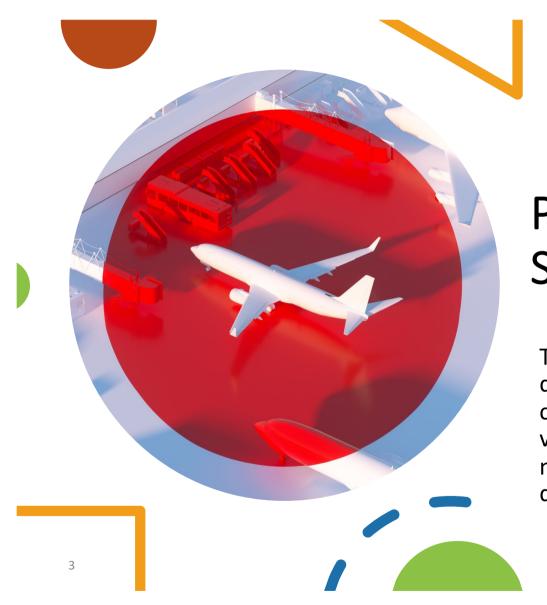




AGENDA

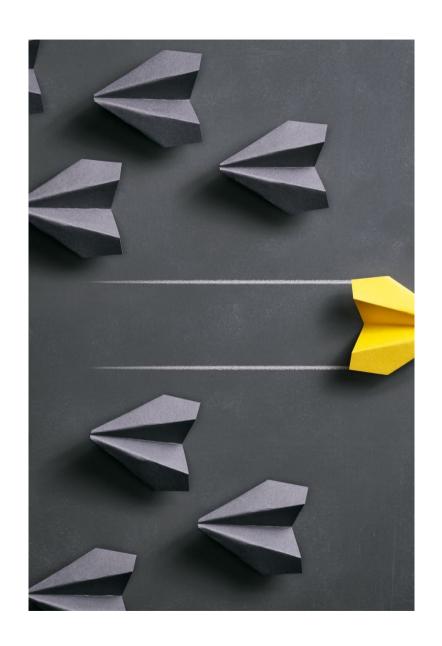
- Problem Statement
- Concept and Motivation
- Dataset Description
- Explore Data Analysis
- Methodology
- Results
- Conclusion

PRESENTATION TITLE



PROBLEM STATEMENT

To classify the Jan_2019 Ontime Flight dataset into Delayed or Not Delayed category(Flight Delayed or Not) by using various Machine Learning classification models and evaluating them using different evaluation metrics.



CONCEPT AND MOTIVATION

- Today, the aviation sector is vital to global transportation, and many companies depend on different airlines to connect them to other regions of the globe. However, due to various reasons there may be flight delays.
- To solve this issue, accurately classifying these flight delays allows passengers to be well prepared.
- The goal is to classify the dataset using sklearn and keras python module to build a classification model for flight delay.

DATASET DESCRIPTIO N



The Jan_2019 Ontime Flight Dataset from Kaggle was used for this project.



Based on several input features, this dataset is used to decide if a flight is delayed or not.



Relevant details regarding the flight are provided in each row of the data.



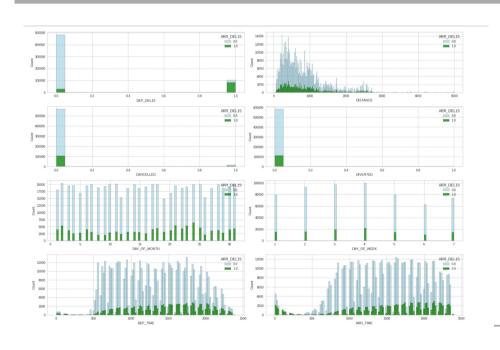
DATASET DESCRIPTION

- The dataset contains 583985 rows with 22 features set.
- Features are:
- Description about some important features are below:
 - DAY_OF_MONTH: indicates day of the month
 - DAY_OF_WEEK: indicates day of the week
 - ORIGIN AIRPORT ID: indicates unique origin airport id
 - DEST_AIRPORT_ID: indicates unique destination id
 - DEP_DEL15: indicates there is departure delay or not(1=yes, 0=no)
 - CANCELLED: indicates whether flight is cancelled or not(1=yes, 0=no)
 - **DIVERTED:** indicates whether flight is diverted or not(1=yes, 0=no)
 - **DISTANCE**: indicates the distance between airports in miles
 - ARR_DEL15: indicates if the flight is delayed or not (1 = yes, 0 = no)(Target feature)

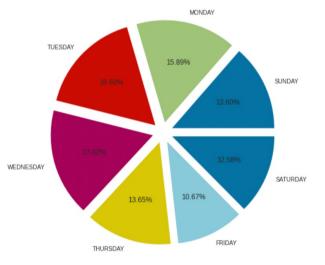
EXPLORE DATA ANALYSIS

,				Heat	map displa	aying the	relationshi	ip between	nthe featu	ires of the	data			
DAY_OF_MONTH -		-0.031	-0.012	-0.005	-0.005	-0.005	-0.005	-0.0014	0.03	0.0035	0.041	0.11	0.0014	-0.02
DAY_OF_WEEK -	-0.031	1	0.06	0.0059	0.0059	0.0057	0.0057	0.004	-0.0035	0.0057	-0.00097	-0.0078	0.00061	0.019
OP_CARRIER_FL_NUM -	-0.012	0.06	1	-0.057	-0.057	-0.061	-0.061	0.0098	0.027	0.0056	0.04	0.047	0.0089	-0.34
ORIGIN_AIRPORT_ID -	-0.005	0.0059	-0.057	1	1	0.018	0.018	-0.036	0.004	-0.0059	0.0085	0.013	-0.00071	0.092
ORIGIN_AIRPORT_SEQ_ID -	-0.005	0.0059	-0.057	1	1	0.018	0.018	-0.036	0.004	-0.0059	0.0085	0.013	-0.00071	0.092
DEST_AIRPORT_ID -	-0.005	0.0057	-0.061	0.018	0.018			0.028	0.012	0.021	0.015	0.012	0.0055	0.092
DEST_AIRPORT_SEQ_ID -	-0.005	0.0057	-0.061	0.018	0.018			0.028	0.012	0.021	0.015	0.012	0.0055	0.092
DEP_TIME -	-0.0014	0.004	0.0098	-0.036	-0.036	0.028	0.028	1		0.67	0.14	-0.00043	0.0031	-0.027
DEP_DEL15 -	0.03	-0.0035	0.027	0.004	0.004	0.012	0.012	0.19	1	0.088	0.72	0.038	0.02	0.0093
ARR_TIME -	0.0035	0.0057	0.0056	-0.0059	-0.0059	0.021	0.021	0.67	0.088	1	0.062	-0.0042	-0.0033	0.021
ARR_DEL15 -	0.041	-0.00097	0.04	0.0085	0.0085	0.015	0.015	0.14	0.72	0.062	1	0.039	0.014	0.00051
CANCELLED -	0.11	-0.0078	0.047	0.013	0.013	0.012	0.012	-0.00043	0.038	-0.0042	0.039	1	-0.0081	-0.049
DIVERTED -	0.0014	0.00061	0.0089	-0.00071	-0.00071	0.0055	0.0055	0.0031	0.02	-0.0033	0.014	-0.0081	1	0.013
DISTANCE -	-0.02	0.019	-0.34	0.092	0.092	0.092	0.092	-0.027	0.0093	0.021	0.00051	-0.049	0.013	
	DAY_OF_MONTH -	DAY_OF_WEEK -	OP_CARRIER_FL_NUM -	ORIGIN_AIRPORT_ID -	ORIGIN_AIRPORT_SEQ_ID -	DEST_AIRPORT_ID -	DEST_AIRPORT_SEQ_ID -	DEP_TIME -	DEP_DEL15 -	ARR_TIME -	ARR_DEL15 -	CANCELLED -	DIVERTED -	DISTANCE -

EXPLORE DATA ANALYSIS



Cancelled Flights Percent Based On Days Of Week



METHODOLOGY

- Initially I performed Data Cleaning.
- Then, I did some exploratory analysis on the data, like performing Uni and Bi- variate analysis on features and also explored more about individual features that affecting the target feature.
- I trained the data set using 8 different classification models and also did tunning hyper parameters in order obtain best model. Below are the models that are used:
 - K-Nearest Neighbor which classifies or predicts how an individual data point will be grouped using proximity.
 - Naive Bayes simple probabilistic classifier
 - Decision Trees Classifier simple decision rules inferred from the data features
 - Logistic Regression uses the sigmoid function to return the probability of a label.
 - **GBDT** a meta-estimator that averages the results of several decision tree classifiers fitted to different dataset sub-samples to increase the predicted accuracy.
 - Random Forest uses bagging and feature randomness.
 - **Neural Networks** A NN Sequential model is appropriate for a plain stack of layers where each layer has exactly one input tensor and one output tensor.
 - **SVM** used for classification, outlier detection and regression



Model	F1-Score	Cross_Val_Score (ROC_AUC)				
KNN	0.8679	0.784				
Naïve Bayes	0.8977	0.851				
Decision Trees	0.9163	0.879				
Logistic Regression	0.9158	0.858				
GBDT	0.9194	0.926				
Random Forest	0.9227	0.937				
Neural Network	0.9158	-				
SVM	0.9173	0.890				

PRESENTATION TITLE 10

CONCLUSION

Successfully able to classify the dataset into delayed and not delayed and also, we can observe that Gradient Boosting Decision Trees, Random Forest and Neural Networks were able to classify the dataset with highest accuracy and cross validation score.

Moreover, with the help of EDA I was able to select the relevant features for the classification which provided me utmost accuracy



REFERENCES

- https://www.kaggle.com/code/godswayjd/flight-delay/data
- https://www.geeksforgeeks.org/
- https://towardsdatascience.com/