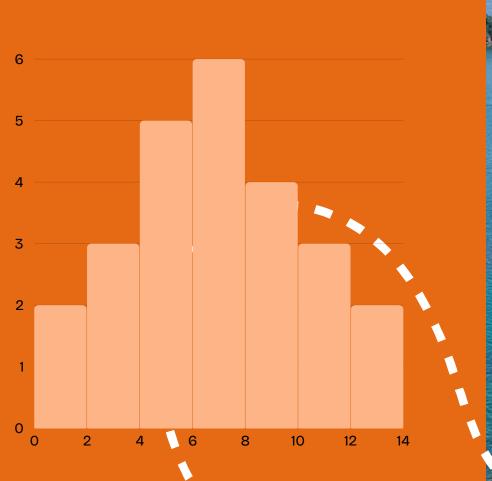
AIRLINES CUSTOMER SATISFACTION ANALYSIS

CODE³



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PROBLEM STATEMENT

- The airline industry, hit hard by the pandemic and recession, with ICAO estimating a \$371 billion loss in 2020, must now focus on improving passenger satisfaction to recover
- This project aims to build a binary classification model
 that predicts whether a passenger is Satisfied or
 Neutral/Dissatisfied, based on flight experience, service
 quality, and customer profile





-HOW WE DO AND WHAT WE DO

- The model considers passenger demographics such as **gender**, **age**, **and customer type** to identify **satisfaction trends**.
- Travel details including type of travel, travel class, and flight distance are used to capture the impact of journey characteristics on satisfaction.
- Departure and arrival delays are important features that reflect the effect of flight disruptions on passenger experience.
- Ratings for booking ease, online check-in, and gate location are included to evaluate pre-boarding service quality.
- Inflight services such as Wi-Fi availability, entertainment, food and drink quality, and seat comfort help measure mid-flight satisfaction.
- On-board service, leg room space, and cleanliness represent the physical comfort and crew interaction during the flight.
- Ground services like **check-in efficiency and baggage handling** complete the assessment of the over<mark>all passenger</mark> journey.
- The target variable is passenger satisfaction, classified as either satisfied or neutral/dissatisfied based on the overall experience





SET

- The dataset is from Kaggle. it provides cutting-edge data science, faster and better than most people ever thought possible. Kaggle offers both public and private data science competitions and on-demand consulting by an elite global talent pool.
- Dataset link
 https://www.kaggle.com/datasets/teejma
 hal20/airline-passenger-satisfaction

Data	columns (total 23 columns):			
#	Column	Non-Nu	ll Count	Dtype
0	Gender	103904	non-null	object
1	Customer Type	103904	non-null	object
2	Age	103904	non-null	int64
3	Type of Travel	103904	non-null	object
4	Class	103904	non-null	object
5	Flight Distance	103904	non-null	int64
6	Inflight wifi service	103904	non-null	int64
7	Departure/Arrival time convenient	103904	non-null	int64
8	Ease of Online booking	103904	non-null	int64
9	Gate location	103904	non-null	int64
10	Food and drink	103904	non-null	int64
11	Online boarding	103904	non-null	int64
12	Seat comfort	103904	non-null	int64
13	Inflight entertainment	103904	non-null	int64
14	On-board service	103904	non-null	int64
15	Leg room service	103904	non-null	int64
16	Baggage handling	103904	non-null	int64
17	Checkin service	103904	non-null	int64
18	Inflight service	103904	non-null	int64
19	Cleanliness	103904	non-null	int64
20	Departure Delay in Minutes	103904	non-null	int64
21	Arrival Delay in Minutes	103594	non-null	float64
22	satisfaction	103904	non-null	object
dtypes: float64(1), int64(17), object(5)				
memory usage: 18.2+ MB				



We used EDA to understand the data, detect patterns, and identify important features.

- Analyzed summary statistics to check distributions and outliers.
- Used visualizations like histograms,
 boxplots, and heatmaps to explore
 relationships between age, travel class,
 delays, and satisfaction.
- Identified key trends showing how service quality and customer profiles affect satisfaction.





HEATMAP



Departure delay in minutes and arrival delay in minutes are highly co-related!

