



# Predicting Customer Satisfaction for Airlines

Final Presentation | Group A | June 17<sup>th</sup>, 2024

# Approach

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## Problem Statement

- Understanding of **issue** in practical context
- Definition of **business goal** for the implementation of ML algorithms

## Dataset Introduction

- Understanding of **underlying numerical and categorical predictors**

## ML Model Selection

- **ML modelling approach**
- **Choice of ML models** for solving the problem

## Model Evaluation & Feature Importance

- Interpretation of **performance metrics** of the model
- Extraction of **feature importance** to determine most relevant predictors

## Impact

- **Suggestion of key measures** to implement going forward
- **Application of ML models** withing a **dashboard**

➤ **Approach**

➤ **Problem Statement**

➤ **Dataset Introduction**

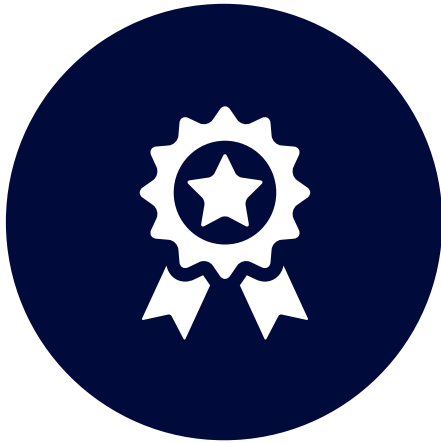
➤ **ML Model Selection**

➤ **Model Evaluation & Feature Importance**

➤ **Impact**

# Dissatisfied customers have detrimental consequences

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## Negative Impact on Reputation

Bad reviews, damaged brand image



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## Decreased Customer Loyalty

Loss of repeat business, higher customer acquisition costs



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## Negative Financial Consequences

Refunds, compensation, costly operational adjustments

➤ **Approach**

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➤ **Impact**

# Our dataset is sourced from an undisclosed airline

## Data Source<sup>1</sup>



**Undisclosed  
airline company**

## Details



### **Customer survey**

Scores from customer survey questions



### **Passenger details**

e.g., age, booked class, travel type



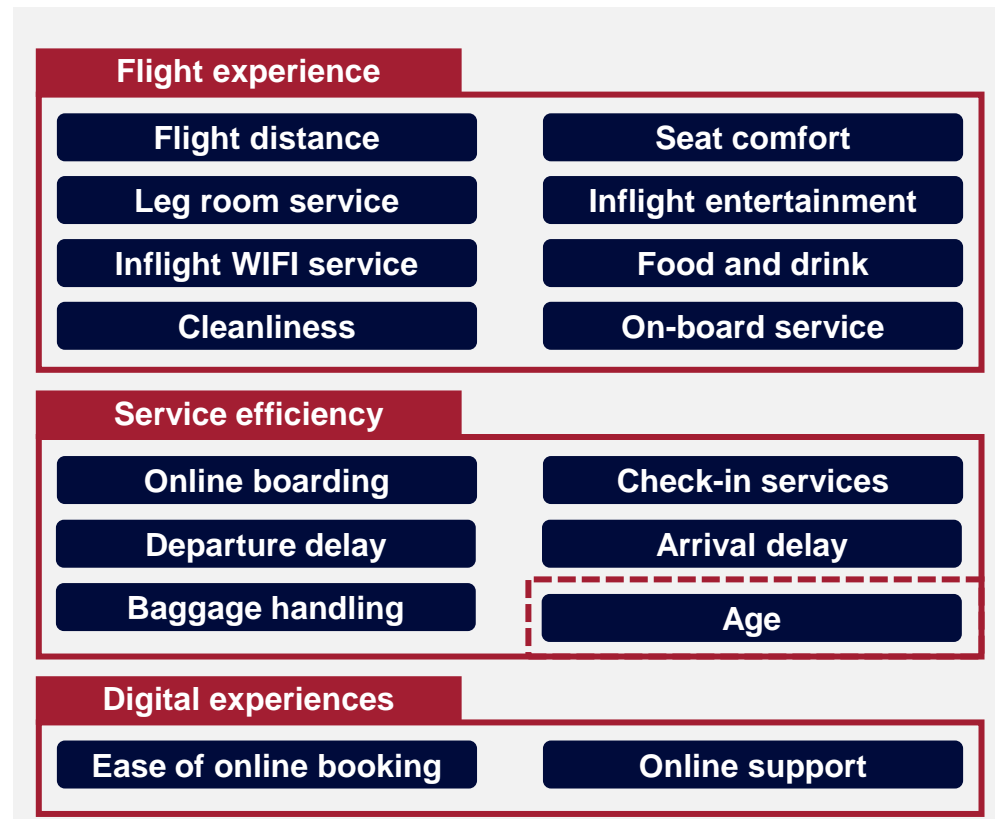
### **Flight information**

Departure and arrival delay

<sup>1</sup> Source: <https://www.kaggle.com/datasets/raminhuseyn/airline-customer-satisfaction>

# In our models, we rely on 19 flight-related predictors

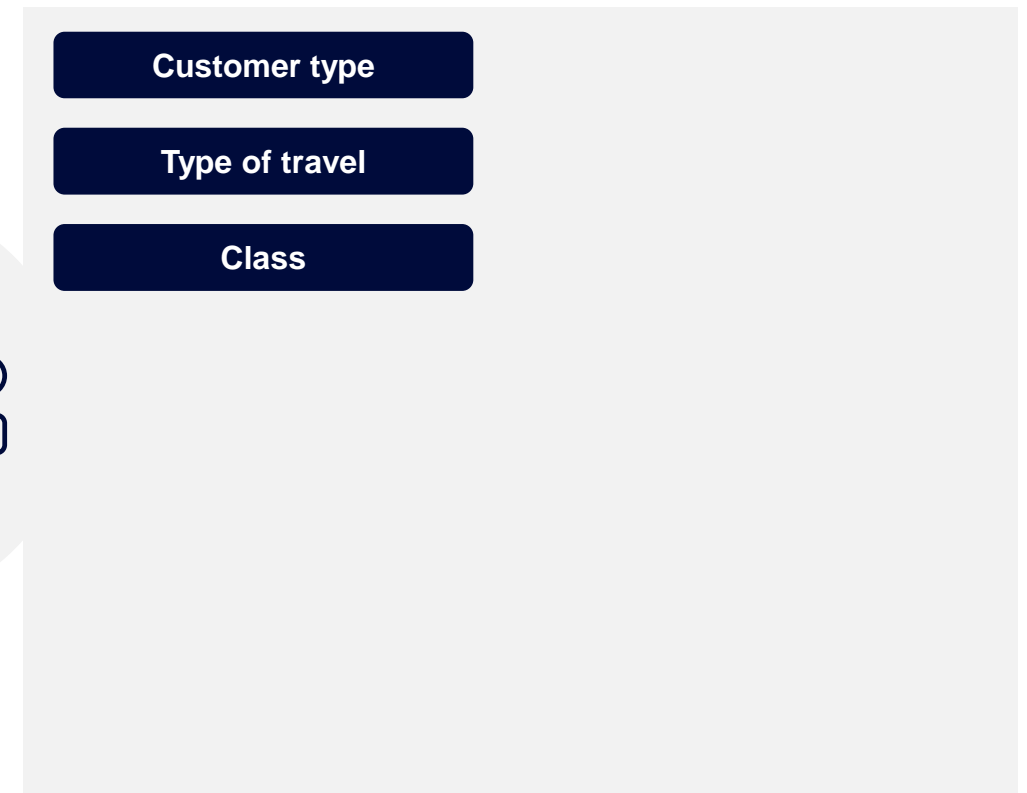
## Numerical Predictors



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## Categorical Predictors



➤ **Approach**

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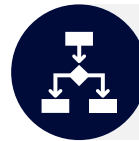
➤ **Impact**



# We decided to compare 6 different ML model algorithms

## Binary classification

Will a customer be satisfied or dissatisfied after the flight?



Decision Tree



Support Vector Classifier



Logistic Regression



NN (MLP Classifier)



XGBoost



AutoML (TPOT Classifier)

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- **Impact**

# XGBoost performs best with a Validation AUC of 0.992

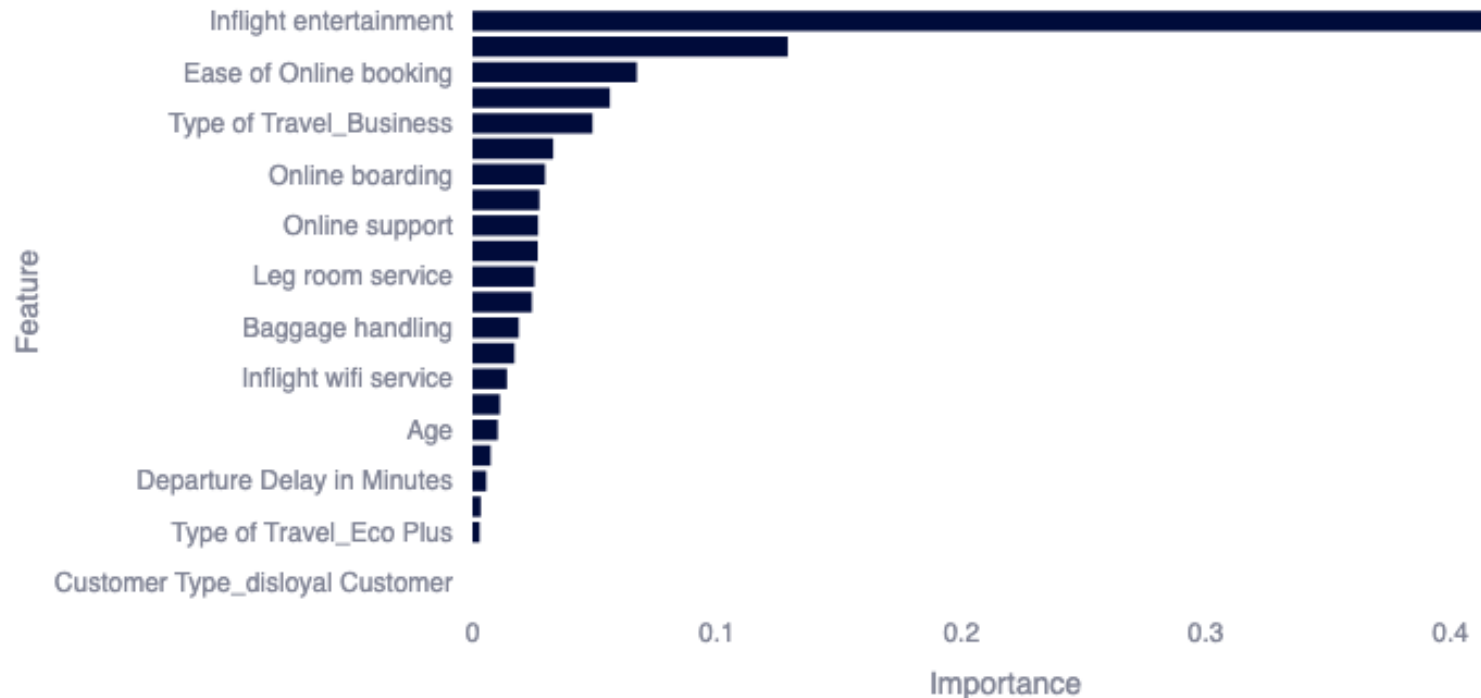
## Criterion: Validation AUC (Area under curve)

- 1 AUC is **independent of setting a classification threshold**, which makes it robust even when the threshold is not known
- 2 AUC **considers both the True Positive Rate (Sensitivity) and the False Positive Rate (Specificity)**, ensuring that the **model performs well in identifying both satisfied and dissatisfied customers**, which is relevant in our business context
- 3 AUC evaluates how well the model distinguishes between the two classes regardless of their distribution, making it **robust against class imbalance**

	XGBoost	NN	AutoML (XGBClassifier)	SVC	Decision Tree	Logistic Regression
Validation AUC	0.992	0.991	0.988	0.986	0.974	0.900
Rank	1	2	3	4	5	6

# Top 3: entertainment, seat comfort & online booking

## Feature Importance (XGBoost model)



## Top 3 Features



**Inflight entertainment**



**Seat comfort**



**Ease of online booking**

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- **Impact**

# Airlines can focus on 3 main areas of improvement



## Inflight entertainment

### Action Items:

- 1 Expand variety of content available** to cater to wider passenger preferences (incl. different age groups, languages)
- 2 Invest in high-quality screens with better resolution and responsiveness** and user interfaces
- 3 Update the entertainment content regularly** to include the latest content



## Seat comfort

### Action Items:

- 1 Work with seat manufacturers** to design **more ergonomic seats that provide better lumbar support** and adjustable features like headrests
- 2 Reconfigure the seating layout** to **offer more legroom**, especially in economy class
- 3 Use higher-quality, more comfortable seat cushions** that offer better support during long flights




## Ease of online booking


### Action Items:


- 1 Simplify the online booking interface** to **reduce the number of steps required to complete the booking**
- 2 Ensure the online booking system** is fully **optimized for mobile devices**, including through an own app
- 3 Integrate real-time customer support options**, such as chatbots and live chat, with the booking process


# A dashboard serves as insight hub to airline managers


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 Overview

 Detailed Analysis

 Predict Satisfaction

 Batch Prediction

 Feature Importance

Deploy ⋮

## Dashboard

### Welcome to the Airline Customer Satisfaction Dashboard!

This dashboard provides an in-depth analysis of airline customer satisfaction data. It includes several subpages that can be accessed from the sidebar on the left:

- **Overview:** Introduction to the dashboard and the data.
- **Detailed Analysis:** Visualizations and analysis of customer satisfaction and related features.
- **Predict Satisfaction:** Predict the satisfaction of individual customers based on input features.
- **Batch Prediction:** Upload an Excel file to predict the satisfaction of multiple customers at once.
- **Feature Importance:** Display the importance of different features in the XGBoost model used for prediction.

### Dataset

Below is the data used in this dashboard and for model training (XGBoost):

	satisfaction	Customer Type	Age	Type of Travel	Class	Flight Distance	Seat comfort	Foo

# A dashboard serves as insight hub to airline managers

