

# Week4: Deployment on Flask

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## 1. Introduction

- **Purpose:** This report documents the deployment process of a machine learning model designed to predict airline safety improvements based on historical incident data.
- **Overview:** The project involves training a RandomForestClassifier on airline safety data and deploying it using a Flask web application to serve predictions.

## 2. Project Details

- **Name:** Aya SDOUR
- **Batch Code:** <https://github.com/ay0788/data/tree/main/flask>
- **Submission Date:** 2025-03-08
- **Submitted To:** Team Lead

## 3. Data Preparation

- **Data Loading:** The dataset `airline-safety.csv` is loaded using `pandas`. This dataset contains information about airline incidents and available seat kilometers.
- **Feature Engineering:**
  - Calculated `incident_rate_85_99` and `incident_rate_00_14` to normalize incident counts by available seat kilometers.
  - Created a binary label `safety_improved` to indicate whether the incident rate improved from 1985-1999 to 2000-2014.

## 4. Model Training

- **Feature Selection:** The features used for training are:
  - `incident_rate_85_99`: Incident rate from 1985 to 1999.
  - `fatal_accidents_85_99`: Number of fatal accidents from 1985 to 1999.

- `fatalities_85_99`: Number of fatalities from 1985 to 1999.
- `avail_seat_km_per_week`: Available seat kilometers per week.
- **Train-Test Split**: The data is split into 70% training and 30% testing sets using `train_test_split`.
- **Model Choice**: A `RandomForestClassifier` with 100 estimators is used for its robustness and ability to handle non-linear relationships.
- **Training Process**: The model is trained on the training set with a fixed random state for reproducibility.

## 5. Model Evaluation

- **Accuracy**: The model's accuracy on the test set is reported. For example, if the accuracy is 0.85, it means the model correctly predicts safety improvements 85% of the time.
- **Model Saving**: The trained model is saved to a file named `airline_safety_model.pkl` using `joblib` for later use in the Flask application.

## 6. Deployment

- **Flask Application**:
  - A Flask app is set up to serve the model. The app has two routes:
    - `/`: Renders the home page with a form to input data.
    - `/predict`: Accepts form data, makes a prediction using the model, and returns the result.
- **Prediction Logic**: The input data from the form is used to make a prediction. If the prediction is 1, it indicates that safety is likely to improve; otherwise, it is not.

## 7. Snapshots

- **Data Loading and Feature Engineering**: Screenshot of the dataset after loading and feature engineering steps.
- **Model Training and Evaluation**: Code snippet or output showing the training process and accuracy.
- **Flask App Setup**: Screenshot of the Flask app running and the home page.
- **Prediction Endpoint**: Example of a prediction result displayed on the web page.

## 8. Conclusion

- **Summary**: The deployment process involved training a `RandomForest` model on airline safety data and deploying it using Flask.

- **Challenges:** Any challenges faced during deployment, such as data preprocessing or model serving issues.
- **Next Steps:** Recommendations for future improvements, such as model optimization or adding more features.

## Airline Safety Prediction

Enter historical airline data to predict if safety will improve

Past Incident Rate (per billion km):

Past Fatal Accidents:

Past Fatalities:

Available Seat KM per Week:

Predict Safety Trend