

The background features a dark blue grid. A white line chart with circular markers is overlaid on the grid, showing a fluctuating trend. The chart starts at a low point, rises to a peak, falls to a trough, and then generally trends upwards with some minor fluctuations.

Aviacion Mexicana: *Facing Post-Pandemic Challenges*

**This investigation is brought to you
by:**

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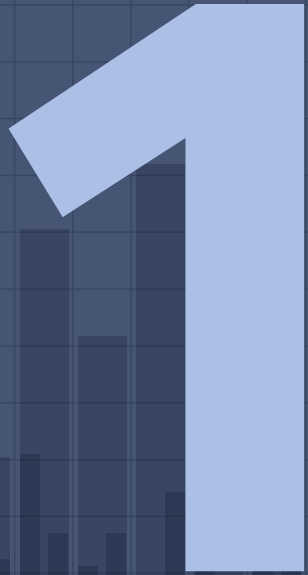




Are **loyalty programs** a valid
strategy for airline companies to
overcome **post-pandemic
hardships**?

Why is this important?

The whole world changed in 2020, but the air transport industry was one of the most affected...



Pandemic impact on Air Travel

- According to IATA, 1.8 billion passengers flew in 2020, a **decrease of 60.2%** compared to the 4.5 billion who flew in 2019.
- The total industry passenger **revenues fell by 69%** to \$189 billion in 2020, and net losses were \$126.4 billion in total.
- On average, there was a **\$71.7 loss incurred per passenger** in 2020.

Loyalty Programs



Companies have been actively looking for ways to spend their resources effectively and efficiently. With no dollar to spare, they are looking into many alternatives that have the most impact in recovering their operations.

Are they relevant? How can they help?

80%

of a company's future revenue will come from just 20% of existing clients.

5%

Higher costs to acquire a new customer over keeping an existing one satisfied and happy.

Database

To accomplish the objective of our investigation, we have chose a database that contains the outcomes of 103,904 airline passenger satisfaction survey.



kaggle

Database

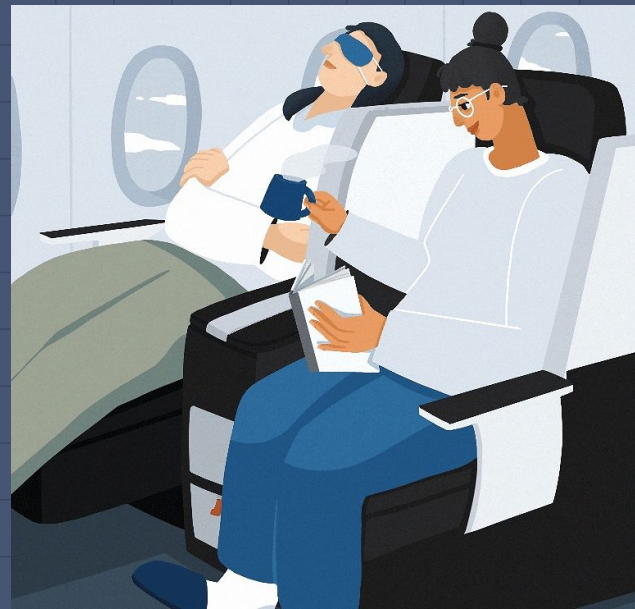
The dataset contains the following variables:

	id	gender	customer_type	age	type_travel	class_name	flight_distance	departure_delay	arrival_delay	satisfaction
sid										
1	70172	Male	Loyal Customer	13	Personal Travel	Eco Plus	460	25	18	neutral or dissatisfied
2	5047	Male	disloyal Customer	25	Business travel	Business	235	1	6	neutral or dissatisfied
3	110028	Female	Loyal Customer	26	Business travel	Business	1142	0	0	satisfied
4	24026	Female	Loyal Customer	25	Business travel	Business	562	11	9	neutral or dissatisfied
5	119299	Male	Loyal Customer	61	Business travel	Business	214	0	0	satisfied

Database

Can we predict whether a customer will become loyal?

The analysis could help the airline invest in loyalty programs based on travel rewards



Project Development:

Building a Machine Learning Model

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STEP 1 - Decision Making

The model chosen was logistic regression. These models are used to analyze data and, mathematically **determining the probability of new samples belonging to a class.** There are only two possible answers.

STEP 2 - Application

For this investigation, the objective was to **predict whether new purchasing clients will be loyal or disloyal.** The purpose of predicting this outcome is to create better loyalty programs and to dedicate resources to the right customers.

STEP 3 - Validation

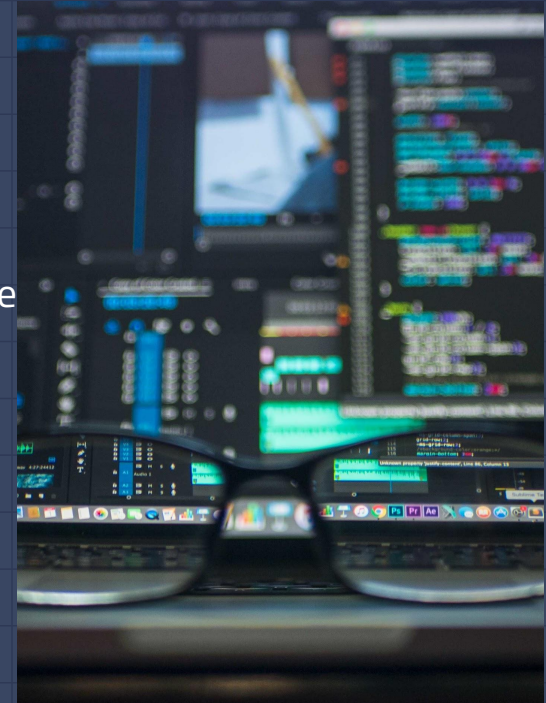
After creating the model, it was confirmed that **this particular data could be correctly analyzed using this strategy.**

Data Preprocessing

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In order to clean and prepare the data for the model, several steps were established. They go as follows:

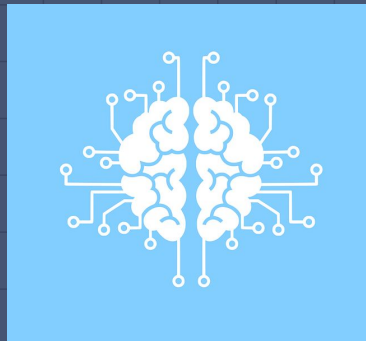
1. Detect **missing values** with the Pandas DataFrame function **isna()**.
2. Make sure we are using the correct variables data types -Pandas DataFrame **dtypes** property.
3. Use Scikit-learns LabelEncoder module to **transform categorical and text variables into numerical data.**
4. **Drop** the identification **id column.**
5. **Verify the information** on the DataFrame, including the index type and columns, non-null values, and memory usage.



Machine Learning

FEATURE ENGINEERING & FEATURE SELECTION

We decided to keep every column except for any that worked as an identifier. Since the rest of the variables are **facts that we consider could have an important impact** on the customer loyalty.



TRAINING & TESTING SETS

We split the dataset into **random** train and test subsets using the Scikit-learn `train_test_split` module. The training subset will be used in the model to learn from it and the testing subset to assess its performance.

CLASSIFICATION MATRIX

After creating the model, we obtained these results:

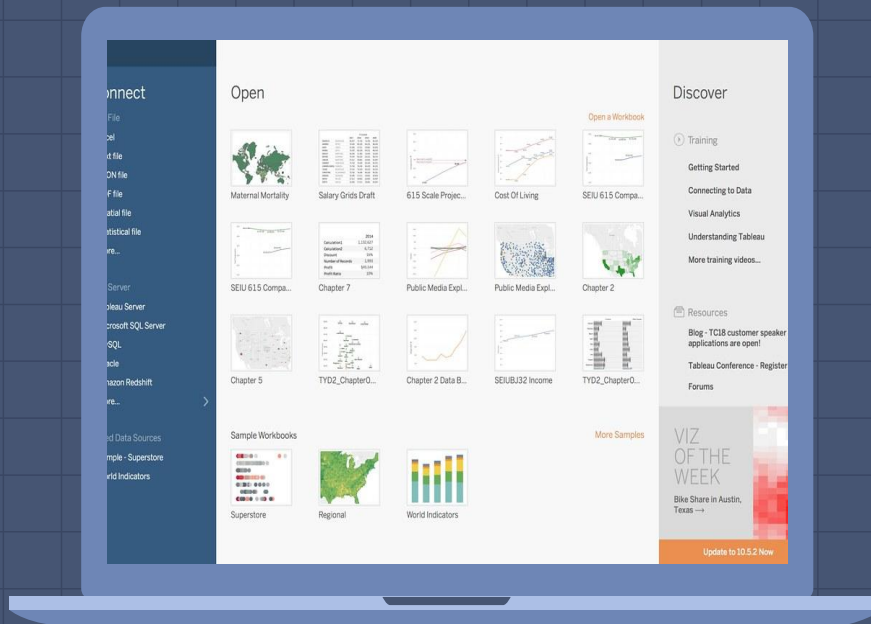
	precision	recall	f1-score	support
0	0.92	0.95	0.93	21231
1	0.72	0.63	0.67	4745
accuracy			0.89	25976
macro avg	0.82	0.79	0.80	25976
weighted avg	0.88	0.89	0.89	25976

A close-up, slightly blurred photograph of a stack of newspapers. The top newspaper's masthead 'WORLD BUSINESS' is clearly visible in bold black letters. Below it, the page number 'B3' is printed. Some text from the article is visible, including '...countries have' and '...Gen. Tha'. The background shows more layers of newspapers, creating a sense of depth and volume. The overall color palette is muted, with greys, blues, and whites.

**Big picture: How
does data look like
today?**

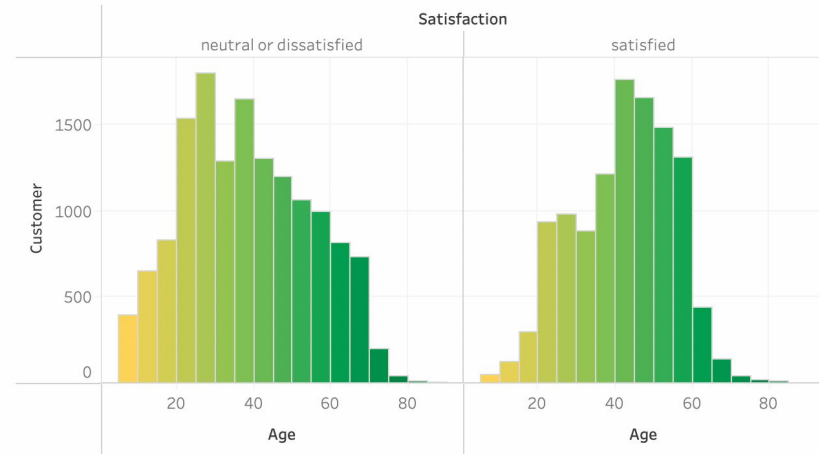
Why Tableau?

- We choose Tableau to create our dashboard because it is a powerful tool with data visualization capabilities and data can be analyzed very quickly.
- One of the most interesting tools that Tableau provides is interactivity. Some of the graphs created can be modified by the viewer using several filters in order to make a deeper analysis.
- The tableau dashboard can be viewed publicly here: https://public.tableau.com/app/profile/jesus.salas6470/viz/Dashboard_16464975652980/FinalDashboard

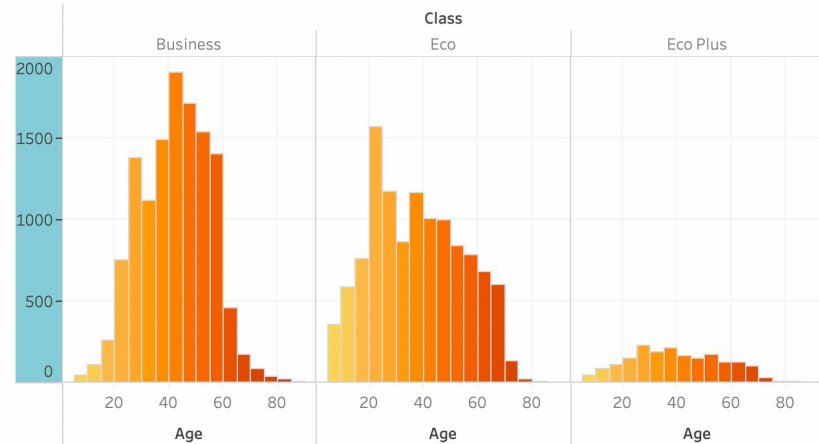


The age distribution between satisfied and dissatisfied compared with the type of class

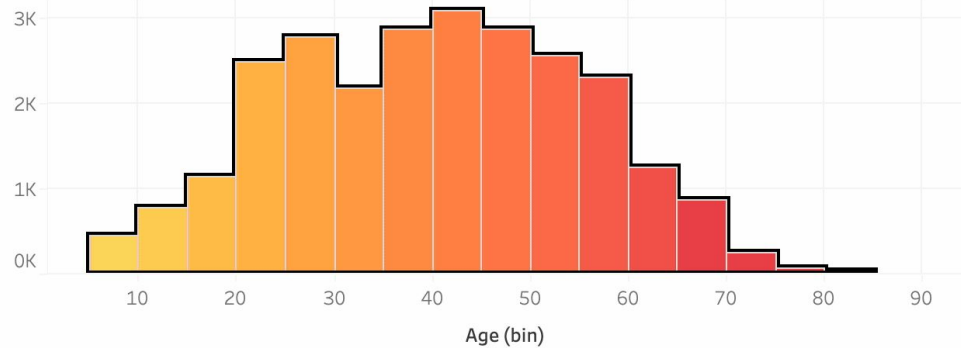
Satisfaction Age



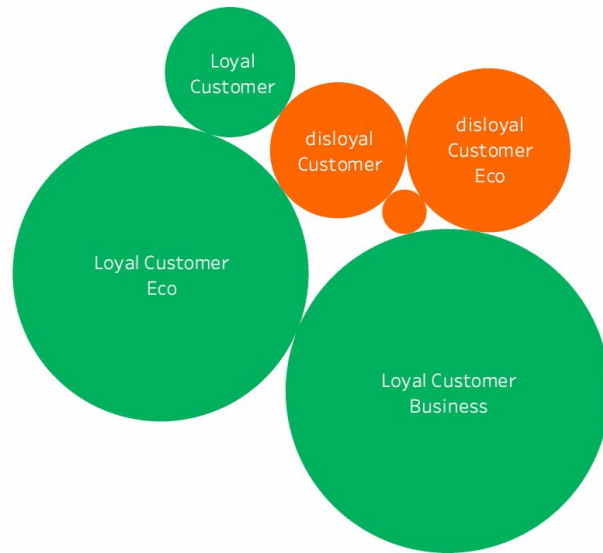
Class Type



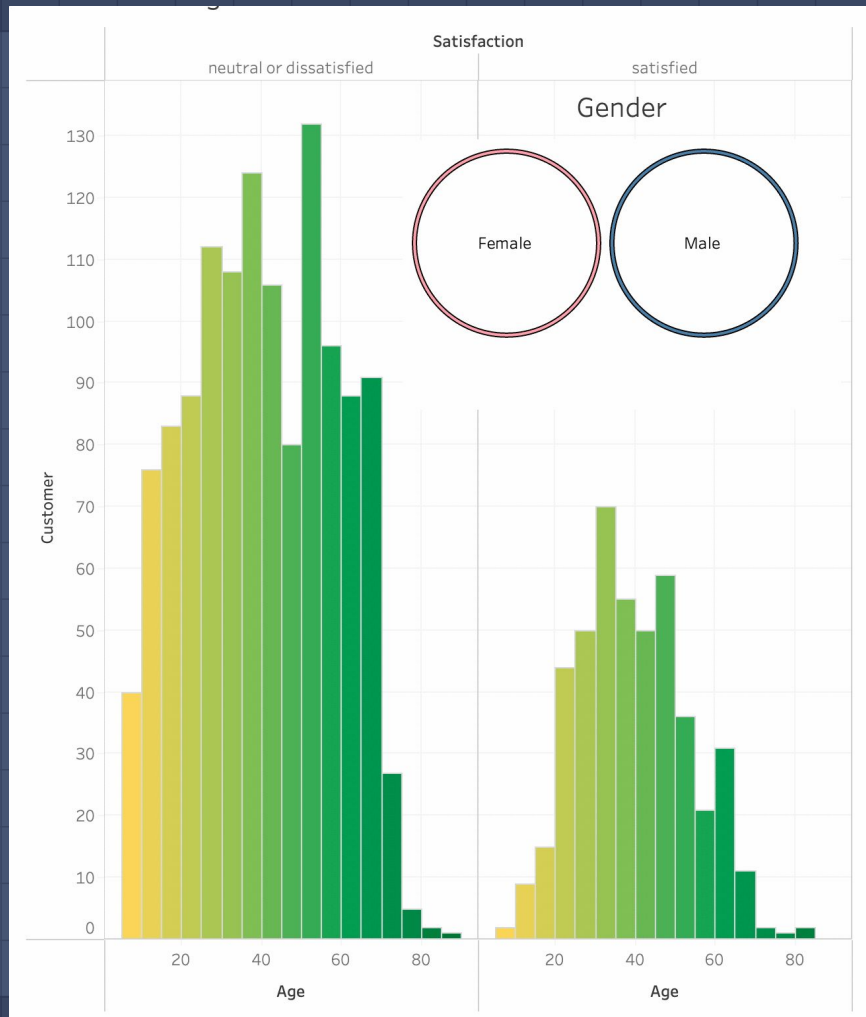
Passengers between 40- 45 and 20-25 are the largest group and all are Loyal customers.



Customer Dist



Males between 30 and 35 years old and females between 50 to 55 years represent the largest group of dissatisfied customers



Comparing
the actual
data with
the
machine
learning
algorithm
prediction

Conclusions & Recommendations

As mentioned by my colleagues, the model worked correctly to identify the customer type which will become part of a bigger marketing strategy related to loyalty programs for Mexicana Aviacion.

We recommend as a topic for further research to look into the features that have the 20-25 age group dissatisfied in order to retain them as loyal clients in the future.

THANKS!

