



Data Collection and Preprocessing Phase

Date	15 July 2024
Team ID	739725
Project Title	Flight Delay Prediction using Machine Learning.
Maximum Marks	6 Marks

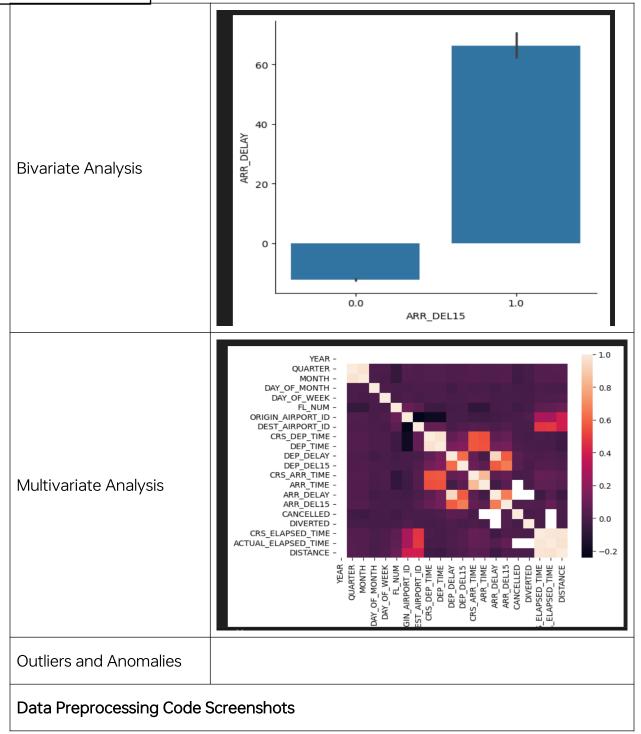
Data Exploration and Preprocessing Template

Dataset variables will be statistically analyzed to identify patterns and outliers, with Python employed for preprocessing tasks like normalization and feature engineering. Data cleaning will address missing values and outliers, ensuring quality for subsequent analysis and modeling, and forming a strong foundation for insights and predictions.

Section	Descr	iptior	1							
	YEA	R QUARTER	MONTH	DAY_OF_MONTH	DAY_OF_WEE	C FL_NUM	ORIGIN_AIRPORT_	ID DEST_AIRPORT_ID	CRS_DEP_TIME	DEP_TIME
	count 11231.	0 11231.000000	11231.000000	11231.000000	11231.00000	11231.000000	11231.0000	00 11231.000000	11231.000000	11124.000000
	mean 2016.	0 2.544475	6.628973	15.790758	3.960199	1334.325617	12334.5166	95 12302.274508	1320.798326	1327.189410
	std 0.	0 1.090701	3.354678	8.782056	1.99525	7 811.875227	1595.0265	10 1601.988550	490.737845	500.306462
	min 2016.	0 1.000000	1.000000	1.000000	1.000000	7.000000	10397.0000	00 10397.000000	10.000000	1.000000
	25% 2016	0 2.000000	4.000000	8.000000	2.00000	624.000000	10397.0000	00 10397.000000	905.000000	905.000000
	50% 2016.	0 3.000000	7.000000	16.000000	4.000000	1267.000000	12478.0000	00 12478.000000	1320.000000	1324.000000
	75% 2016.	0 3.000000	9.000000	23.000000	6.00000	2032.000000	13487.0000	00 13487.000000	1735.000000	1739.000000
	max 2016.	0 4.000000	12.000000	31.000000	7.00000	2853.000000	14747.0000	00 14747.000000	2359.000000	2400.000000
Data Overview	8 rows × 21 colu	imns								
Data Overview	DEP DEI 15	CRS ARR TIME	ARR TIME	ARR DELAY	ARR DEL15	CANCELLED	DIVERTED CRS FI	APSED TIME ACTUA	AL FLAPSED TIME	DISTANCE
	11124.000000	11231.000000			11045.000000 1			11231.000000	11043.000000	11231.000000
	0.142844	1537.312795	1523.978499	-2.573123	0.124672	0.010150	0.006589	190.652124	179.661233	1161.031965
	0.349930	502.512494	512.536041	39.232521	0.330361	0.100241	0.080908	78.386317	77.940399	643.683379
	0.000000	2.000000	1.000000	-67.000000	0.000000	0.000000	0.000000	93.000000	75.000000	509.000000
	0.000000	1130.000000	1135.000000	-19.000000	0.000000	0.000000	0.000000	127.000000	117.000000	594.000000
	0.000000	1559.000000	1547.000000	-10.000000	0.000000	0.000000	0.000000	159.000000	149.000000	907.000000
	0.000000	1952.000000	1945.000000	1.000000	0.000000	0.000000	0.000000	255.000000	236.000000	1927.000000
	1.000000	2359.000000	2400.000000	615.000000	1.000000	1.000000	1.000000	397.000000	428.000000	2422.000000
Univariate Analysis	f1 pl	ights['t.show(YEAR'].	b.pyplo			ind='pie	',autopct=	·'%. Of')	











	flights-pd.read_csv("flightdata.csv") flights								
		P_DEL15							
	0 2016 1 1 1 5 DL N836DN 1399 10397 ATL 1 2016 1 1 1 5 DL N964DN 1476 11433 DTW	0.0							
	2 2016 1 1 1 5 DL N813DN 1597 10397 ATL	0.0							
	3 2016 1 1 1 5 DL NS87NW 1768 14747 SEA 4 2016 1 1 1 5 DL N836DN 1823 14747 SEA	0.0							
	4 2010 1 1 1 3 DL ROSOLIN 1023 14/41 SEA								
	11226 2016 4 12 30 5 DL N940DL 1715 11433 DTW	0.0							
	11227 2016 4 12 30 5 DL N836DN 1770 14747 SEA 11228 2016 4 12 30 5 DL N583NW 1823 11433 DTW	0.0							
	11229 2016 4 12 30 5 DL N554NW 1901 10397 ATL	0.0							
oading Data	11220 2016 4 12 20 5 DI MB/JOM 2005 10207 ATI	0.0							
oading Data	_								
		STANCE							
	2143 2102.0 -41.0 0.0 0 0 338 295.0 1435 1439.0 4.0 0.0 0 0 110 115.0	2182 528							
	1215 1142.0 -33.0 0.0 0 0 335 300.0	2182							
	1335 1345.0 10.0 0.0 0 0 196 205.0	1399							
	607 615.0 8.0 0.0 0 0 247 259.0	1927							
		594							
	2046 2100.0 14.0 0.0 0 0 201 181.0	1399							
	2210 2154.0 -16.0 0.0 0 0 311 295.0	1927							
	1806 1801.0 -5.0 0.0 0 0 336 332.0 925 913.0 -12.0 0.0 0 0 120 110.0	2182 594							
	925 915.0 -12.0 0.0 0 0 120 110.0	594							
	flights=flights.fillna({'dep_dell5':e}) flights.iloc[177:185] / 00s FL NUM MONTH DAY OF MONTH DAY OF WEEK ORIGIN DEST CRS ARR TIME DEP DEL15 ARR DEL15 177 2834 1 9 6 MSP SEA 852 0.0 1.0								
	177 2834 1 9 6 MSP SEA 852 0.0 1.0 178 2839 1 9 6 DTW JFK 1724 0.0 0.0								
	179 86 1 10 7 MSP DTW 1632 NaN 1.0								
	180 87 1 10 7 DTW MSP 1649 1.0 0.0								
	181 423 1 10 7 JFK ATL 1600 0.0 0.0 182 440 1 10 7 JFK ATL 849 0.0 0.0								
	183 485 1 10 7 JFK SEA 1945 1.0 0.0								
	184 557 1 10 7 MSP DTW 912 0.0 1.0								
Handling Missing Data	<pre>import math for index,row in flights.iterrows(): [flights.loc[index,'CR5_ARR_TIME']=math.floor(row['CR5_ARR_TIME']/188) flights.head()</pre>								
	FL NUM MONTH DAY OF MONTH DAY OF WEEK ORIGIN DEST CRS ARR TIME DEP DEL15 AF	RR DEL15							
	0 1399 1 1 5 ATL SEA 21 0.0	0.0							
	1 1476 1 1 5 DTW MSP 14 0.0	0.0							
	2 1597 1 1 5 ATL SEA 12 0.0	0.0							
	3 1768 1 1 5 SEA MSP 13 0.0	0.0							
	4 1823 1 1 5 SEA DTW 6 0.0	0.0							
	<pre>from sklearn.preprocessing import OneHotEncoder oh=OneHotEncoder()</pre>)							
Data Transformation	<pre>z=oh.fit_transform(flights.iloc[:,4:5]).toarray() t=oh.fit_transform(flights.iloc[:,5:6]).toarray() z // 0.0s array([[1., 0., 0., 0., 0.],</pre>								
ata Transformation	<pre>z=oh.fit_transform(flights.iloc[:,4:5]).toarray() t=oh.fit_transform(flights.iloc[:,5:6]).toarray() z / 0.0s array([[1., 0., 0., 0., 0.],</pre>								
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Feature Engineering	Attached the codes in the final submission.
Save Processed Data	-