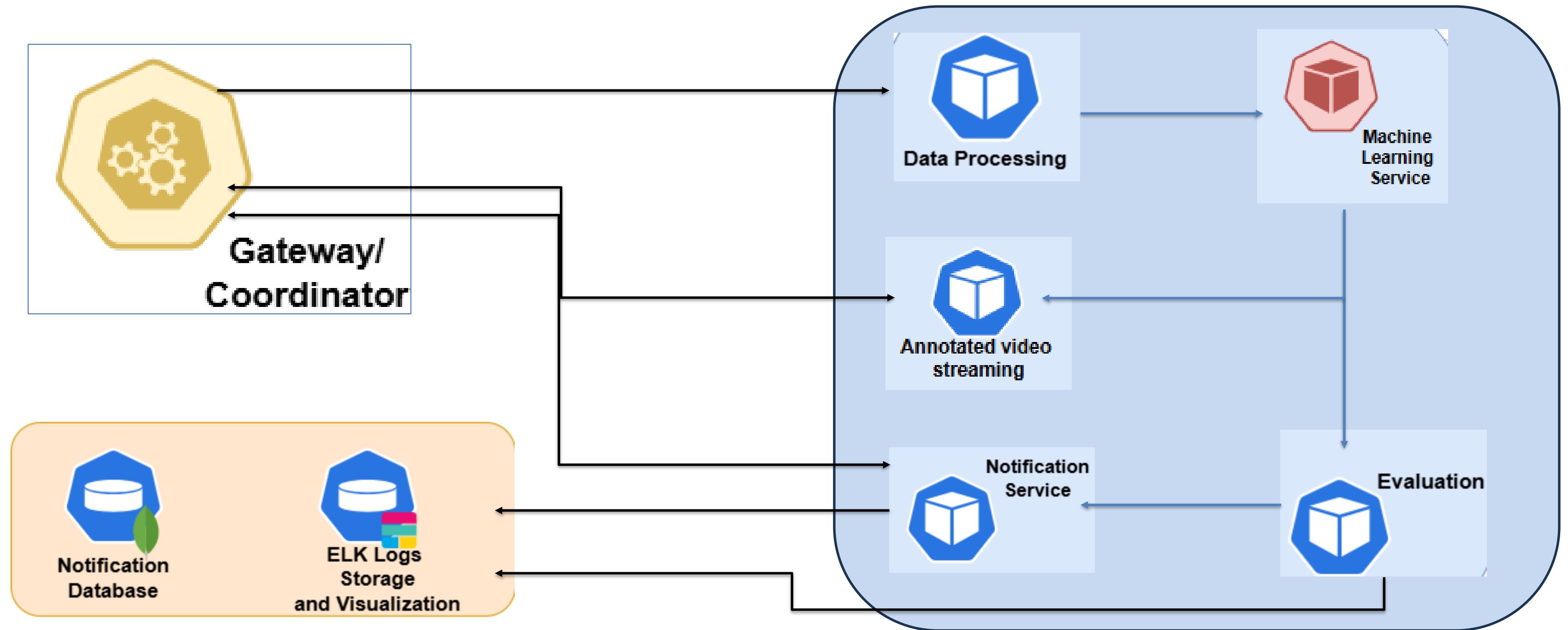
“Our computer vision-based PPE detection solution follows a robust architecture that ensures real-time, scalable, and efficient monitoring, ensuring safety compliance with instant alerts and low-latency processing.”



# ARCHITECTURE

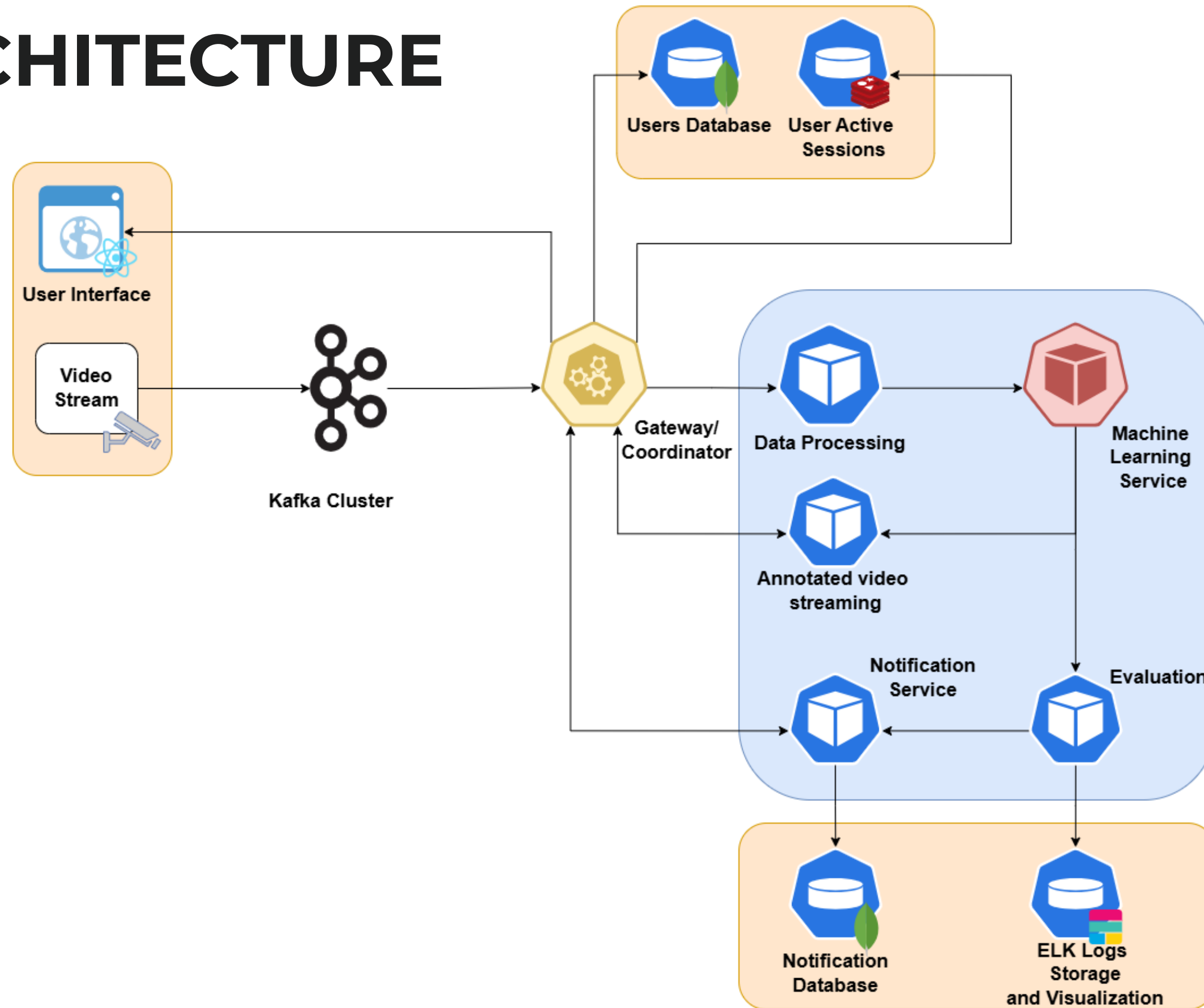


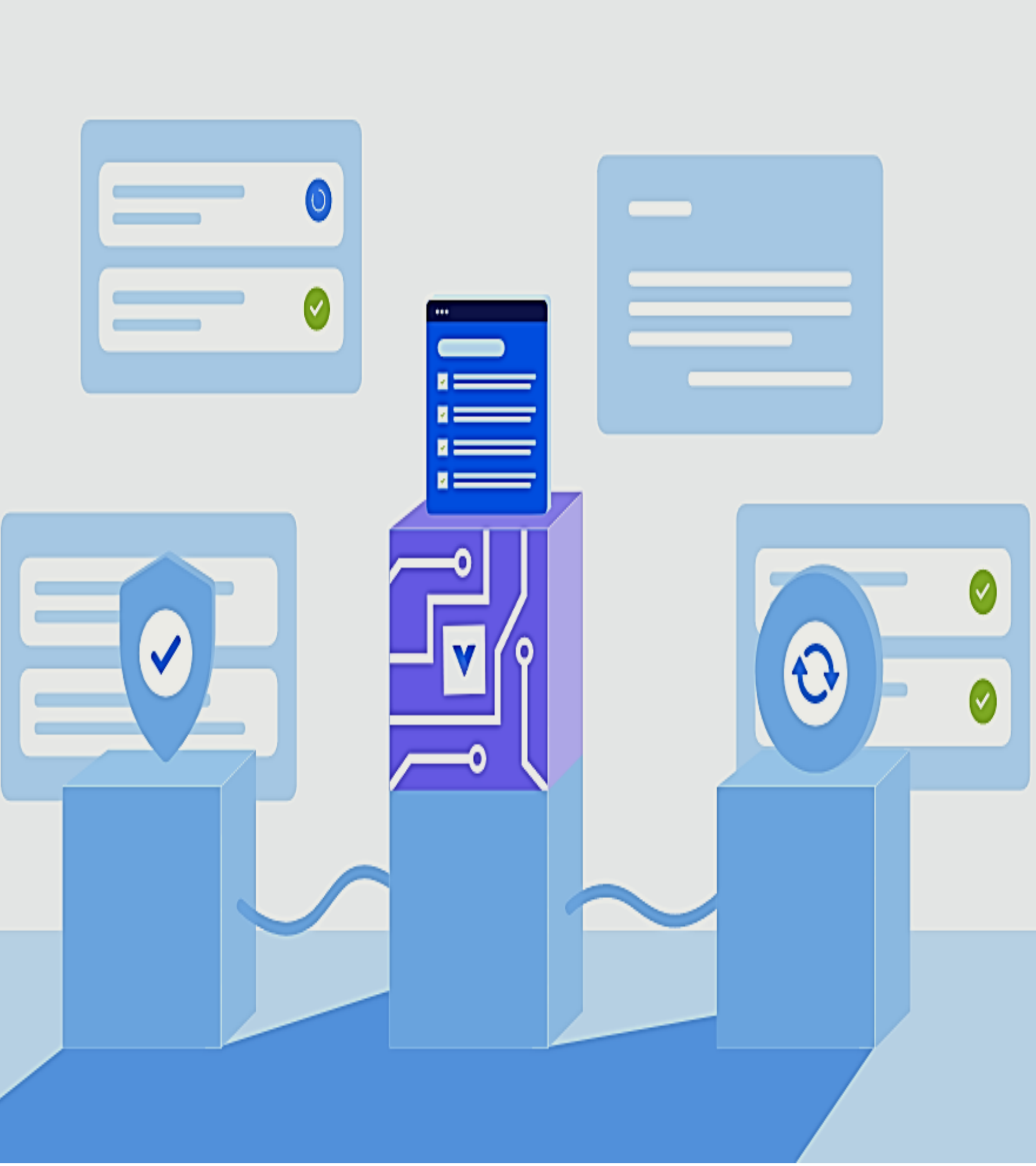
# ARCHITECTURE





# ARCHITECTURE





## Rationale

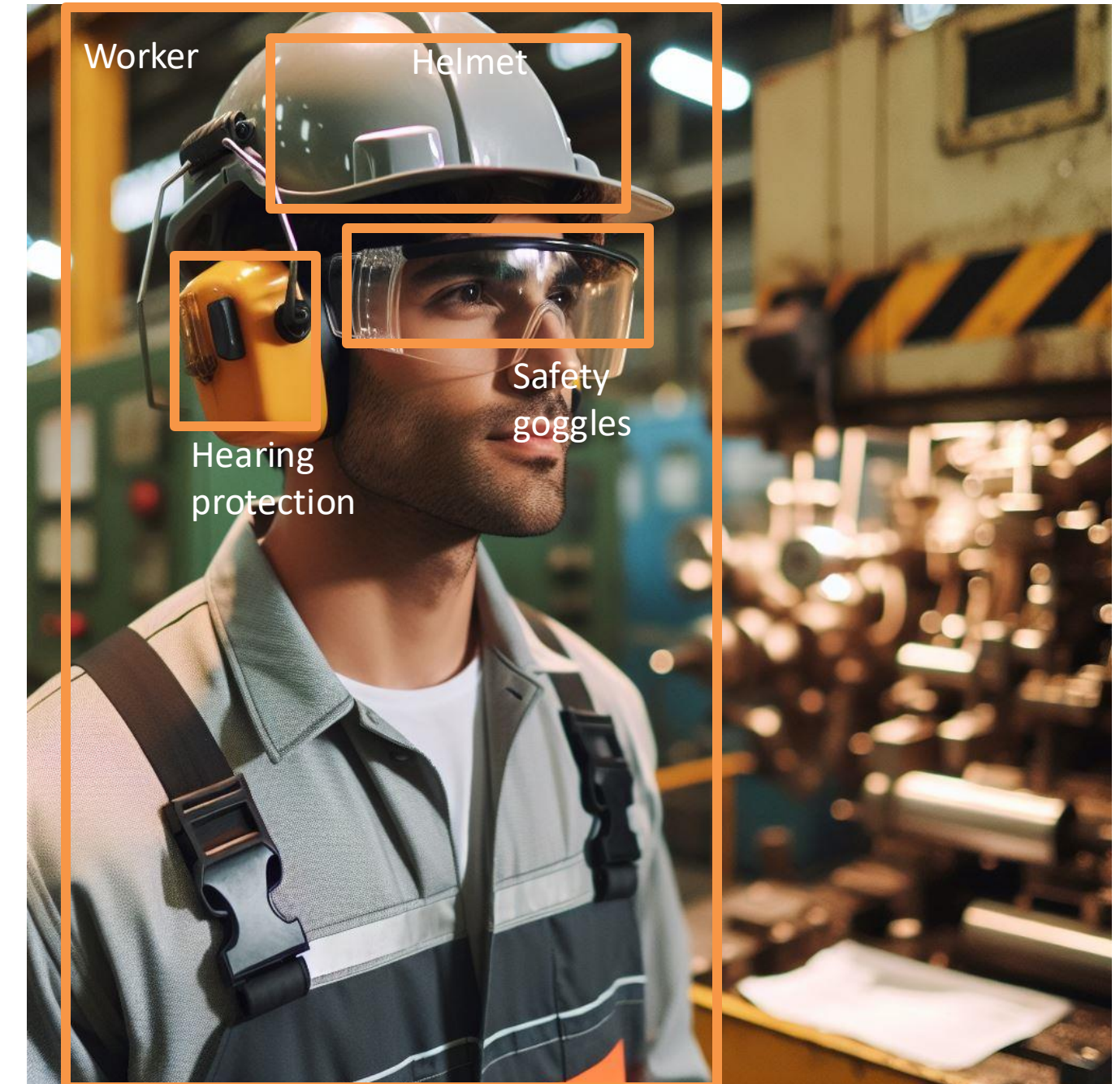
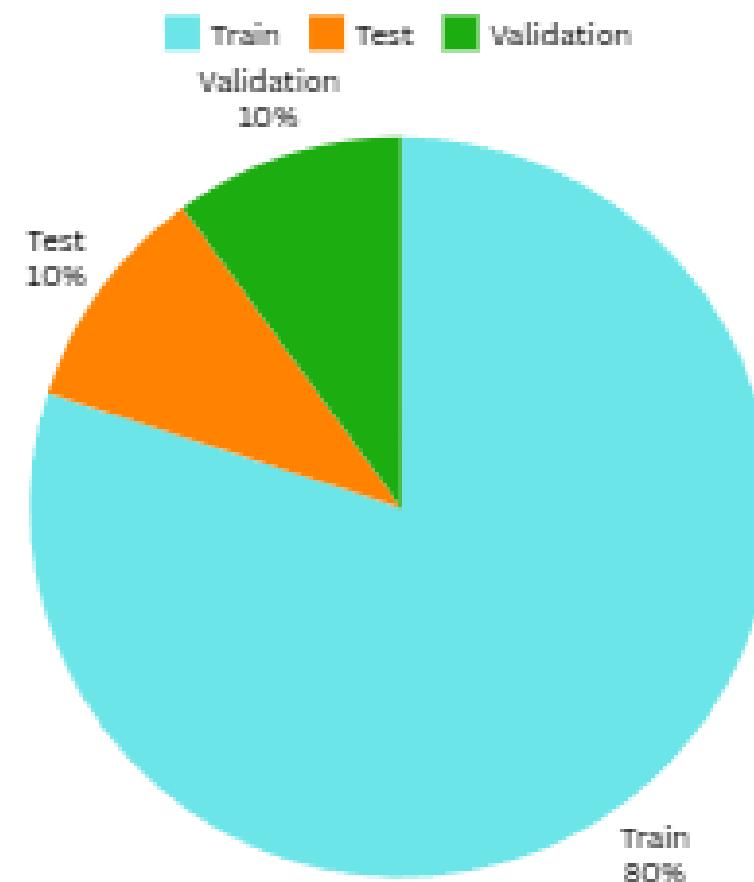
- Scalability
- Load balancing
- YOLOv8 is optimized for real-time inference
- Kafka ensures real-time, scalable, and fault-tolerant video stream processing.

# Dataset

- 536 images of PPEs and workers with/without PPEs
- Duplicated were removed --> 445 images
- All images were resized to 640x640 (for YOLO training)
- Images were manually annotated using Labellmg
- 80/10/10 : train/validation/val

- **Classes**

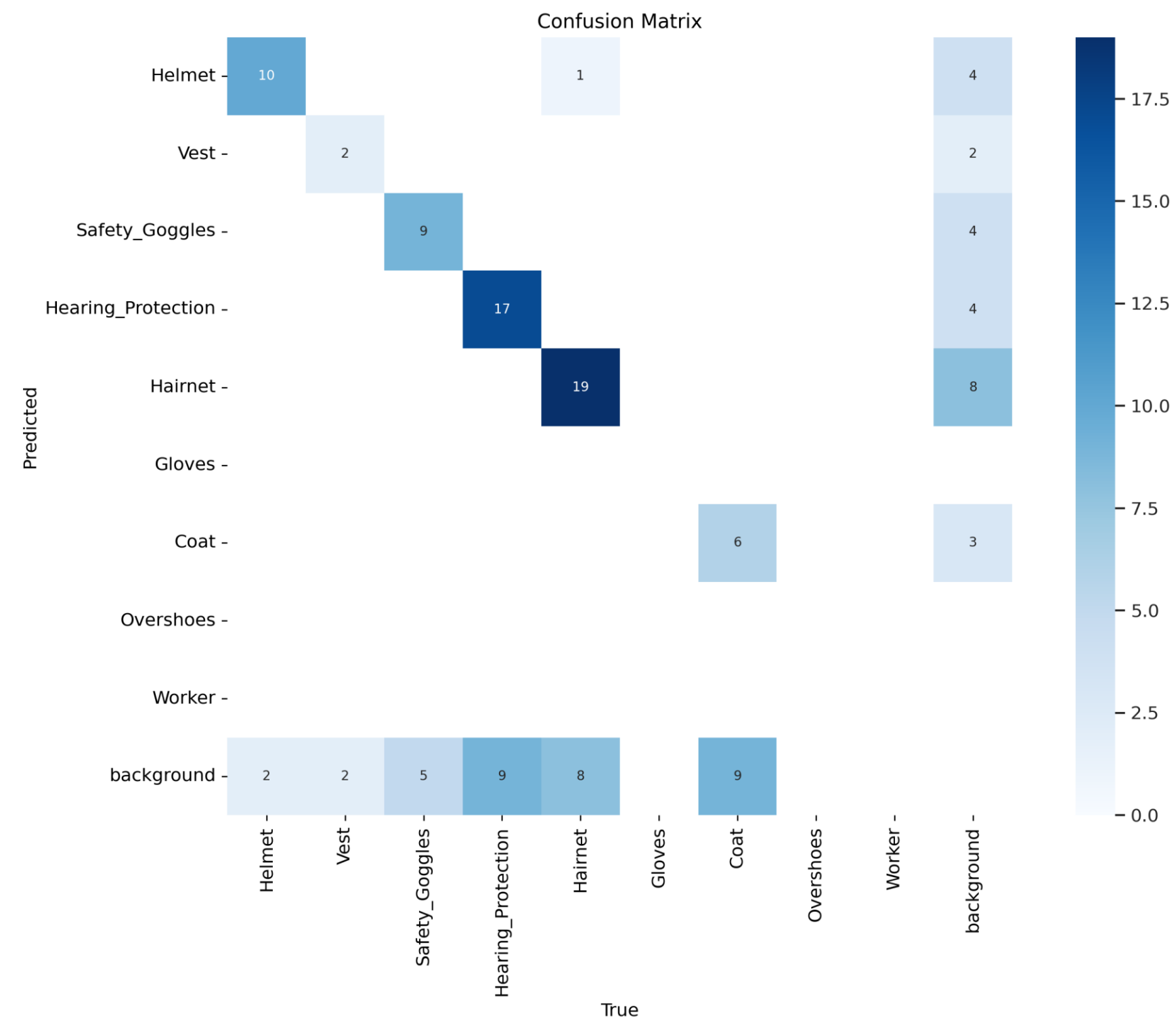
- "Helmet",
- "Vest",
- "Safety\_Goggles",
- "Hearing\_Protection",
- "Hairnet",
- "Gloves",
- "Coat",
- "Overshoes",
- "Worker"

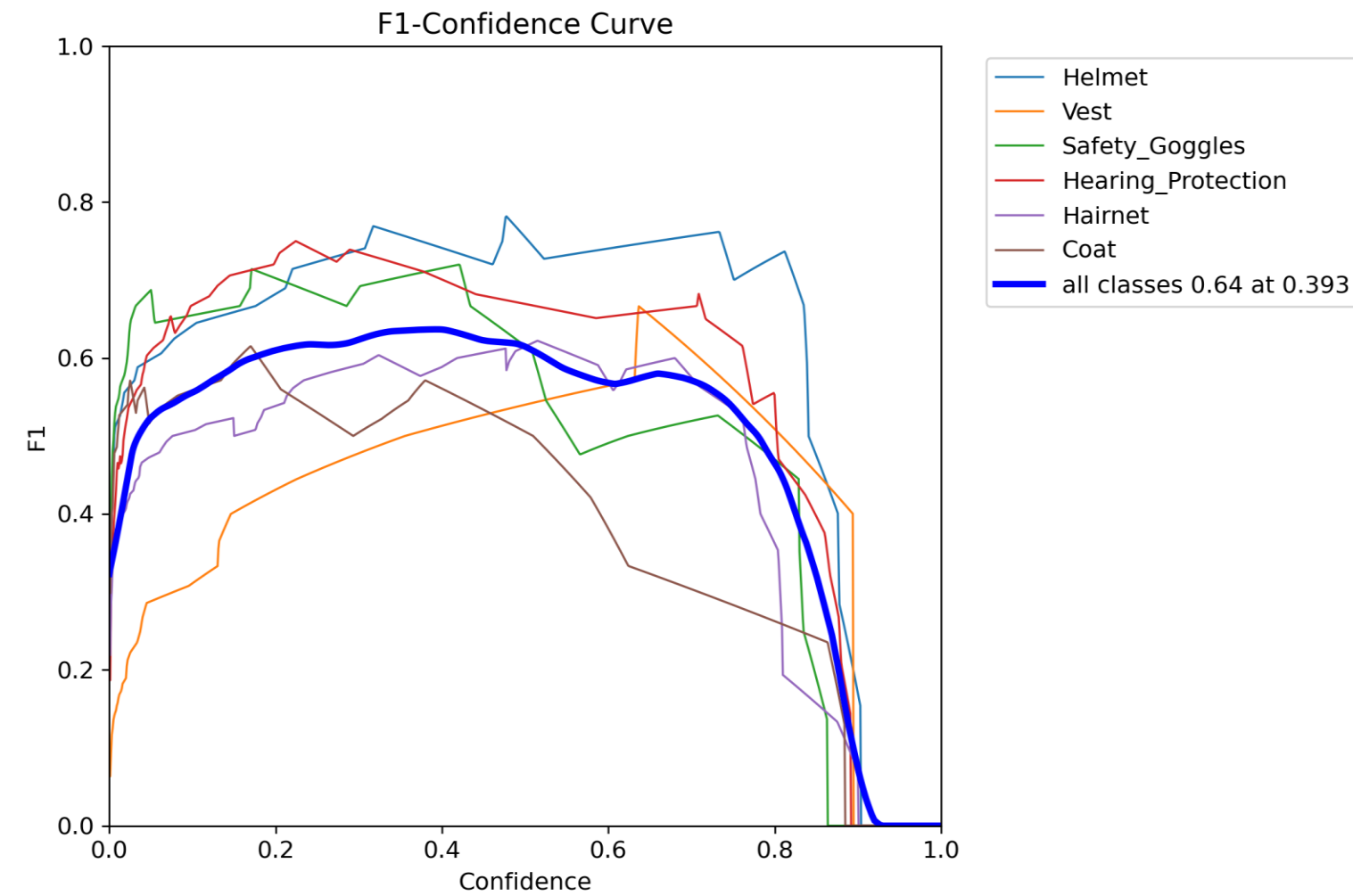
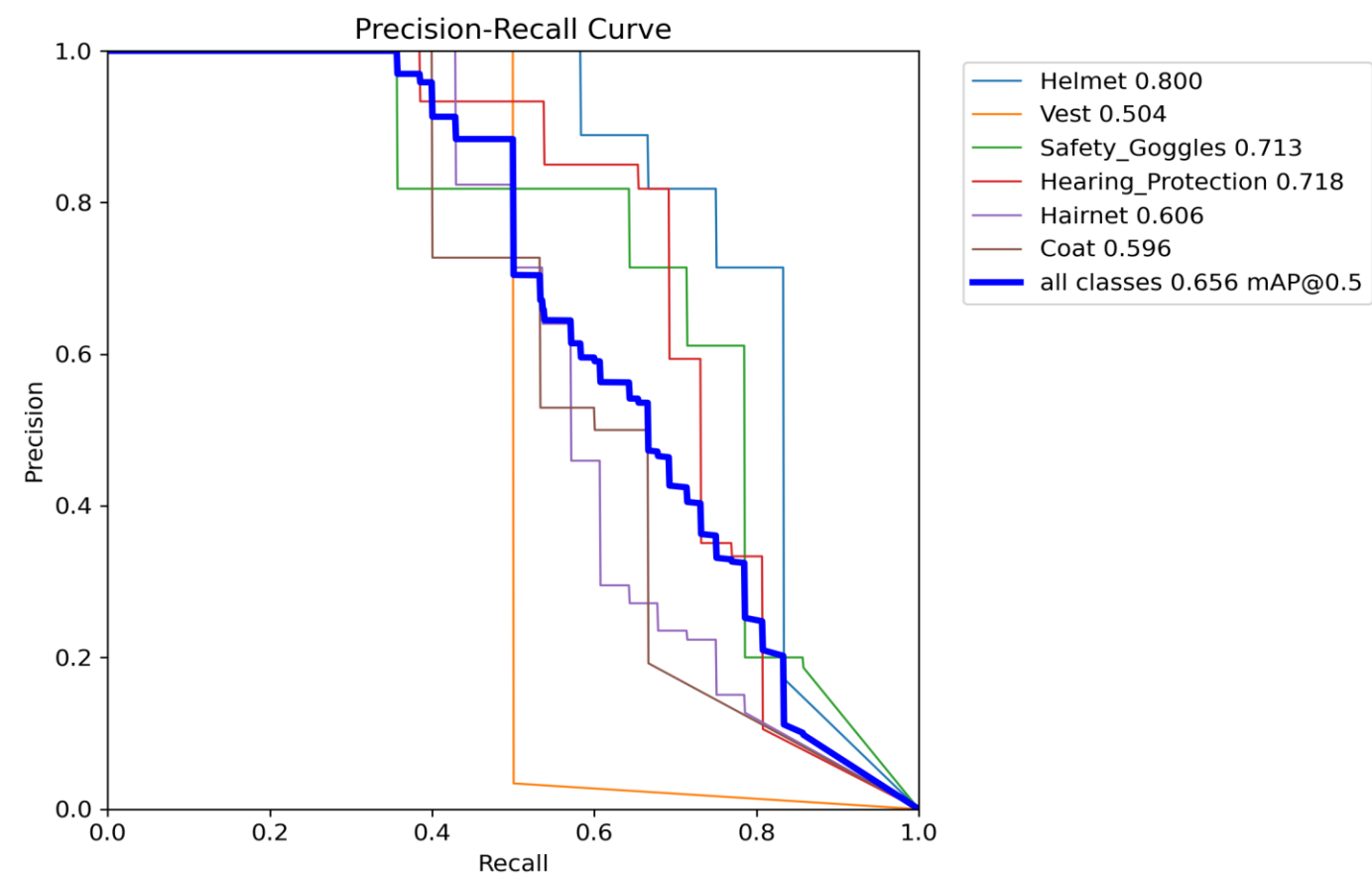
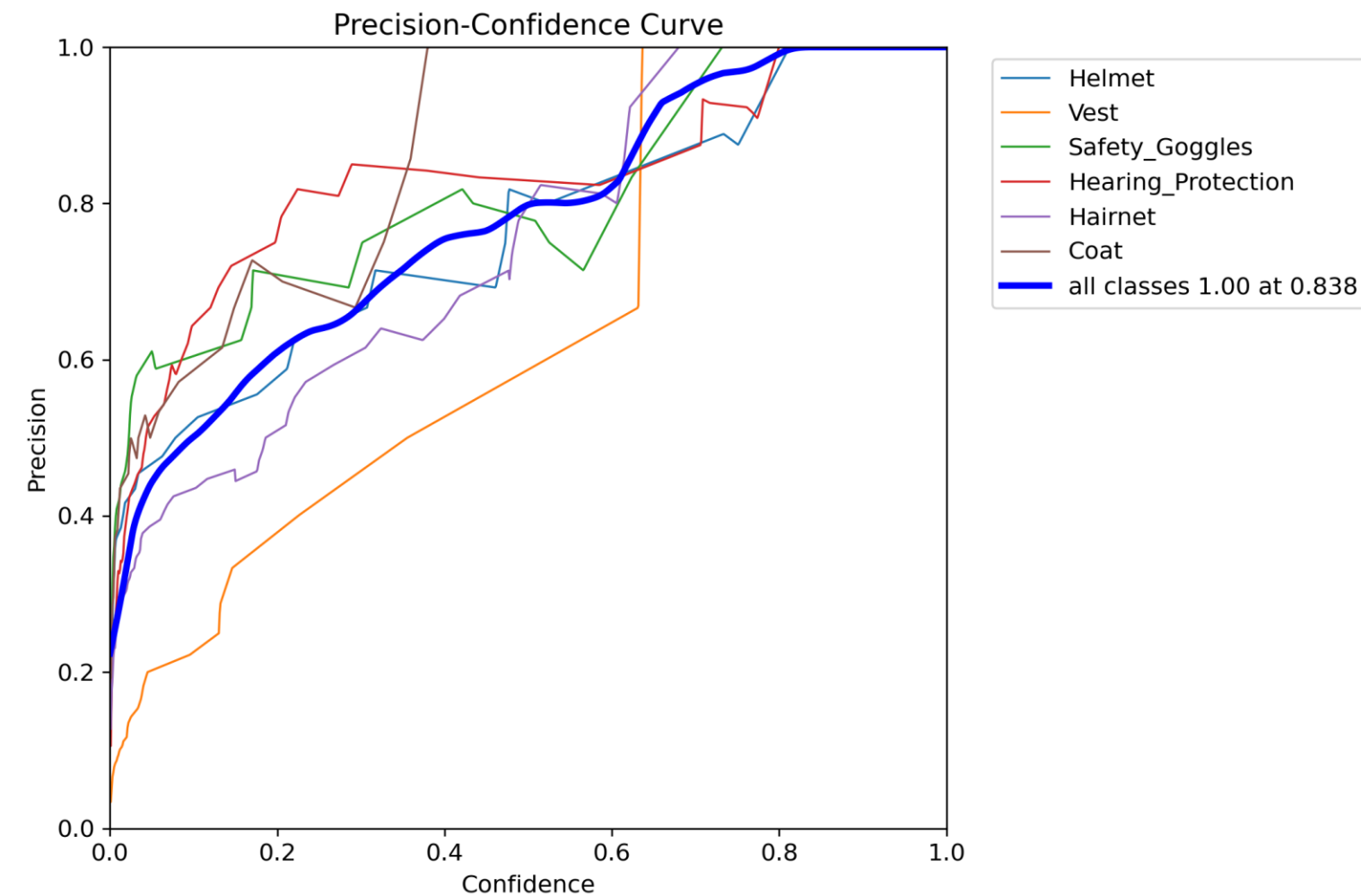
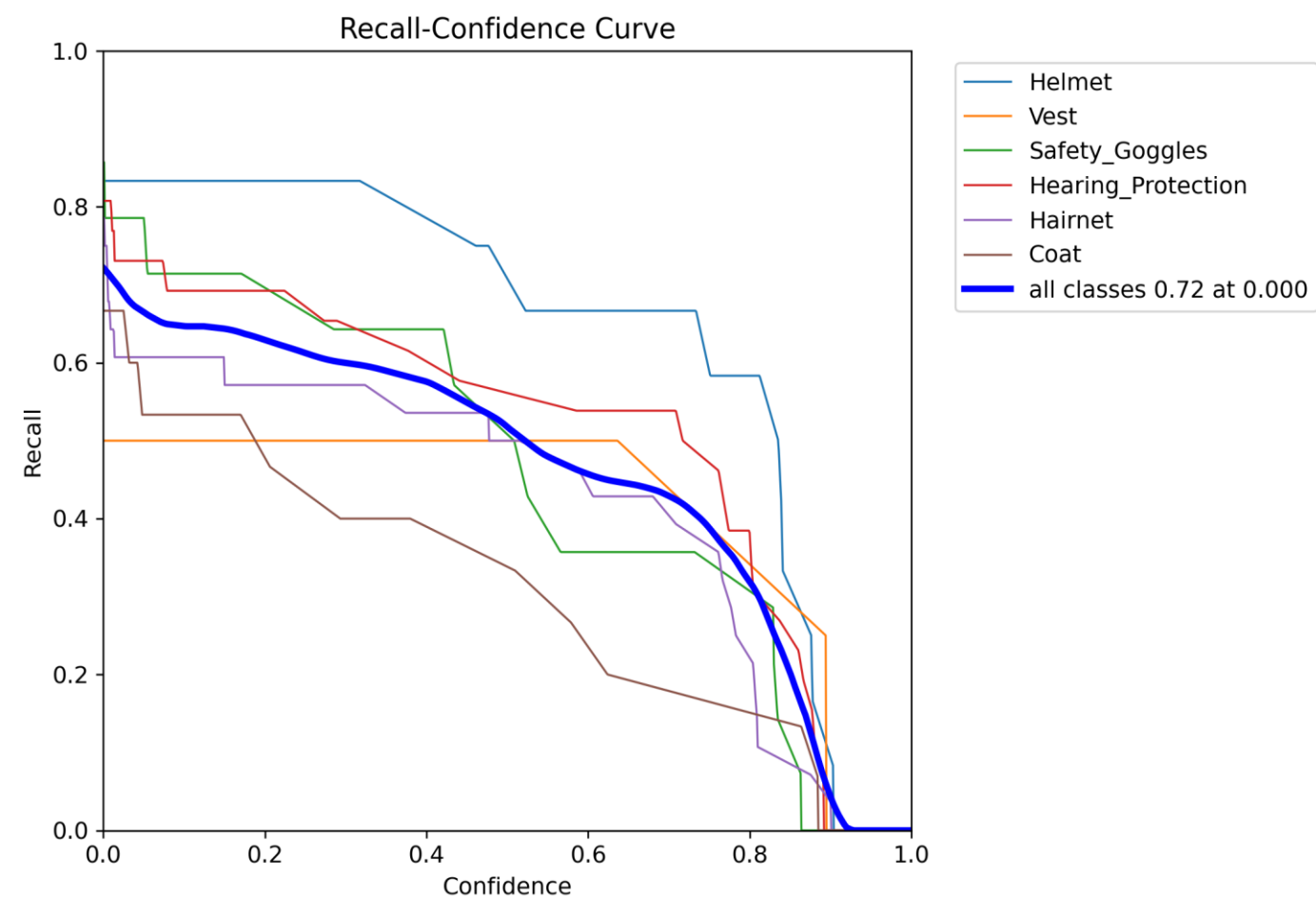




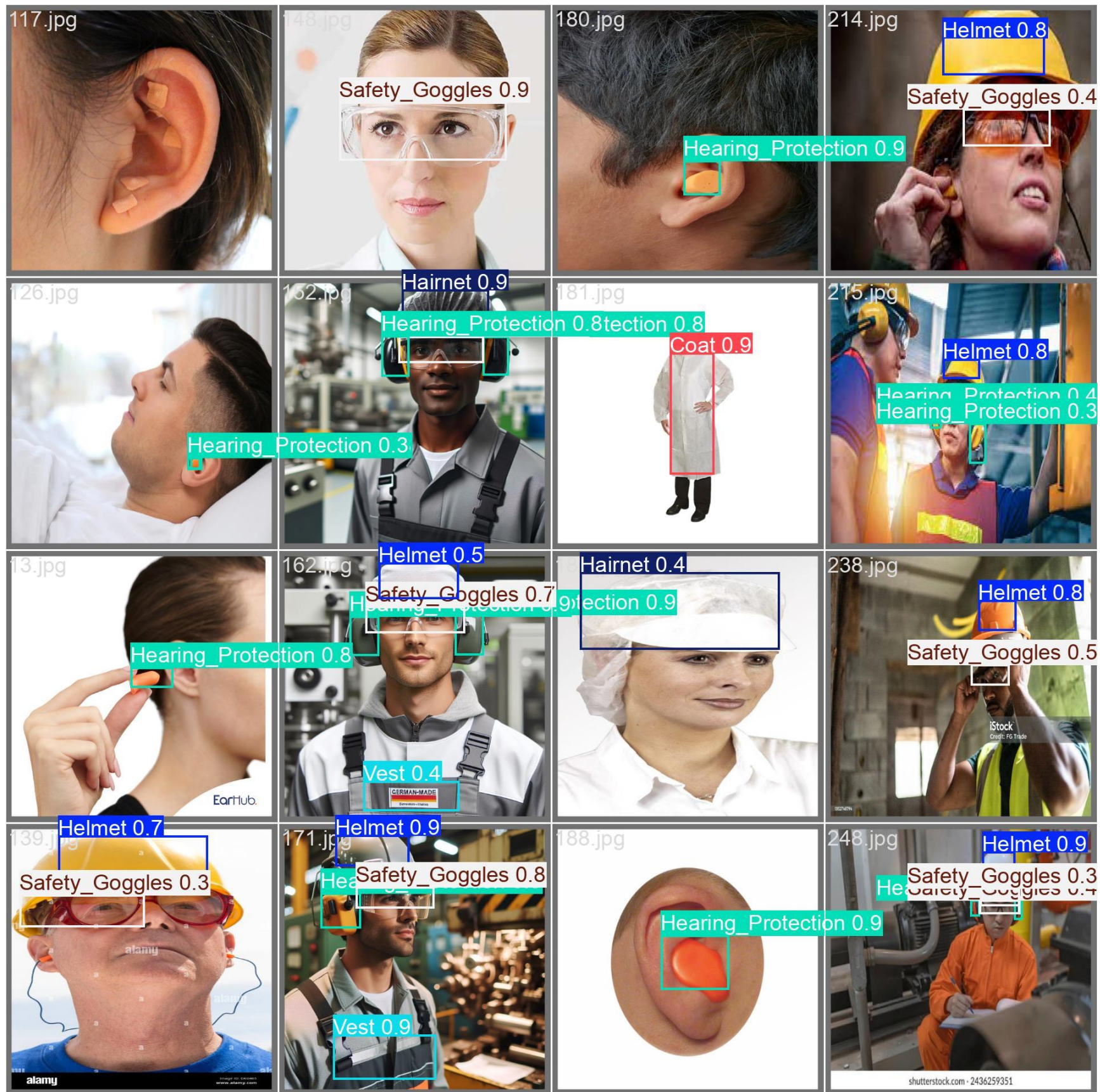
# Model Results

| Metric        | All Classes | Helmet | Vest  | Safety Goggles | Hearing Protection | Hairnet | Coat  |
|---------------|-------------|--------|-------|----------------|--------------------|---------|-------|
| Precision (P) | 0.738       | 0.706  | 0.511 | 0.791          | 0.842              | 0.625   | 0.955 |
| Recall (R)    | 0.583       | 0.801  | 0.500 | 0.643          | 0.617              | 0.536   | 0.400 |
| mAP@50        | 0.656       | 0.800  | 0.504 | 0.713          | 0.718              | 0.606   | 0.596 |







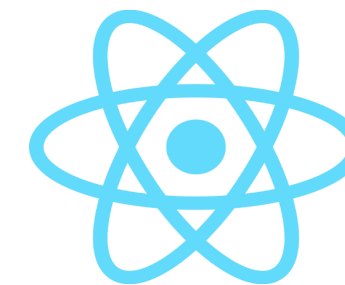




# Technology Stack

## Frontend

- Provides a concise and clear way for the user to enter the system, read notifications, check in real time the detection process & also get notified when there are anomalies detected.
- Build using ReactJS and Material UI



## Backend

- Built using Flask and is based on a microservice architecture to offer consistent and easier improvement in scalability and load balancing.
- It is also fed with information by a Kafka cluster from which the entry point of the app is subscribed to.



Flask

# Technology Stack

## Machine Learning

- YOLOv8 was used for PPE detection
- Trained on a custom dataset with nine classes.
- Evaluation was conducted using precision, recall, mAP scores to ensure reliable detection.



## Storage

- Different levels of storage are considered.
- A Redis cache for storing user sessions used to open sockets for notifications, MongoDB
- Databases for Users, Cameras and Notification, and an Elastic Search instance for inserting validation logs and connecting to Kibana for visualizations.



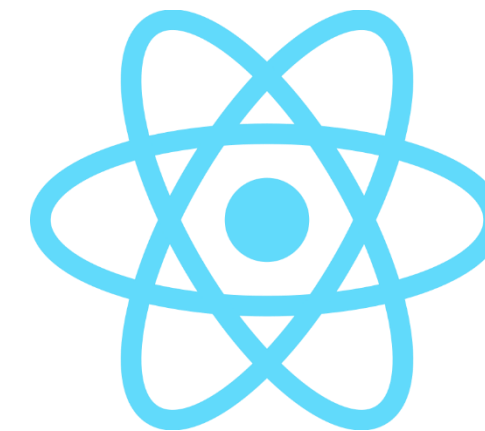
# TOOLS



LabelImg



mongoDB®



Microservices



redis



Flask



Material UI



# User Interface Overview

## Dashboard

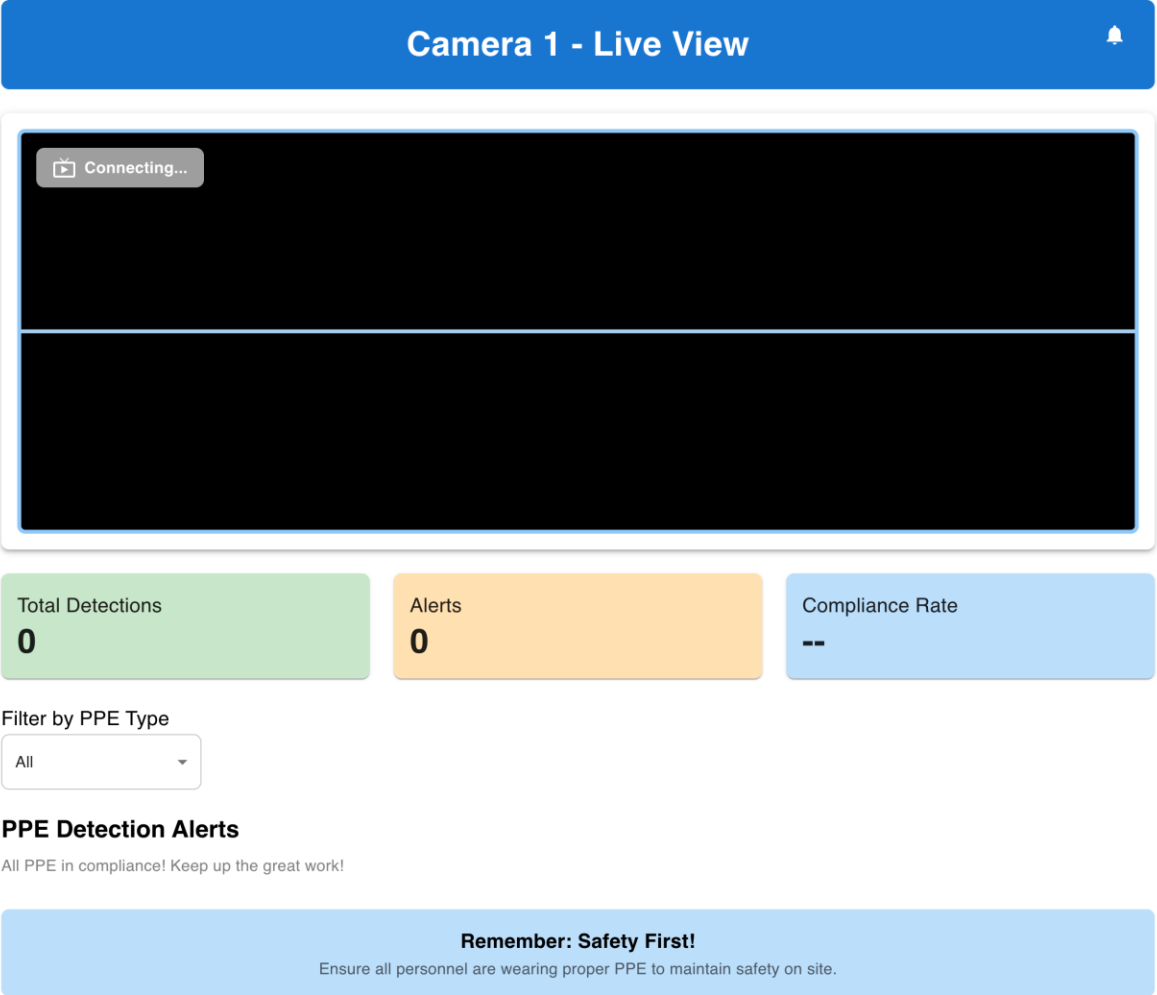
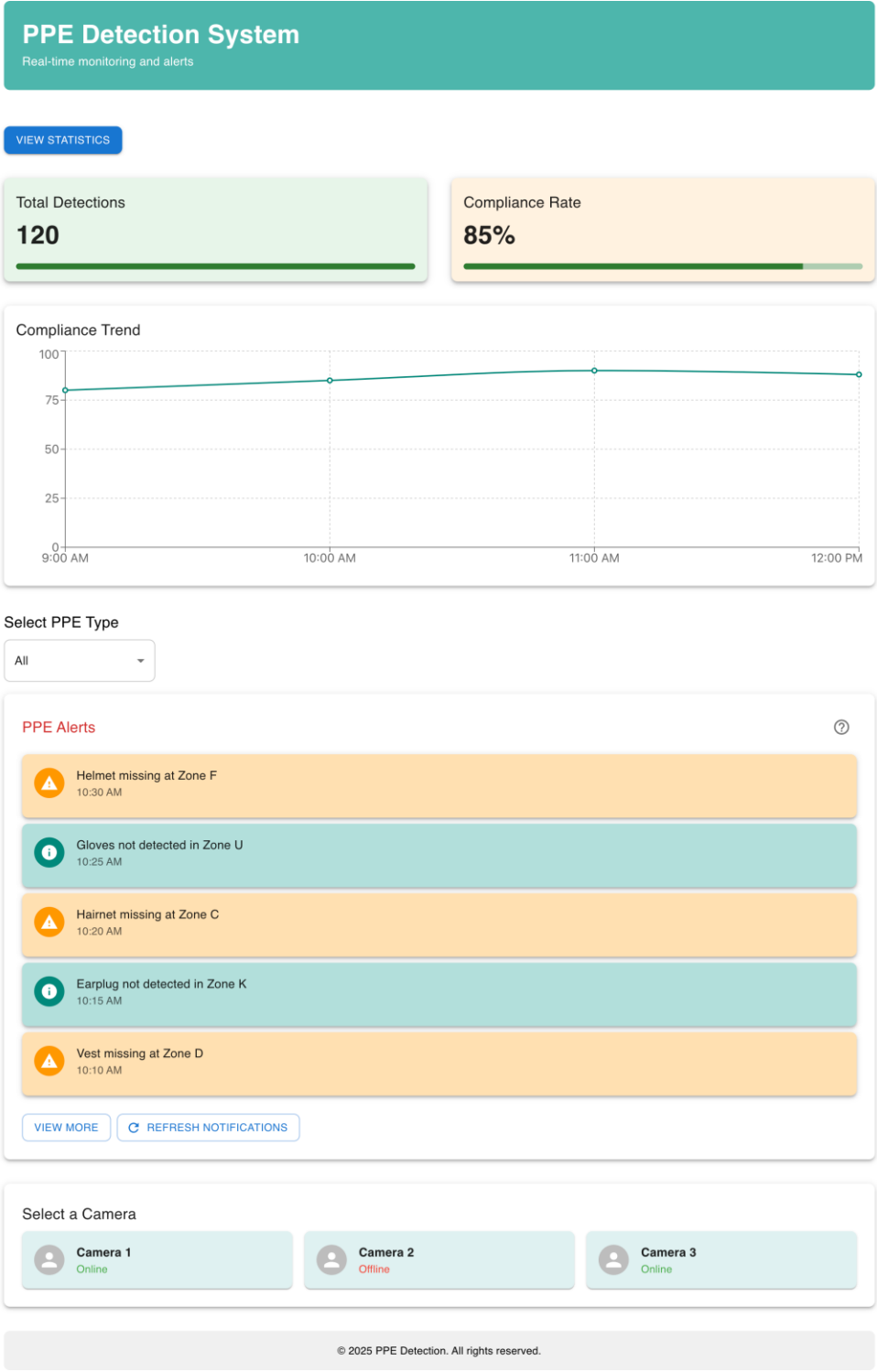
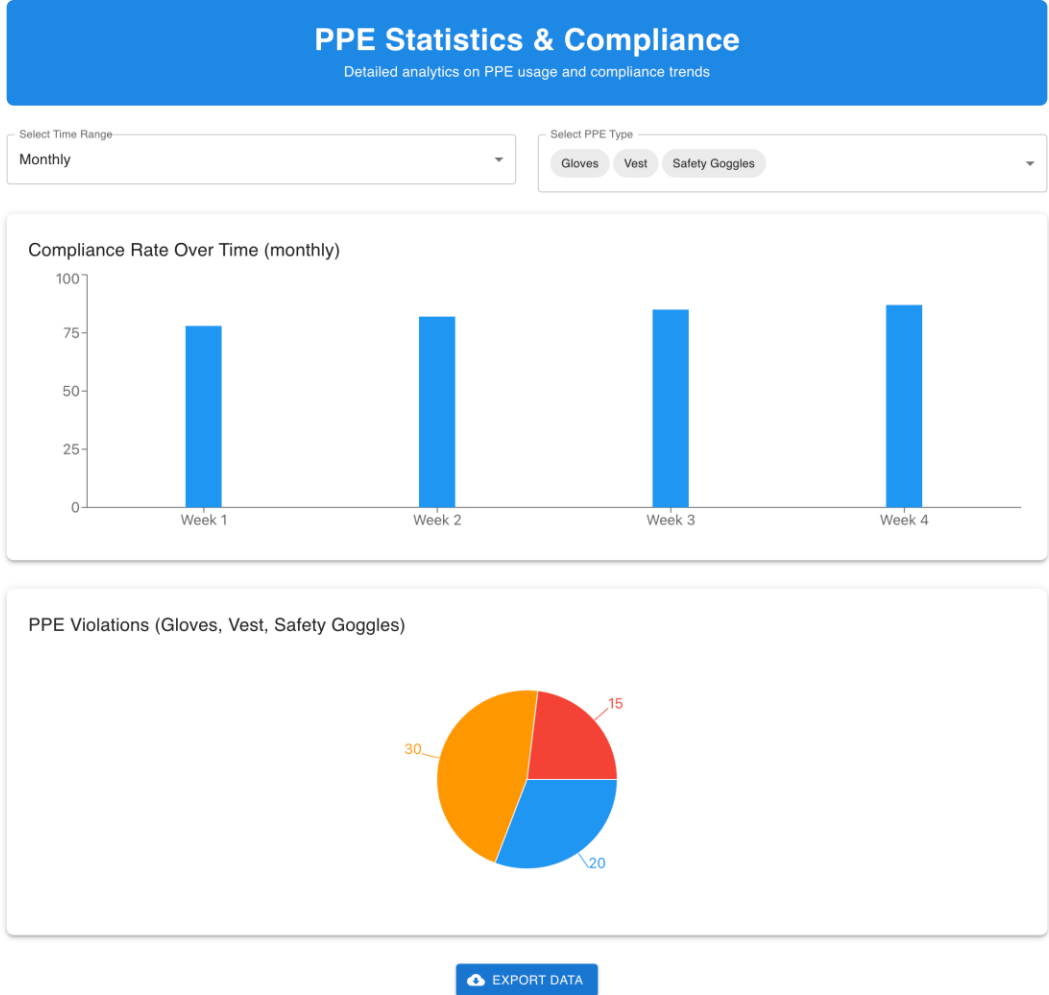
- Total Detections: Displays the number of detected PPE usages.
- Compliance Rate: Indicates the percentage of users adhering to PPE guidelines.
- PPE Alerts: Users can select specific alerts to view details about compliance issues.

## Camera View

- Live Stream: Show the current camera feed for specific cameras.
- Notification Alerts: Users receive real-time notifications specific to the selected camera.
- Key Metrics Displayed: Total Detections, Alerts, Compliance Rate

## Statistics

- Weekly, Monthly, Yearly Statistics: Visualize data trends over time.
- PPE Type Breakdown: Show compliance rates and alerts categorized by PPE type
- Download the data as a CSV



User Interface Design



# Impact

The solution strives to:

- Efficiently detecting existing personal protecting equipment
- Reporting correctly the missing equipment and notifying the user in real time
- **Offering analytics** for understanding better staff behavior in time.
- Allowing the user to monitor the video stream with the detected equipment.



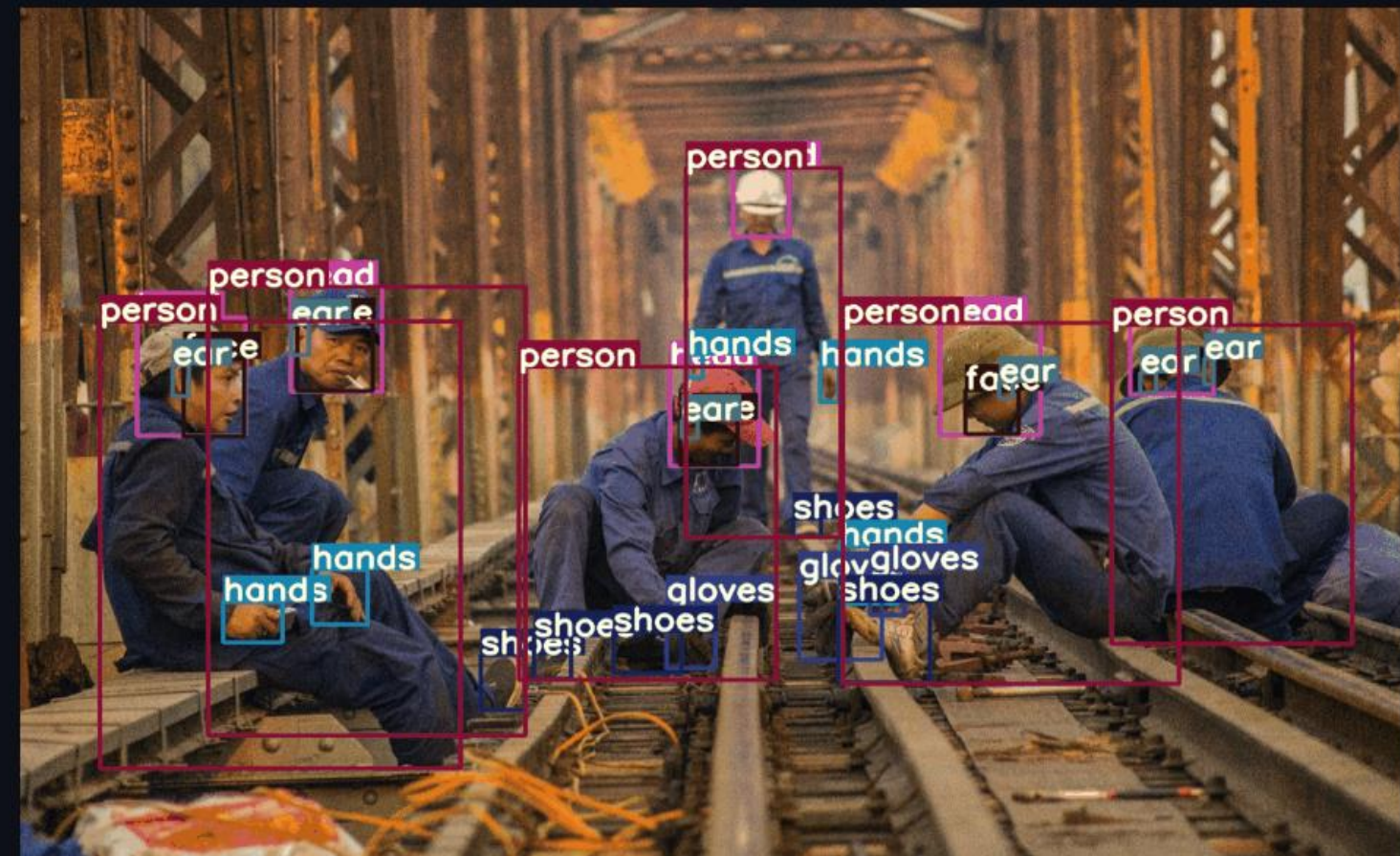


# Future work

- Experimenting with different AI models. ( **R-CNN**, **SSD** )
- Expanding the dataset for better accuracy
- Integrating with edge devices for real-time processing
- Experimenting with cloud or server by orchestrating through Kubernetes and implementing load Balancing

## SH17: A Dataset for Human Safety and Personal Protective Equipment Detection in Manufacturing Industry

arXiv Paper Accepted Paper JSSR KAGGLE





# Bigger Picture

***Ensure work safety with PPE  
compliance and protecting  
lives***



# THANK YOU

