

BACKGROUND

As someone who has always been interested in travel, I understand that the journey, primarily consisting of inflight time, is an essential part of the traveling experience. Growing up in Singapore, every year I would read the news about the [Skytrax World Airline Awards](#), seeing Singapore Airlines in one of the top spots, often first place. But, **what exactly makes an airline the best airline in the world?**

The problem questions:

- How can airlines know which factors are most important for/the strongest contributors to customer satisfaction?
- Which factors contributing to satisfaction correlate with one another?
- Are a random set of customers more likely to be satisfied or neutral/dissatisfied with an airline experience?

OBJECTIVE

- Clean, fit a model to predict, and visualize & interpret the passenger satisfaction data to find the relationships between category ratings and overall satisfaction of the passenger, as well as evaluate the results.

DATA

	A	B	C	D	E	F	G	H	I	J	K	L	M
	id	Gender	Customer Type	Age	Type of Travel	Class	Flight Distance	Inflight wifi service	Departure/Arrival time convenient	Ease of Online booking	Gate location	Food and drink	
1	0	70172 Male	Loyal Customer	13	Personal Travel	Eco Plus	460	3	3	4	3	1	5
2	1	5047 Male	disloyal Customer	25	Business travel	Business	235	3	2	2	3	3	1
3	2	110028 Female	Loyal Customer	26	Business travel	Business	1142	2	2	5	2	2	5
4	3	24026 Female	Loyal Customer	25	Business travel	Business	562	2	5	3	5	5	2
5	4	119299 Male	Loyal Customer	61	Business travel	Business	214	3	3	3	3	3	4
6	5	111157 Female	Loyal Customer	26	Personal Travel	Eco	1180	3	4	1	2	1	1
7	6	82113 Male	Loyal Customer	47	Personal Travel	Eco	1276	2	4	2	3	3	2
8	7	96462 Female	Loyal Customer	52	Business travel	Business	2035	4	3	3	4	4	5
9	8	79485 Female	Loyal Customer	41	Business travel	Business	853	1	2	2	2	2	4
10	9	65725 Male	disloyal Customer	20	Business travel	Eco	1061	3	3	3	3	4	2
11	10	34991 Female	disloyal Customer	24	Business travel	Eco	1182	4	5	2	5	4	2
12	11	51412 Female	Loyal Customer	12	Personal Travel	Eco Plus	308	2	4	4	2	2	1
13	12	98628 Male	Loyal Customer	53	Business travel	Eco	834	1	4	4	4	4	1
14	13	83502 Male	Loyal Customer	33	Personal Travel	Eco	946	4	2	4	4	3	4
15	14	95789 Female	Loyal Customer	26	Personal Travel	Eco	453	3	2	2	3	2	2
16	15	100580 Male	disloyal Customer	13	Business travel	Eco	486	2	1	2	2	3	4
17	16	71142 Female	Loyal Customer	26	Business travel	Business	2123	3	3	3	3	3	4
18	17	127461 Male	Loyal Customer	41	Business travel	Business	2075	4	4	2	2	4	4
19	18	70354 Female	Loyal Customer	45	Business travel	Business	2486	4	4	4	4	4	3
20	19	66246 Male	Loyal Customer	38	Personal Travel	Eco	460	2	3	3	3	2	5

	O	P	Q	R	S	T	U	V	W	X	
	Seat comfort	Inflight entertainment	On-board service	Leg room service	Baggage handling	Checkin service	Inflight service	Cleanliness	Departure Delay in Minutes	Arrival Delay in Minutes	satisfac
1	2	5	5	4	3	4	4	5	25	18	neutral
2	3	1	1	1	5	3	1	4	1	6	neutral
3	4	5	4	4	3	4	4	5	0	0	satisfie
4	5	2	2	5	2	3	1	4	11	9	neutral
5	6	3	3	4	4	4	3	3	0	0	satisfie
6	7	1	1	3	4	4	4	1	0	0	neutral
7	8	2	2	3	3	4	3	2	9	23	neutral
8	9	5	5	5	5	5	4	4	4	0	satisfie
9	10	3	1	2	1	4	4	1	0	0	neutral
10	11	3	2	2	2	4	4	3	0	0	neutral
11	12	2	2	3	3	5	3	2	0	0	neutral
12	13	1	1	1	2	5	5	1	0	0	neutral
13	14	1	1	1	1	3	4	4	28	8	neutral
14	15	4	4	4	5	2	2	4	0	0	satisfie
15	16	2	2	4	3	2	1	2	43	35	neutral
16	17	1	4	2	1	4	3	4	1	0	neutral
17	18	4	4	5	3	4	5	4	49	51	satisfie
18	19	4	5	5	5	3	5	5	0	10	satisfie
19	20	5	5	5	5	3	3	4	7	5	satisfie
20	21	5	5	1	2	4	2	5	17	18	neutral

WHAT WE HAVE DONE ACTUALLY?

Table of Contents:

1. Data Cleaning
2. Data Visualization and Analysis
3. Fitting a logistic regression model to the data
4. Clustering
5. Interpretation and Evaluation

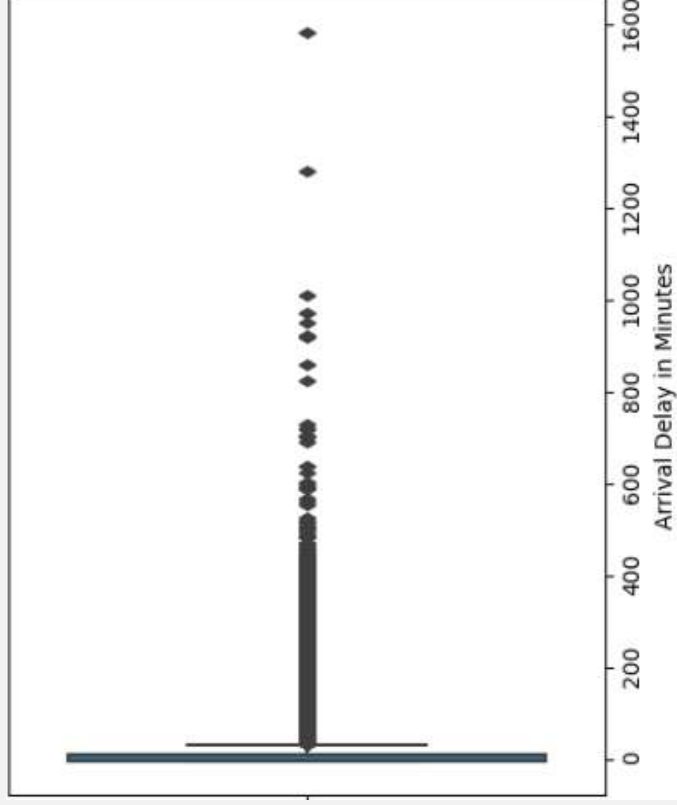
DATA CLEANING

Understanding the Structure of the Dataset

- The data has 103,904 entries of 25 columns.
- We can see that each row represents the data for one customer, and that there is information about their 1-5 ratings (0 if there is no data on what they rated that category) for each category.
- *There are a few things to notice in terms of data cleaning:*
 - a) The column "Arrival Delay in Minutes" has 103904-103594 = **310** missing values
 - b) The first two columns are unnecessary, as they are not relevant to the analysis

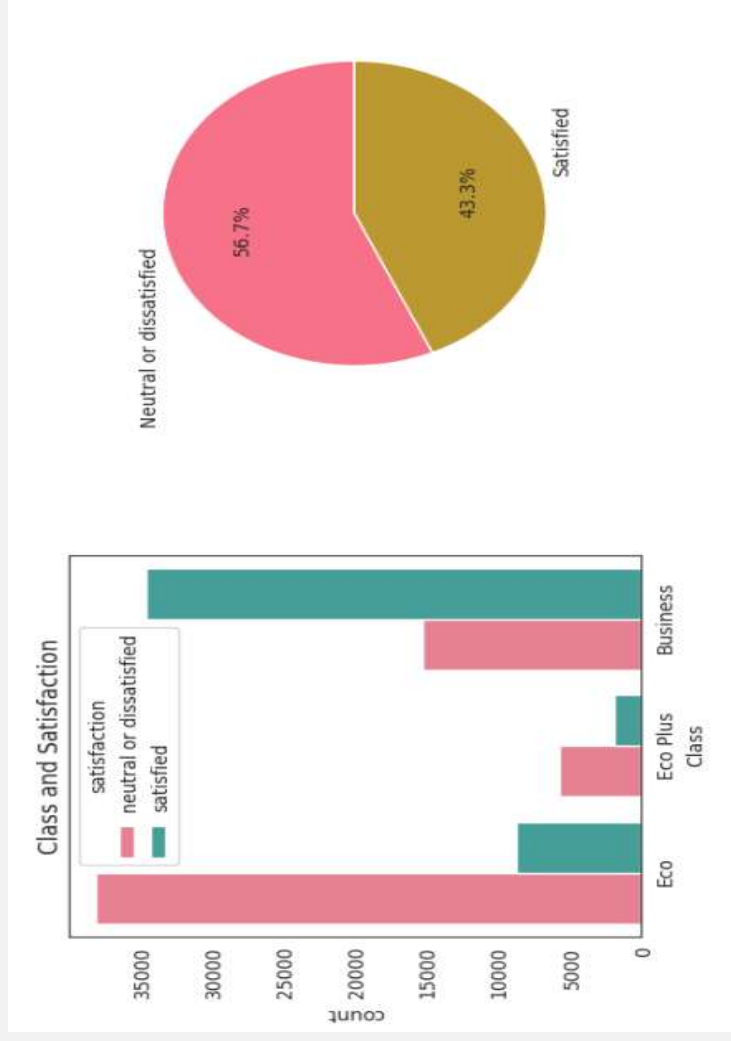
PREPARING DATA FOR ANALYSIS

(THAT MISSING VALUE)



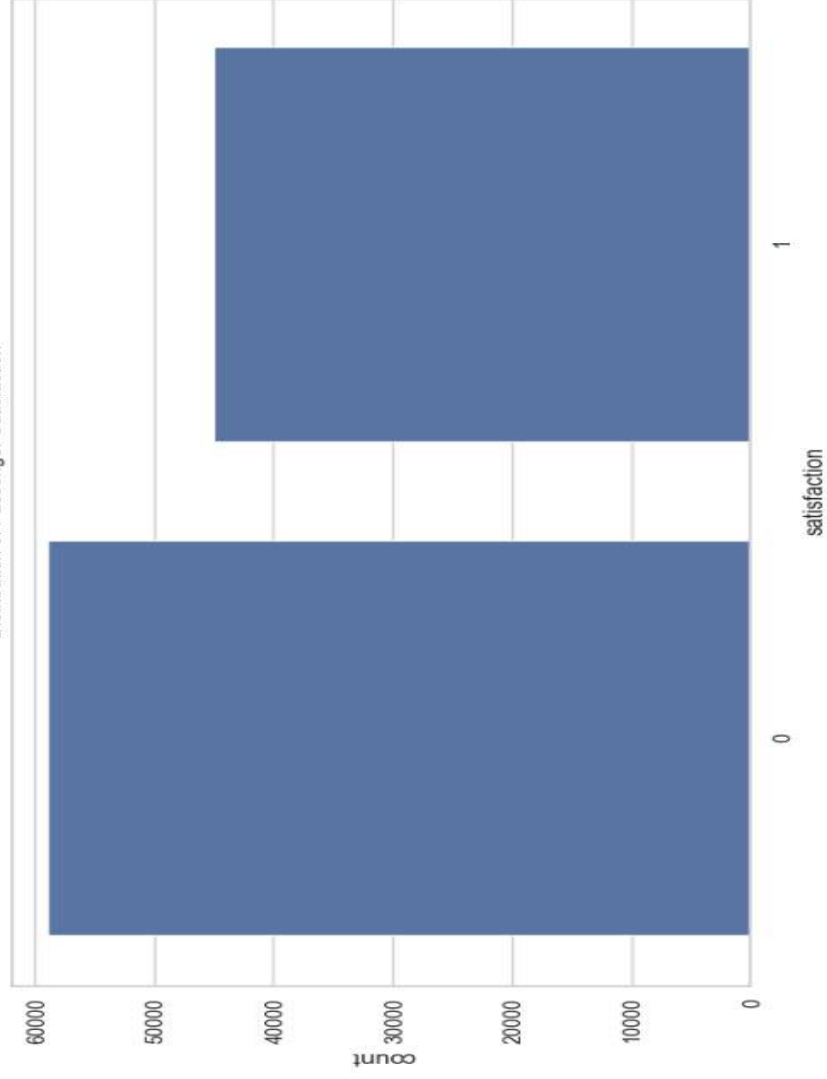
- Despite the mean of the 'Arrival Delay in Minutes' column being 15.13, we can see from the boxplot that almost any value except 0 is an outlier.
- The mode of the column is 0 by a large margin, and the 310 missing values are few compared to the large number of total entries.
- In considering this along with most values being outliers other than 0, I will fill in the Average Delay in Minutes column's null values with 0.

DATA VISUALIZATION AND ANALYSIS

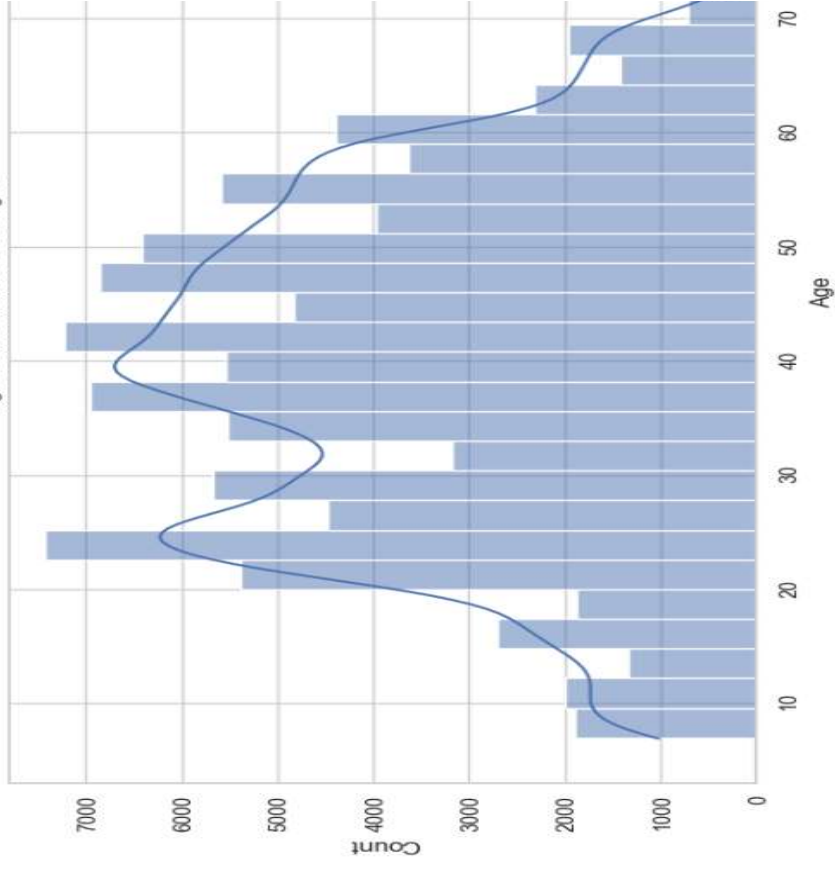


- 56.7% of the passengers in the survey were neutral or dissatisfied, the rest being satisfied with the airline.
- It is reasonable to assume that the class passengers are traveling in affects their satisfaction, which is corroborated by the bar plot on the left.
- Economy and Economy Plus tend to have more neutral or dissatisfied customers than satisfied customers, whereas Business Class has more satisfied customers than dissatisfied customers.
- Notably, Economy Plus has the lowest count, therefore, there is an unequal distribution in the class of customers in the dataset, however, this may also be because Economy Plus is generally flown the least.

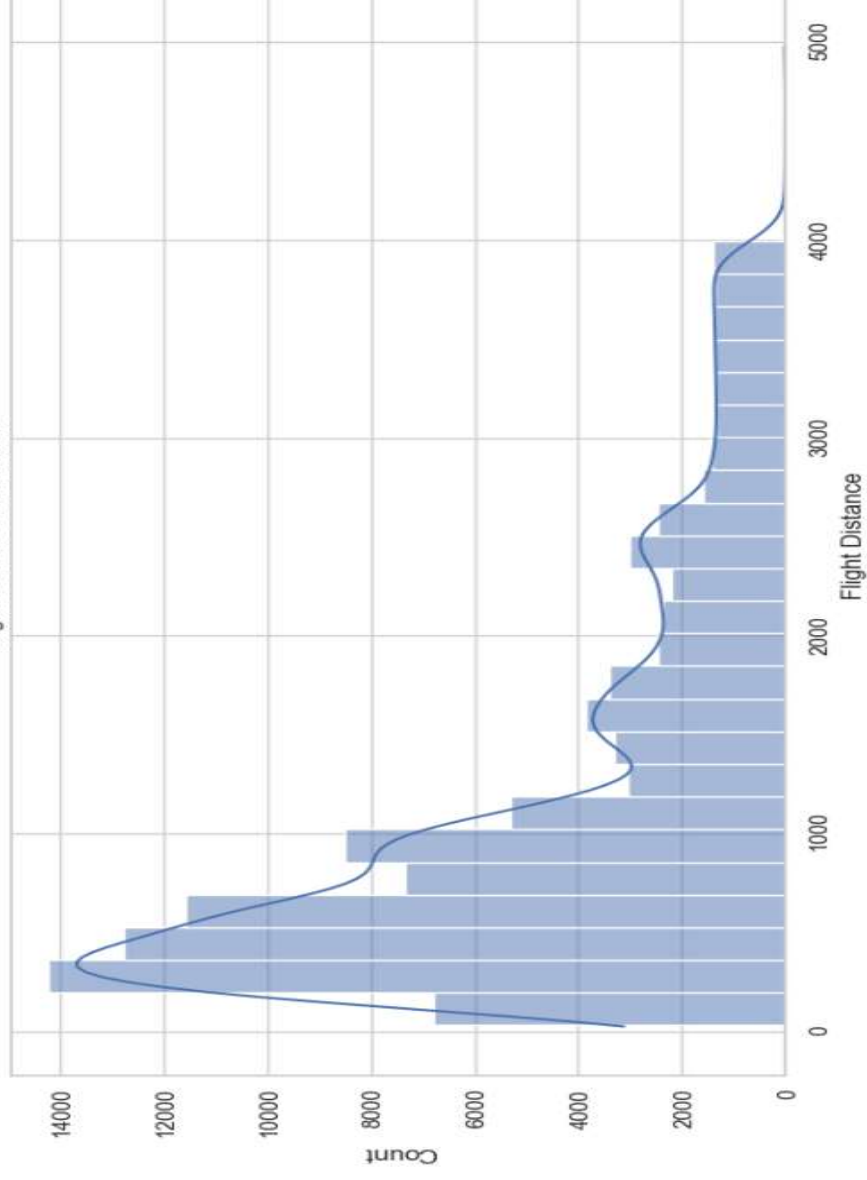
Distribution of Passenger Satisfaction



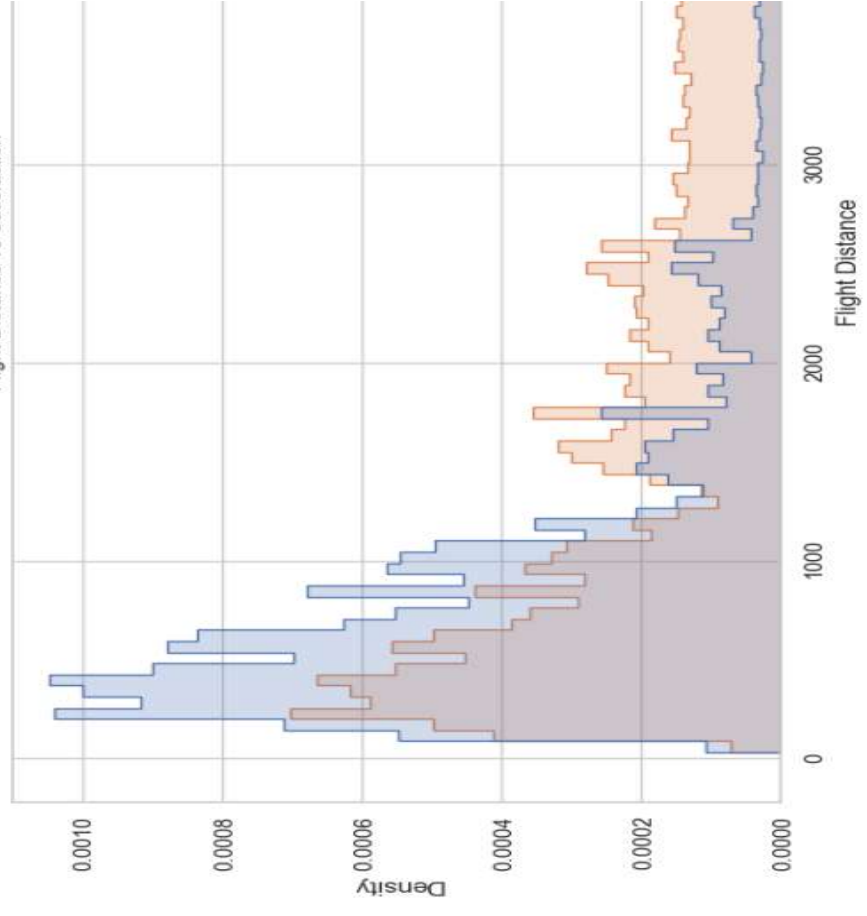
Age Distribution of Passengers



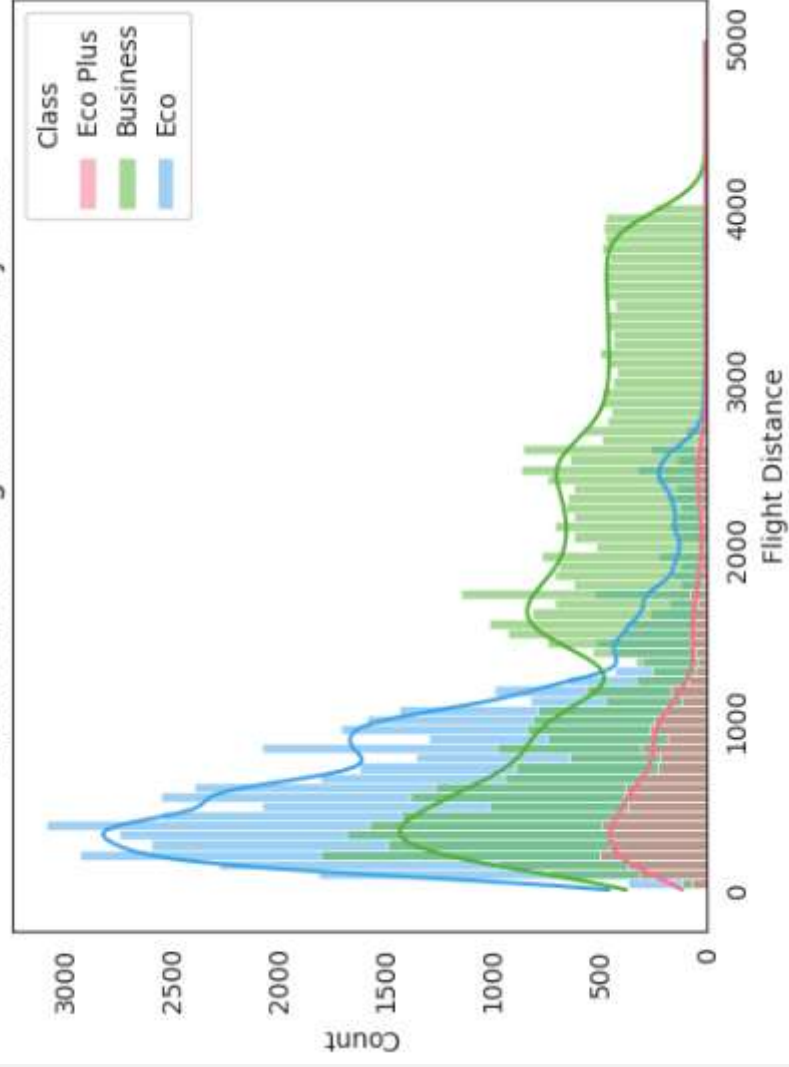
Flight Distance Distribution



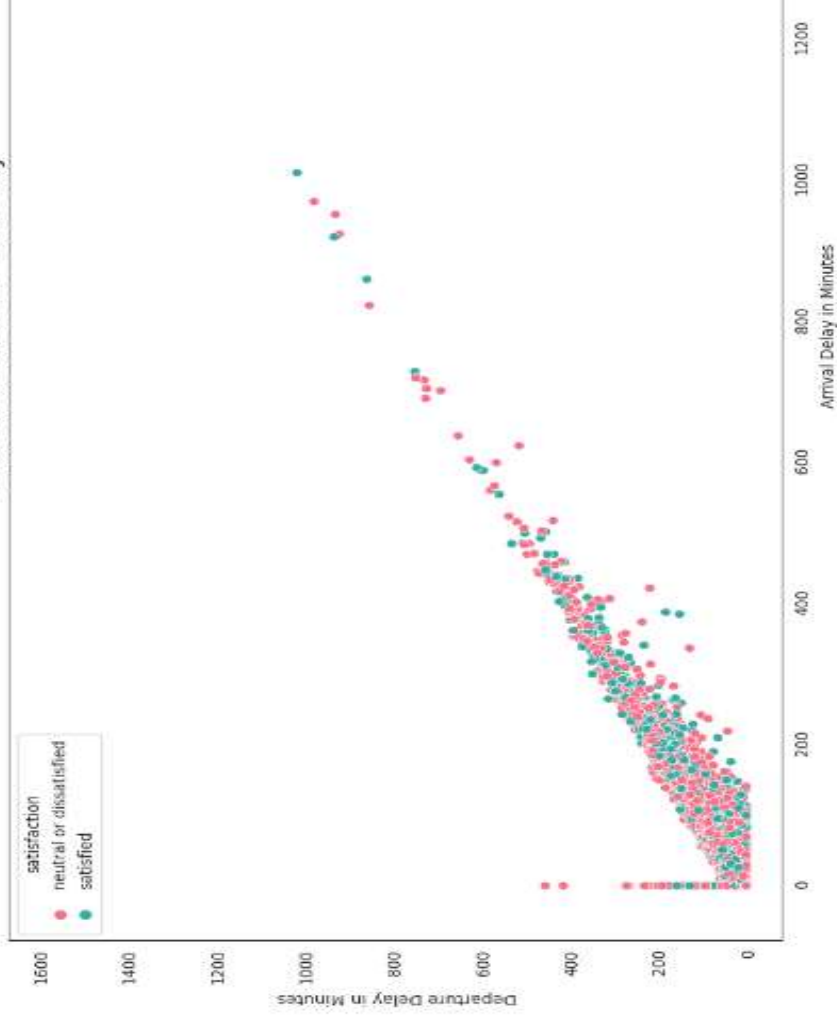
Flight Distance vs Satisfaction



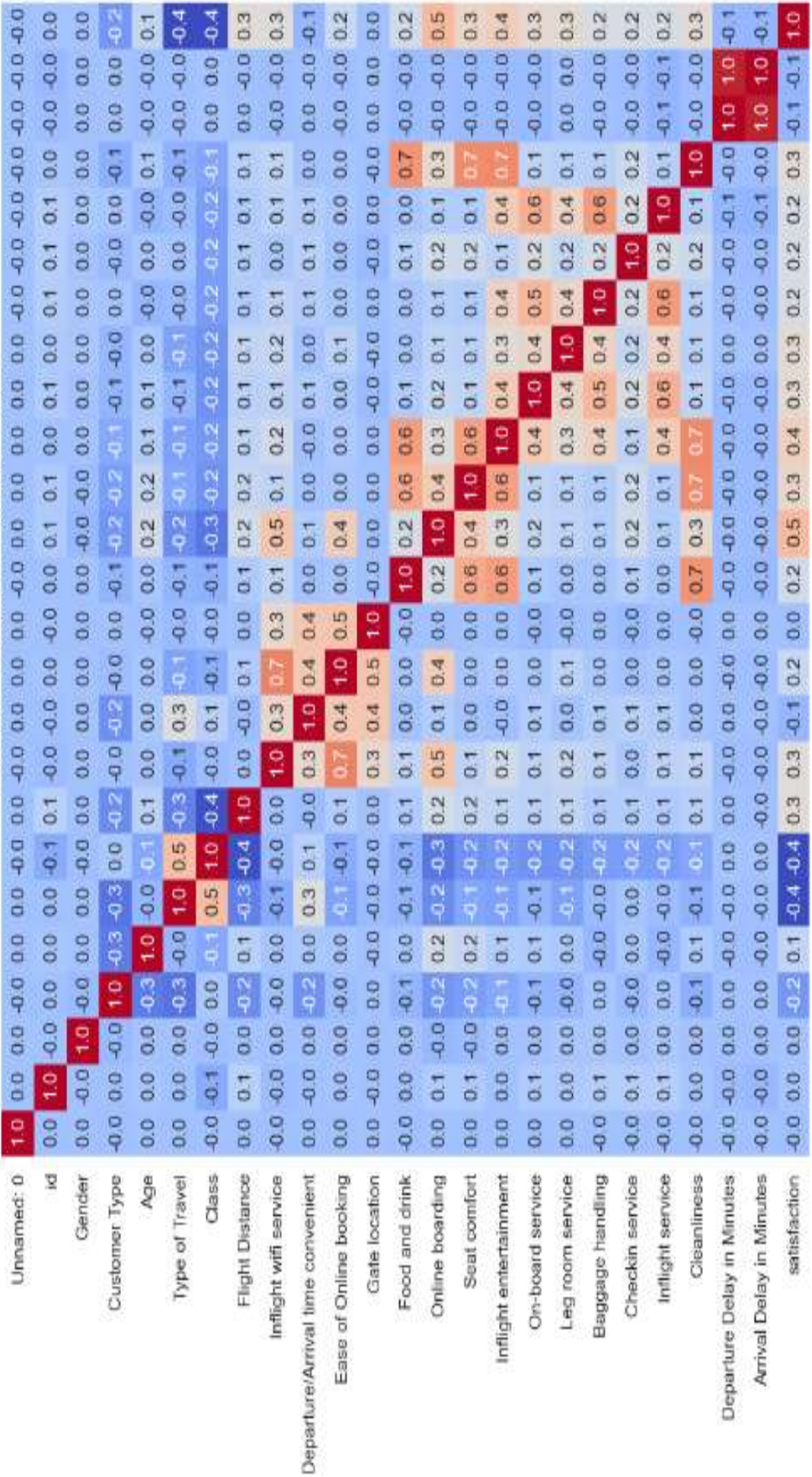
Distribution of Flight Distance by Class



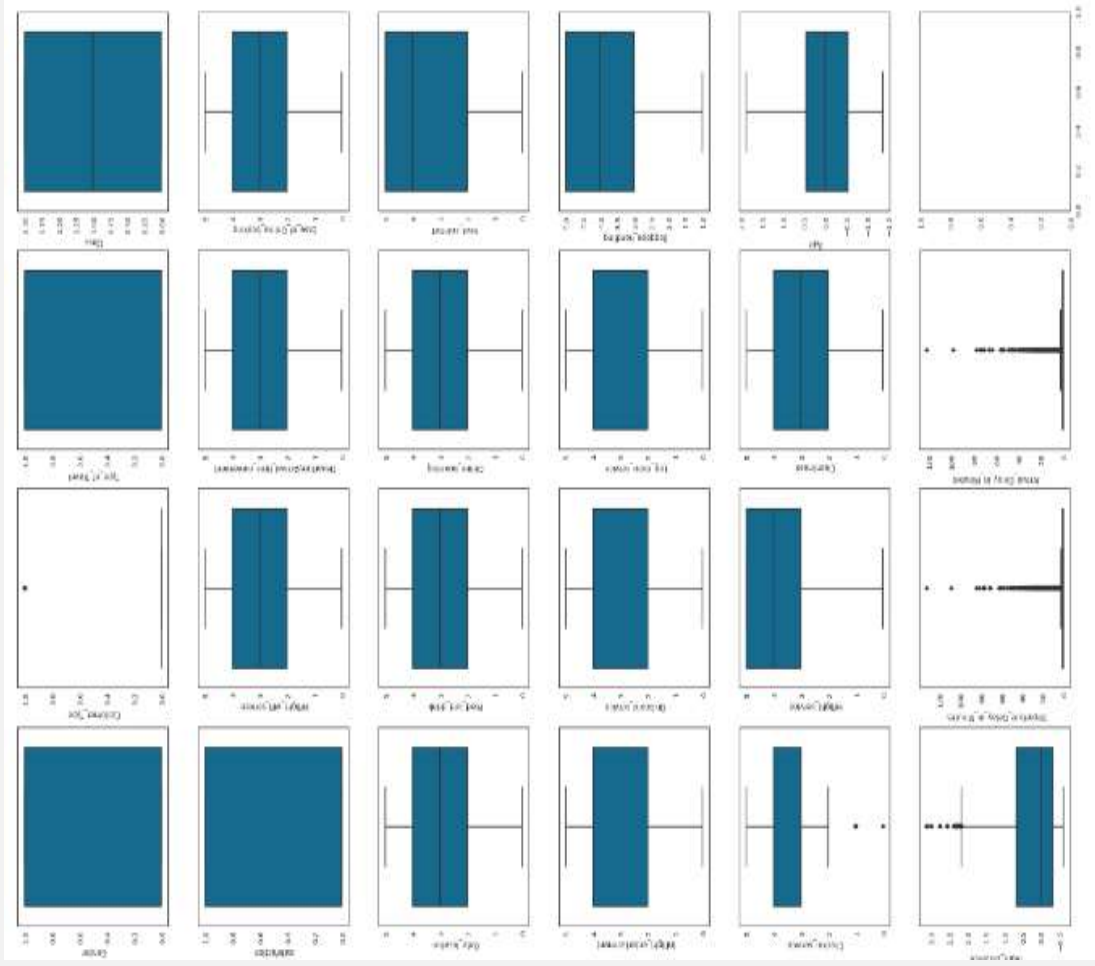
Satisfaction based on Delay



Correlation Heatmap of Numerical Features



OUTLIER DETECTION



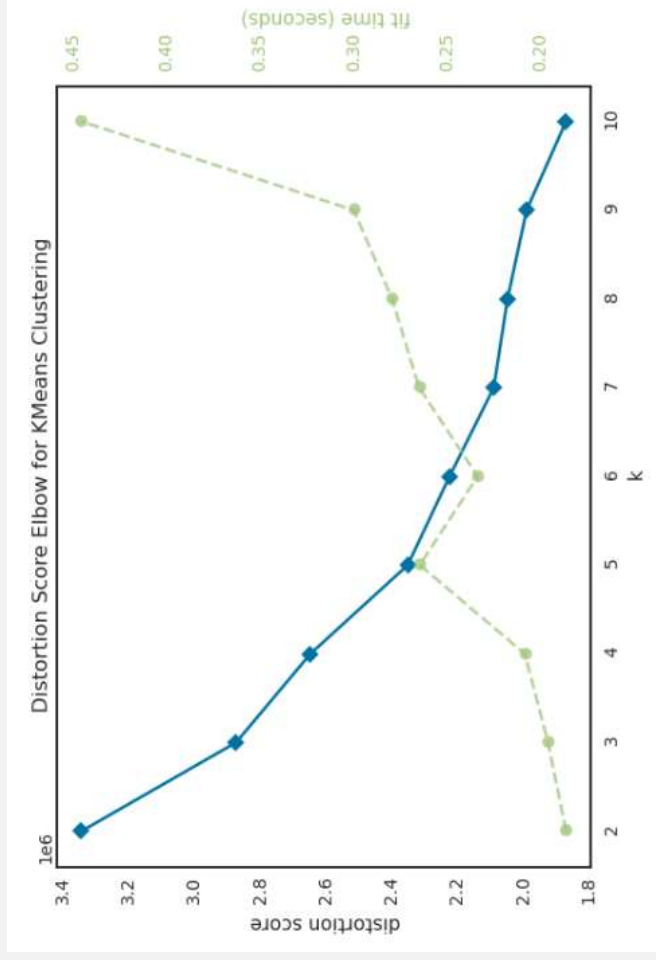
- There are four variables that catch our eye in terms of
- 'Flight Distance' > 2.5: 58 observations (~0.05%)
 - 'Check-in service' < 1: 1 observation (~0%)
 - 'Departure Delay in Minutes' > 100: 2 observations (~0)
 - 'Arrival Delay in Minutes' > 100: 1 observation (~0%)

DIMENSIONALITY REDUCTION (PCA METHOD)

	0	1	2	3	4	5	6	7	8	9	10	11	12
0	0.643481	-2.838255	1.528182	0.416362	1.525427	0.455497	1.696304	-0.958217	-0.635927	-0.612674	0.803281	1.032057	0.000000
1	-1.139223	4.151076	-1.740010	-1.240153	-2.275673	-2.218773	-1.359591	-0.771596	-0.129489	-1.628197	-0.369051	-0.299546	-0.000000
2	-1.823286	-2.830938	2.882431	1.003769	-0.619484	0.646969	0.459538	0.622320	-0.066356	-0.250733	0.190945	-0.480558	-0.000000
3	-0.429559	1.984216	-3.671168	-0.678198	0.751446	-1.945542	-2.566615	0.207154	-0.608445	-1.069061	0.279280	0.010339	-0.000000
4	-1.730046	-1.267699	-0.075760	0.944302	-1.408517	0.170834	-0.730823	0.082743	-0.701223	-0.326915	-0.593180	-0.996033	1.000000

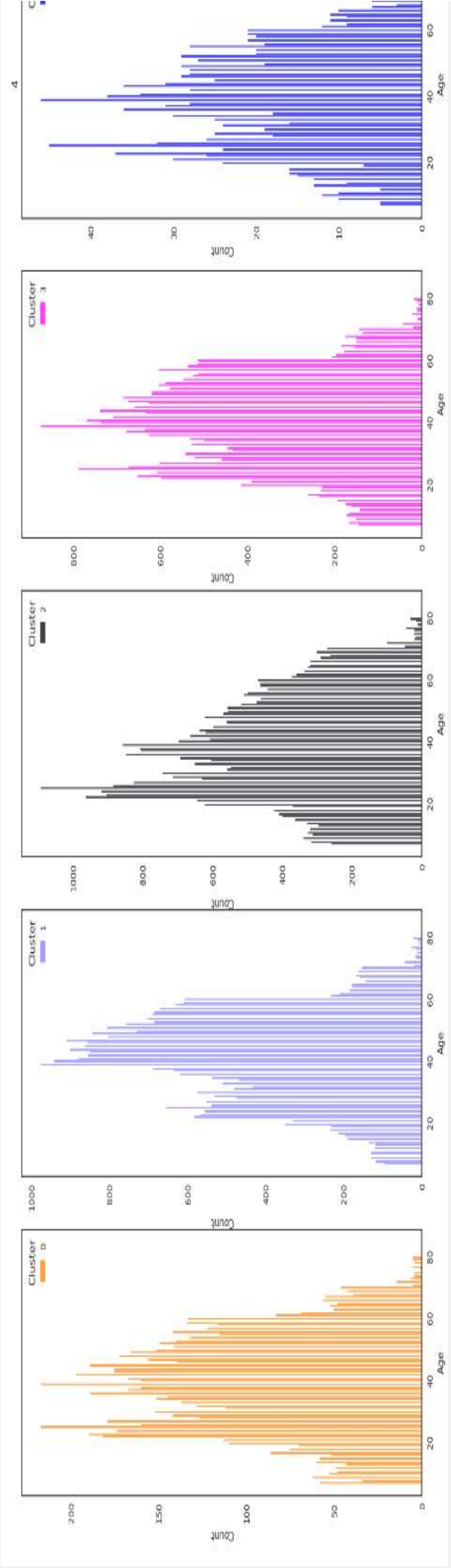
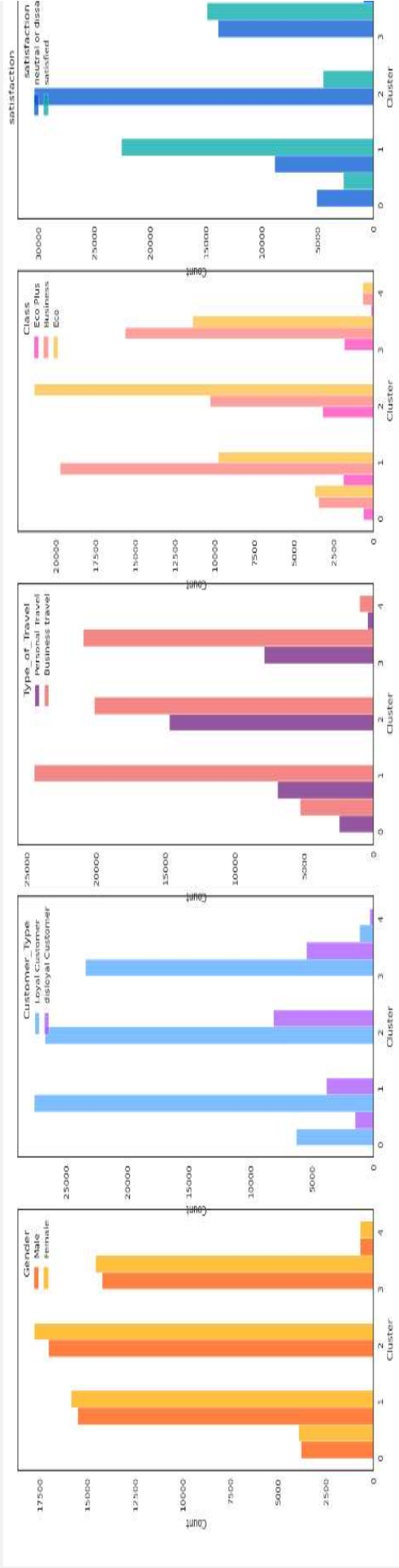
- **Conclusion:** The "dimensionally reduced" data frame above has 13 nameless variables (principle components) which would explain 95% of variance.

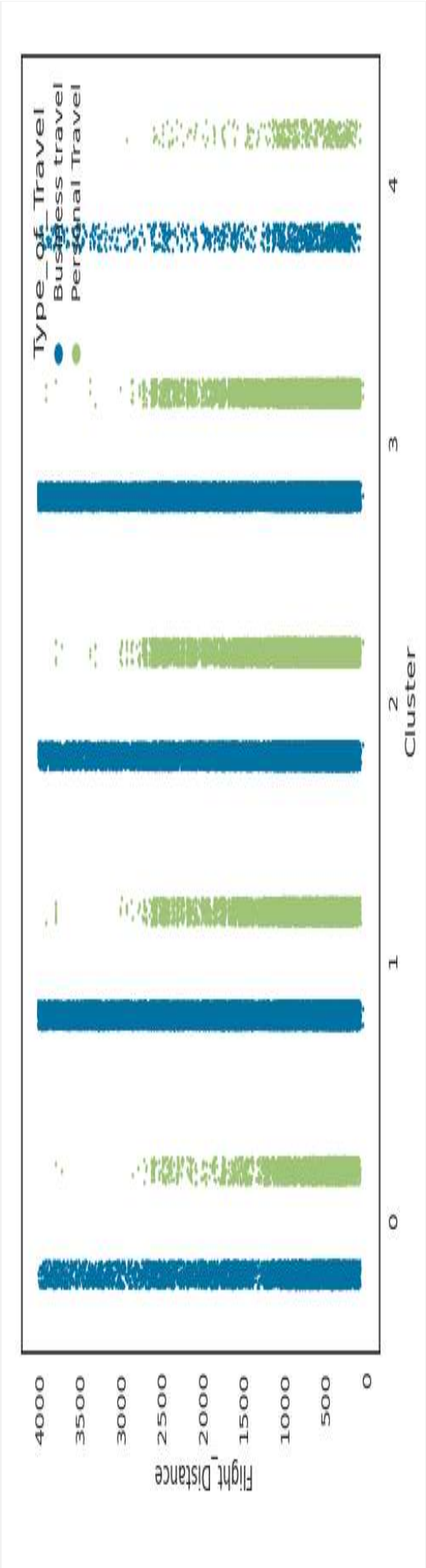
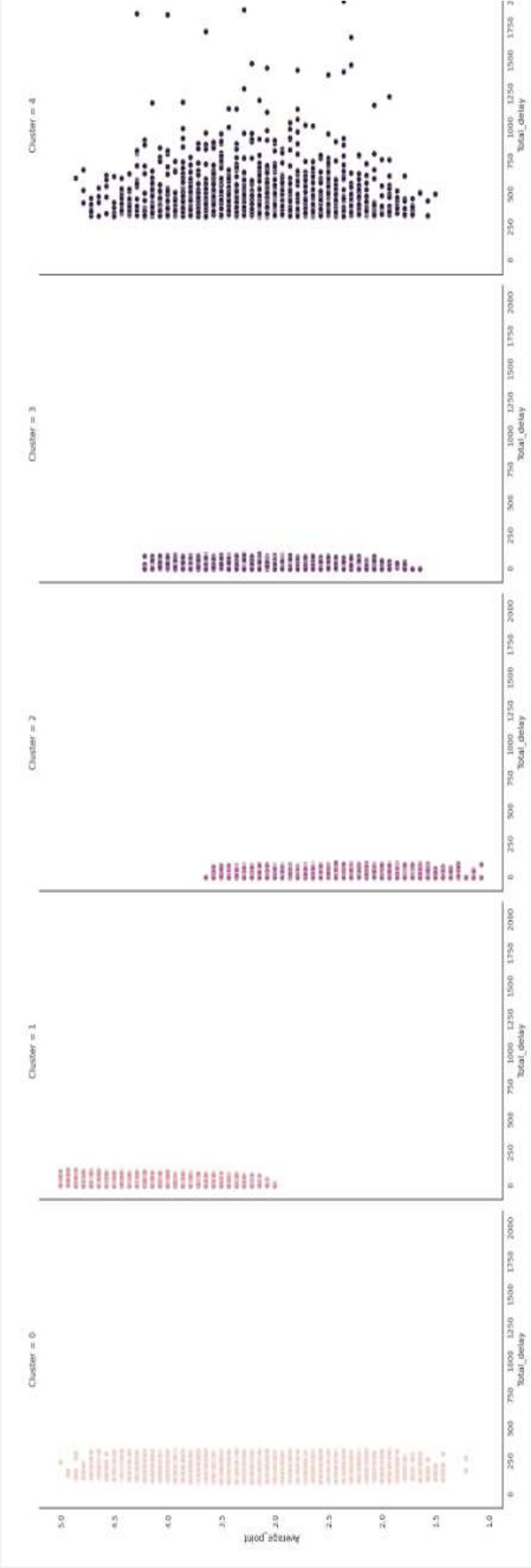
CLUSTER SEGMENTATION



With every new iteration, the curves may be different, but the result always boils down to 5 or 6 as the optimal number of clusters. **Let's take 5 clusters.**

DIFFERENT CLUSTERS ANALYSIS





CLUSTER ANALYSIS

Cluster 0

- not numerous cluster
- nearly equal proportion between women and men
- most customers are loyal
- 2 times more "business travels" than "personal travels"
- "Eco" and "Business" seats are booked equally; "Eco Plus" is booked the least
- 1/3 of passengers is "satisfied"
- absolute majority of passengers is of age between 20 and 60 year old
- air flights distances are usually not more than 1000 miles

Cluster 1

- a numerous cluster
- contains slightly more women than men
- most customers are loyal
- 3 times more "business travels" than "personal travels"
- "Business" seats represent 70% of bookings; "Eco Plus" is booked the least
- 72% of passengers are "satisfied" (best result)
- absolute majority of passengers is of age between 20 and 60 y.o.
- passengers tend to give the highest rating points: almost all of them lie between "3" and "5"

Cluster 2

- the most numerous cluster
- contains slightly more women than men
- most customers are loyal
- 30% more "business travels" than "personal travels"
- "Eco" seats are booked the most (65% of bookings); "Eco Plus" is booked the least
- 13% of passengers are "satisfied" (worst result)
- the biggest share of young (~20 y.o.) passengers; the share of older passengers constantly lowers until 65+ age

Cluster 3

- a numerus cluster
- contains slightly more women than men
- most customers are loyal
- 3 times more "business travels" than "personal travels"
- "Business" seats represent 55% of bookings; "Eco" booked the least
- 50% of passengers are "satisfied"
- absolute majority of passengers is of age between 20 and 60 y.o.
- passengers are conservative in giving ranking points; usually, figures lie within "2", "3" and "4"

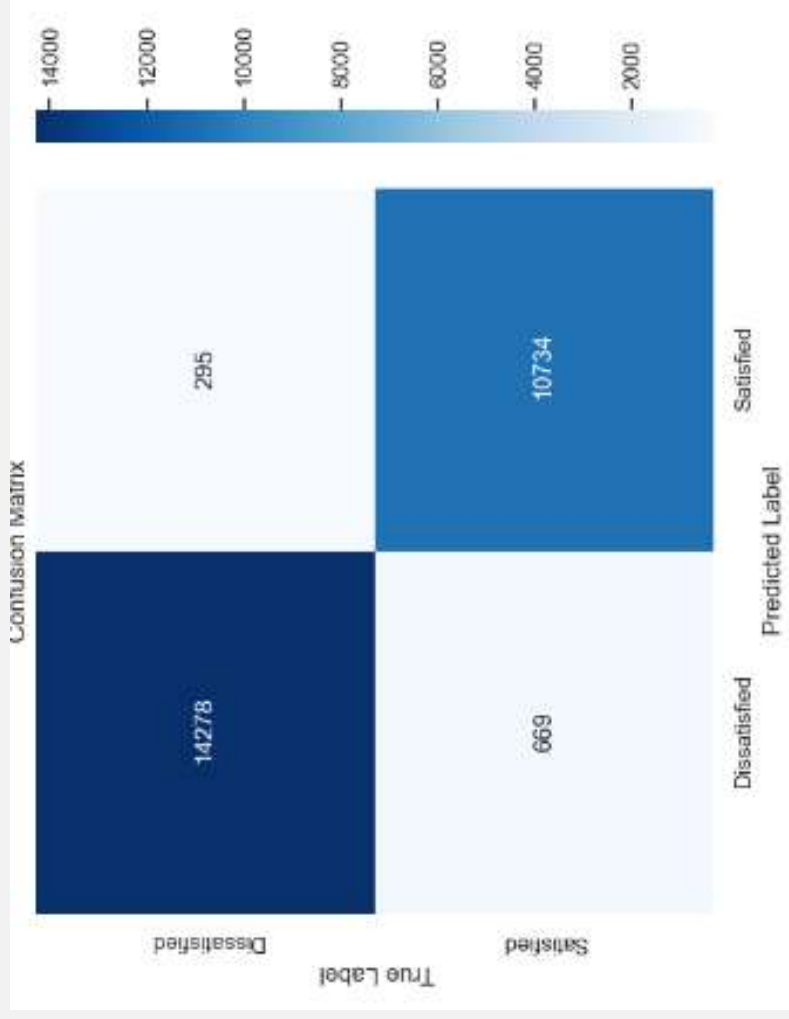
Cluster 4

- the least numerous cluster
- nearly equal proportion between women and men
- most customers are loyal
- 2 times more "business travels" than "personal travels"
- "Eco" and "Business" seats are booked equally; "Eco Plus" is booked the least
- 1/3 of passengers is "satisfied"
- majority of passengers is of age between 20 and 60 y.o., but children and older passengers (65+ y.o.) are also present to some extent
- most travels are up to 1000 miles long, but 2000 miles-long flights occurs as well

MODEL SELECTION

- It is an classification problem, so our basic instinct was to implement decision tree but decision tree is not much robust than input variable so in the end we decided to choose Random forest
- Model Building Random Forest Classifier because it handles both numerical and categorical data well and is robust against overfitting.
- Accuracy: 96.35% Precision, Recall, and F1-Score: Class 0 (Neutral or Dissatisfied): Precision = 95%, Recall = 98%, F1-score = 97% Class 1 (Satisfied): Precision = 98%, Recall = 94%, F1-score 96% =
- The metrics indicate that the model performs excellently in predicting passenger satisfaction, with high scores in both precision and recall across the satisfaction categories 96%

ACCURACY



The matrix shows a strong predictive performance true positive and true negative rates, indicating that is effective at classifying both satisfied and dissatisfied passengers correctly. True Positives (Satisfied correctly identified): The model has a high number of correct predictions for satisfied passengers. True Negatives (Dissatisfied correctly identified): Similarly, it accurately identifies a large number of dissatisfied passengers

RECOMMENDATION

	Feature	Importance
11	Online boarding	0.171534
6	Inflight wifi service	0.151017
4	Class	0.099753
3	Type of Travel	0.097657
13	Inflight entertainment	0.057292
12	Seat comfort	0.045061
5	Flight Distance	0.043389
15	Leg room service	0.039845
8	Ease of Online booking	0.037885
1	Customer Type	0.035961
2	Age	0.035749
14	On-board service	0.030285
18	Inflight service	0.026105
19	Cleanliness	0.025393
17	Checkin service	0.024873
16	Baggage handling	0.024517
9	Gate location	0.018942
7	Departure/Arrival time convenient	0.017735
10	Food and drink	0.012204
0	Gender	0.004804

Feature Importance Analysis Online Boarding (17.15%): The most influential feature, indicating that the online boarding experience plays a crucial role in determining passenger satisfaction.

Inflight Wifi Service (15.10%): Another significant factor, emphasizing the value of connectivity during flights. Class (9.98%): Reflects the impact of travel class on passenger satisfaction, with business class likely offering a more satisfying experience.

Type of Travel (9.77%): Differentiates between personal and business travel, affecting satisfaction levels.

Inflight Entertainment (5.73%): Entertainment options available inflight significantly influence satisfaction, especially on longer Recommendations Based on Analysis Enhance Online Services: Improving the online boarding process and inflight wifi service could lead to significant gains in passenger satisfaction.

Focus on Inflight Experience: Entertainment and comfort in different classes should be areas of focus for service improvements.