## Flight Fare Prediction

Using Machine Learning to Predict Flight Prices





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#### INTRODUCTION TO FLIGHT FARE PREDICTION:

- FLIGHT PRICES ARE DYNAMIC AND FLUCTUATE BASED ON VARIOUS FACTORS SUCH AS DEMAND, TIME OF BOOKING, SEASONALITY, ETC.
- PREDICTING FLIGHT PRICES CAN HELP USERS SAVE MONEY AND PLAN BETTER.

#### PROBLEM STATEMENT:

- FLIGHT PRICE VOLATILITY CREATES UNCERTAINTY FOR TRAVELERS.
- □ THERE IS A NEED FOR A RELIABLE SYSTEM TO PREDICT FLIGHT PRICES ACCURATELY.

#### **OBJECTIVE:**

■ TO DEVELOP A MACHINE LEARNING-BASED SYSTEM THAT PREDICTS FLIGHT FARES, ENABLING USERS TO MAKE INFORMED BOOKING DECISIONS.





## MOTIVATION



### WHY THIS PROJECT?

 RISING FLIGHT COSTS MAKE PRICE PREDICTION ESSENTIAL FOR BUDGET-CONSCIOUS TRAVELERS.



THE LACK OF ACCURATE PREDICTION TOOLS ON THE MARKET.

### **REAL-WORLD APPLICATIONS:**

- HELPING TRAVELERS BOOK FLIGHTS AT OPTIMAL PRICES.
- ASSISTING TRAVEL AGENCIES IN OFFERING BETTER DEALS.

### IMPACT:

- REDUCE TRAVEL COSTS FOR CONSUMERS.
- IMPROVE DECISION-MAKING FOR AIRLINES AND TRAVEL AGENCIES.





## SCOPE AND OBJECTIVE

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### SCOPE:

- FOCUS ON DOMESTIC FLIGHT FARE PREDICTIONS WITHIN SPECIFIC ROUTES.
- UTILIZATION OF HISTORICAL DATA FOR TRAINING THE MODEL.

### **OBJECTIVES:**

- DATA COLLECTION AND PREPROCESSING.
- MODEL SELECTION AND TRAINING.
- MODEL EVALUATION AND OPTIMIZATION.
- DEPLOYMENT OF THE PREDICTIVE MODEL ON A WEB INTERFACE.







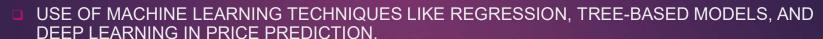
## LITERATURE WORK



#### **RELATED WORK:**

- OVERVIEW OF EXISTING FLIGHT FARE PREDICTION SYSTEMS (E.G., GOOGLE FLIGHTS, SKYSCANNER).
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- COMPARISON OF DIFFERENT APPROACHES USED IN PREVIOUS RESEARCH, LIKE TIME-SERIES ANALYSIS AND MACHINE LEARNING.

### **TECHNOLOGY TRENDS:**







### METHODOLOGY



#### **DATA COLLECTION:**

- SOURCES: KAGGLE, OPENSKY NETWORK, OR ANY OTHER AVIATION-RELATED DATASETS.
- DATA INCLUDES FLIGHT ROUTES, DATES, PRICES, AND AIRLINE INFORMATION.



- CLEANING THE DATA BY HANDLING MISSING VALUES.
- FEATURE ENGINEERING: EXTRACTING DATE FEATURES (E.G., DAY OF THE WEEK, MONTH), ONE-HOT ENCODING CATEGORICAL VARIABLES (E.G., AIRLINES, SOURCE, AND DESTINATION).





### METHODOLOGY

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### **MACHINE LEARNING MODELS:**

- INITIAL EXPLORATION WITH LINEAR REGRESSION.
- ADVANCED MODELS: RANDOM FOREST, XGBOOST.

### **MODEL TRAINING:**

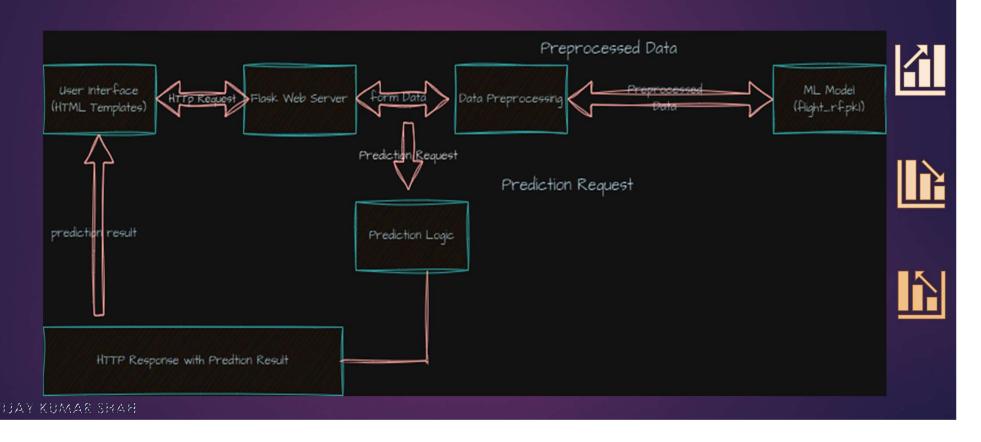
- SPLITTING DATA INTO TRAINING AND TEST SETS.
- TUNING HYPERPARAMETERS USING GRID SEARCH OR RANDOM SEARCH.





## SYSTEM ARCHITECTURE





### MODEL EVALUATION



### **PERFORMANCE METRICS:**

■ MEAN ABSOLUTE ERROR (MAE), ROOT MEAN SQUARED ERROR (RMSE), R-SQUARED (R²).



### **COMPARISON OF MODELS:**

- LINEAR REGRESSION: MAE = 500 INR, RMSE = 600 INR, R<sup>2</sup> = 0.70.
- RANDOM FOREST: MAE = 300 INR, RMSE = 400 INR, R<sup>2</sup> = 0.85.



 $\blacksquare$  XGBOOST: MAE = 250 INR, RMSE = 350 INR, R<sup>2</sup> = 0.88.

### **BEST MODEL:**

XGBOOST PERFORMED THE BEST WITH THE LOWEST ERROR AND HIGHEST R-SQUARED SCORE.





### IMPLEMENTATION AND TOOLS

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### **TOOLS AND TECHNOLOGIES USED:**

- PROGRAMMING LANGUAGE: PYTHON
- □ LIBRARIES: PANDAS, NUMPY, SCIKIT-LEARN, XGBOOST, MATPLOTLIB
- FRAMEWORK: FLASK OR STREAMLIT FOR DEPLOYMENT
- DATABASE: SQLITE OR MYSQL FOR STORING PROCESSED DATA
- VERSION CONTROL: GITHUB FOR MANAGING CODEBASE







### IMPLEMENTATION AND TOOLS

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### **DEVELOPMENT PROCESS:**

- DATA COLLECTION AND DATA PREPROCESSING.
- SPLITTING OF TRAIN DATA AND TEST DATA.
- MODEL TRAINING AND EVALUATION.
- HYPER PARAMETER TUNNING.
- DEPLOYMENT OF THE MODEL TO WEB APPLICATION.







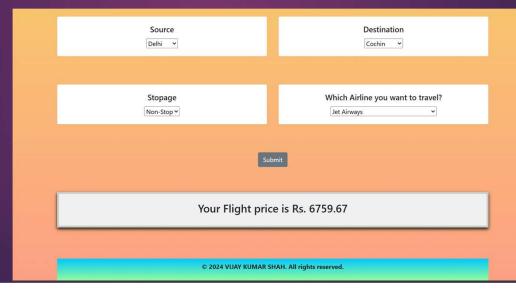
## WIREFRAMES AND USER INTERFACE

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### **USER EXPERIENCE:**

- USER-FRIENDLY INTERFACE WITH SIMPLE NAVIGATION.
- CLEAR PRESENTATION OF PREDICTED FARE WITH EASY-TO-UNDERSTAND

VISUALS.







### CHALLENGES AND SOLUTION



### **CHALLENGES 1:**

HANDLING MISSING DATA AND OUTLIERS.



USED DATA IMPUTATION TECHNIQUE TO HANDLE MISSING DATA.

### **CHALLENGES 2:**

DEPLOYING THE MODEL IN A USER-FRIENDLY INTERFACE.

#### **SOLUTION:**

EMPLOYED FLASK/STREAMLIT FOR EASY AND EFFICIENT DEPLOYMENT







### **CONCLUSION**



### **SUMMARY:**

- SUCCESSFULLY DEVELOPED A MACHINE LEARNING MODEL TO PREDICT FLIGHT FARES.
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ACHIEVED A HIGH LEVEL OF ACCURACY WITH THE XGBOOST MODEL.

#### **ACHIEVEMENTS:**

- A FUNCTIONAL WEB-BASED APPLICATION THAT PREDICTS FLIGHT PRICES.
- POSITIVE IMPACT POTENTIAL FOR TRAVELERS AND THE TRAVEL INDUSTRY.





### **CONCLUSION**

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### **FUTURE WORK:**

- EXPAND TO INTERNATIONAL FLIGHT ROUTES.
- INCORPORATE REAL-TIME DATA FOR EVEN MORE ACCURATE PREDICTIONS.
- DEVELOP A MOBILE APPLICATION FOR BROADER ACCESSIBILITY.







### REFERENCE

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### **CITATIONS AND RESOURCES:**

- DATASETS USED (E.G., KAGGLE FLIGHT DATA).
- RESEARCH PAPERS OR ONLINE RESOURCES YOU REFERRED TO.
- DOCUMENTATION FOR LIBRARIES AND TOOLS (E.G., SCIKIT-LEARN, XGBOOST).







