

## Title for your project

### Measuring customer satisfaction with airports and airplanes of the air transportation industry

Name of first dataset: Airline customer Satisfaction

- **Link:** <https://www.kaggle.com/datasets/sjleshac/airlines-customer-satisfaction>

Name of second dataset: Airport Quarterly Passenger Survey

- **Link:** <https://catalog.data.gov/dataset/airport-quarterly-passenger-survey>

## Names of teammates (Team 1)

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## Brief description/context

### First dataset context:

Customers have given their feedback in a variety of airline trips. The problem we are having has to do with customer satisfaction, there are many users that have been a long time with the airline and have shown dissatisfaction and likewise those users that are not loyal have also showed dissatisfaction with the service provided. It would be interesting to understand the role some variables play in the satisfaction result for these trips.

### Second dataset context:

Airport customers have given feedback on a variety of features of the airport. For each of the features customers provide their level of satisfaction and also give an overall satisfaction. The problem we have is that there are many features in which customers show a dissatisfaction with the value provided by the airport; it would be interesting to understand which variables are the most important to the overall customer satisfaction.

## Description of the dataset

The **first dataset** has 129,880 records and has 23 variables that were measured per trip as detailed below:

VARIABLES			
satisfaction	Gender	Customer Type	Age
Type of Travel	Class	Flight Distance	Seat comfort
Departure/Arrival time convenient	Food and drink	Gate location	Inflight wifi service
Inflight entertainment	Online support	Ease of Online booking	On-board service
Leg room service	Baggage handling	Checkin service	Cleanliness
Online boarding	Departure Delay in Minutes	Arrival Delay in Minutes	

The data was included in the source in 2020 and the information was anonymously given. The name for the airline used for this example is Invistico airlines. The rows represent the individual customer feedback, and the column names are the measures that capture the

customer related information as well as their experience of travelling with the company including how they have rated several airplane features. The columns also mention details about the flight that the customer has taken, as well as features of the particular airplane.

The **second dataset** has 3501 records and has 37 variables that were measured per customer as detailed below:

VARIABLES					
Quarter	Availability of baggage carts	Courtesy of security staff	Walking distance inside terminal	Shopping facilities	Customs inspection
Date recorded	Efficiency of check-in staff	Thoroughness of security inspection	Ease of making connections	Shopping facilities (value for money)	Overall satisfaction
Departure time	Check-in wait time	Wait time of security inspection	Courtesy of airport staff	Comfort of waiting/gate areas	
Ground transportation to/from airport	Courtesy of of check-in staff	Feeling of safety and security	Internet access	Cleanliness of airport terminal	
Parking facilities	Wait time at passport inspection	Restaurants	Business/executive lounges	Ambience of airport	
Parking facilities (value for money)	Ease of finding your way through the airport	Restaurants (value for money)	Availability of washrooms	Arrivals passport and visa inspection	
Courtesy of inspection staff	Flight information screens	Availability of banks/ATM/money changing	Cleanliness of washrooms	Speed of baggage delivery	

The data was first created for use in 2020. The information comes from customer surveys in Austin-Bergstrom airport. The surveys were conducted from 2015-2017 and are reported per quarters. The rows represent individual customer feedback and the columns detail the users overall satisfaction in many variables such as restaurants for example.

## Initial idea of what to do with the data

### First dataset:

- One business question we would be interested in is if a customer would be satisfied or dissatisfied in their trip taking into account their attributes and the attributes of the flight
- Another business question we would like to address is which type of users do we have. We would like to cluster our users by the type of customer they are and other features such as satisfaction to see if we can fit them into different groups depending on their characteristics and addressing them differently in the future

### Second dataset:

- One business question we are interested in is which are the features that influence the most our users satisfaction so we may work on those which are the most relevant to achieving a greater customer satisfaction

## Reading and cleaning your datasets

## First dataset

### Task a

As we can see in the image below, these are the number of missing values per column

satisfaction	0
Gender	0
Customer Type	0
Age	0
Type of Travel	0
Class	0
Flight Distance	0
Seat comfort	0
Departure/Arrival time convenient	0
Food and drink	0
Gate location	0
Inflight wifi service	0
Inflight entertainment	0
Online support	0
Ease of Online booking	0
On-board service	0
Leg room service	0
Baggage handling	0
Checkin service	0
Cleanliness	0
Online boarding	0
Departure Delay in Minutes	0
Arrival Delay in Minutes	393
dtype: int64	

Given that the dataset is quite big and 393 ends up only being 0.3% of the total rows we have decided to drop the rows with this missing values.

### Task b

The following image shows the count per class for the satisfaction variable

satisfied	70882
dissatisfied	58605

Task c

#	Column	Non-Null Count	Dtype
0	satisfaction	129487 non-null	object
1	Gender	129487 non-null	object
2	Customer Type	129487 non-null	object
3	Age	129487 non-null	int64
4	Type of Travel	129487 non-null	object
5	Class	129487 non-null	object
6	Flight Distance	129487 non-null	int64
7	Seat comfort	129487 non-null	int64
8	Departure/Arrival time convenient	129487 non-null	int64
9	Food and drink	129487 non-null	int64
10	Gate location	129487 non-null	int64
11	Inflight wifi service	129487 non-null	int64
12	Inflight entertainment	129487 non-null	int64
13	Online support	129487 non-null	int64
14	Ease of Online booking	129487 non-null	int64
15	On-board service	129487 non-null	int64
16	Leg room service	129487 non-null	int64
17	Baggage handling	129487 non-null	int64
18	Checkin service	129487 non-null	int64
19	Cleanliness	129487 non-null	int64
20	Online boarding	129487 non-null	int64
21	Departure Delay in Minutes	129487 non-null	int64
22	Arrival Delay in Minutes	129487 non-null	float64

dtypes: float64(1), int64(17), object(5)

After validating the data types, they are all the right type of variable. We have classes for our categorical values and ordinal values for our ratings.

## Second dataset

Task a

Below we can see the number of missing values per column we have for the airport customer satisfaction dataset

Quarter	0
Date recorded	0
Departure time	0
Ground transportation to/from airport	54
Parking facilities	39
Parking facilities (value for money)	46
Availability of baggage carts	91
Efficiency of check-in staff	38
Check-in wait time	39
Courtesy of of check-in staff	52
Wait time at passport inspection	69
Courtesy of inspection staff	96
Courtesy of security staff	31
Thoroughness of security inspection	46
Wait time of security inspection	50
Feeling of safety and security	43
Ease of finding your way through the airport	36
Flight information screens	26
Walking distance inside terminal	37
Ease of making connections	83
Courtesy of airport staff	40
Restaurants	59
Restaurants (value for money)	60
Availability of banks/ATM/money changing	41
Shopping facilities	46
Shopping facilities (value for money)	57
Internet access	73
Business/executive lounges	91
Availability of washrooms	35
Cleanliness of washrooms	37
Comfort of waiting/gate areas	41
Cleanliness of airport terminal	32
Ambience of airport	54
Arrivals passport and visa inspection	143
Speed of baggage delivery	181
Customs inspection	201
Overall satisfaction	172

Since we want to use the Overall satisfaction column as our classifier, we will drop the 172 rows which don't have a value since they only make up a 5% of the data.

On the other hand we will impute the rest of the columns with the mode (given that as seen from the histograms, the data is very skewed) to make up for the missing values in each of the columns. We wish to keep them so as to not drop any more information that will come in handy for the classification task.

### Task b

The following shows the count per label. It is important to note that users that gave a rating of 4 or 5 are classified as satisfied; all the other have been classified as not satisfied

```
not satisfied    2088
satisfied        1241
```

### Task c

#	Column	Non-Null Count	Dtype
0	Quarter	3329 non-null	object
1	Date recorded	3329 non-null	object
2	Departure time	3329 non-null	object
3	Ground transportation to/from airport	3329 non-null	float64
4	Parking facilities	3329 non-null	float64
5	Parking facilities (value for money)	3329 non-null	float64
6	Availability of baggage carts	3329 non-null	float64
7	Efficiency of check-in staff	3329 non-null	float64
8	Check-in wait time	3329 non-null	float64
9	Courtesy of of check-in staff	3329 non-null	float64
10	Wait time at passport inspection	3329 non-null	float64
11	Courtesy of inspection staff	3329 non-null	float64
12	Courtesy of security staff	3329 non-null	float64
13	Thoroughness of security inspection	3329 non-null	float64
14	Wait time of security inspection	3329 non-null	float64
15	Feeling of safety and security	3329 non-null	float64
16	Ease of finding your way through the airport	3329 non-null	float64
17	Flight information screens	3329 non-null	float64
18	Walking distance inside terminal	3329 non-null	float64
19	Ease of making connections	3329 non-null	float64
20	Courtesy of airport staff	3329 non-null	float64
21	Restaurants	3329 non-null	float64
22	Restaurants (value for money)	3329 non-null	float64
23	Availability of banks/ATM/money changing	3329 non-null	float64
24	Shopping facilities	3329 non-null	float64
25	Shopping facilities (value for money)	3329 non-null	float64
26	Internet access	3329 non-null	float64
27	Business/executive lounges	3329 non-null	float64
28	Availability of washrooms	3329 non-null	float64
29	Cleanliness of washrooms	3329 non-null	float64
30	Comfort of waiting/gate areas	3329 non-null	float64
31	Cleanliness of airport terminal	3329 non-null	float64
32	Ambience of airport	3329 non-null	float64
33	Arrivals passport and visa inspection	3329 non-null	float64
34	Speed of baggage delivery	3329 non-null	float64
35	Customs inspection	3329 non-null	float64
36	Overall satisfaction	3329 non-null	float64
37	satisfied	3329 non-null	object

After observing all the variables we see that they have the correct type needed for our classification

## Decision on keeping the datasets separate

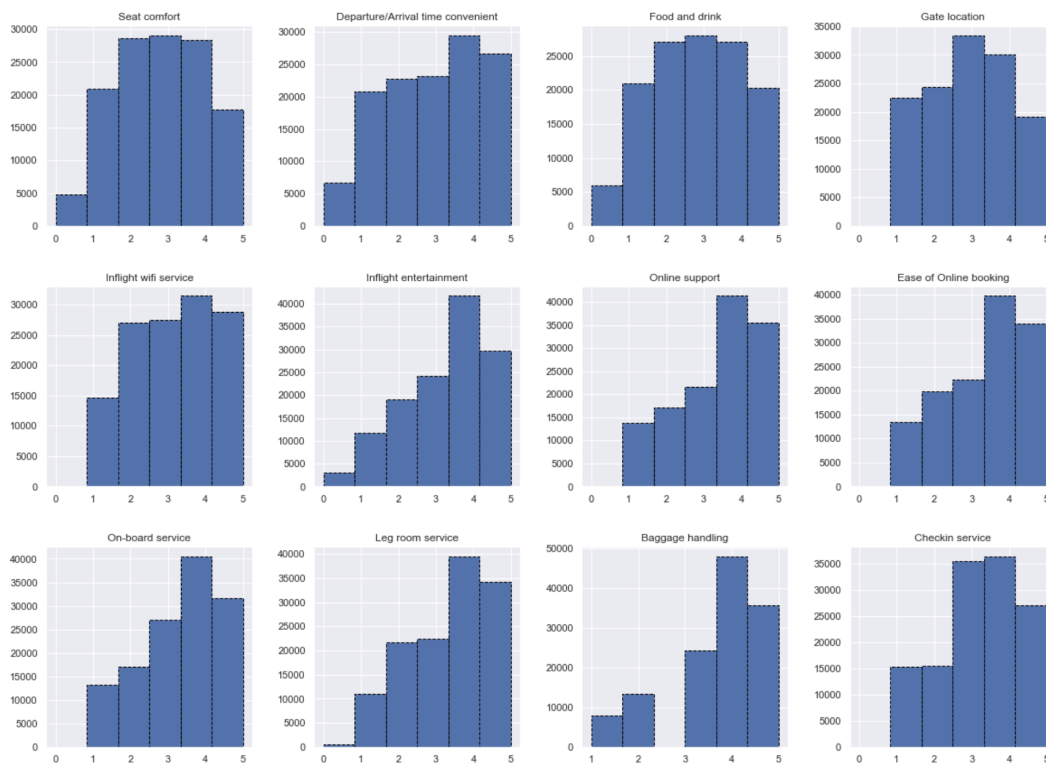
We have decided to keep both the datasets separate as the questions answered from each data set can be answered by using the data in each separately. Both have different numbers of rows and columns, which will make it harder for the information to be merged. For data set one, we will predict whether the customer will take another flight with the airline or not as well as clustering the customer into different groups. For the second data set, we will mainly deal with the attributes that can have an impact on customer satisfaction using binary classification. For this reason, merging the datasets is not required and it is better to treat them separately.

## Data Visualization

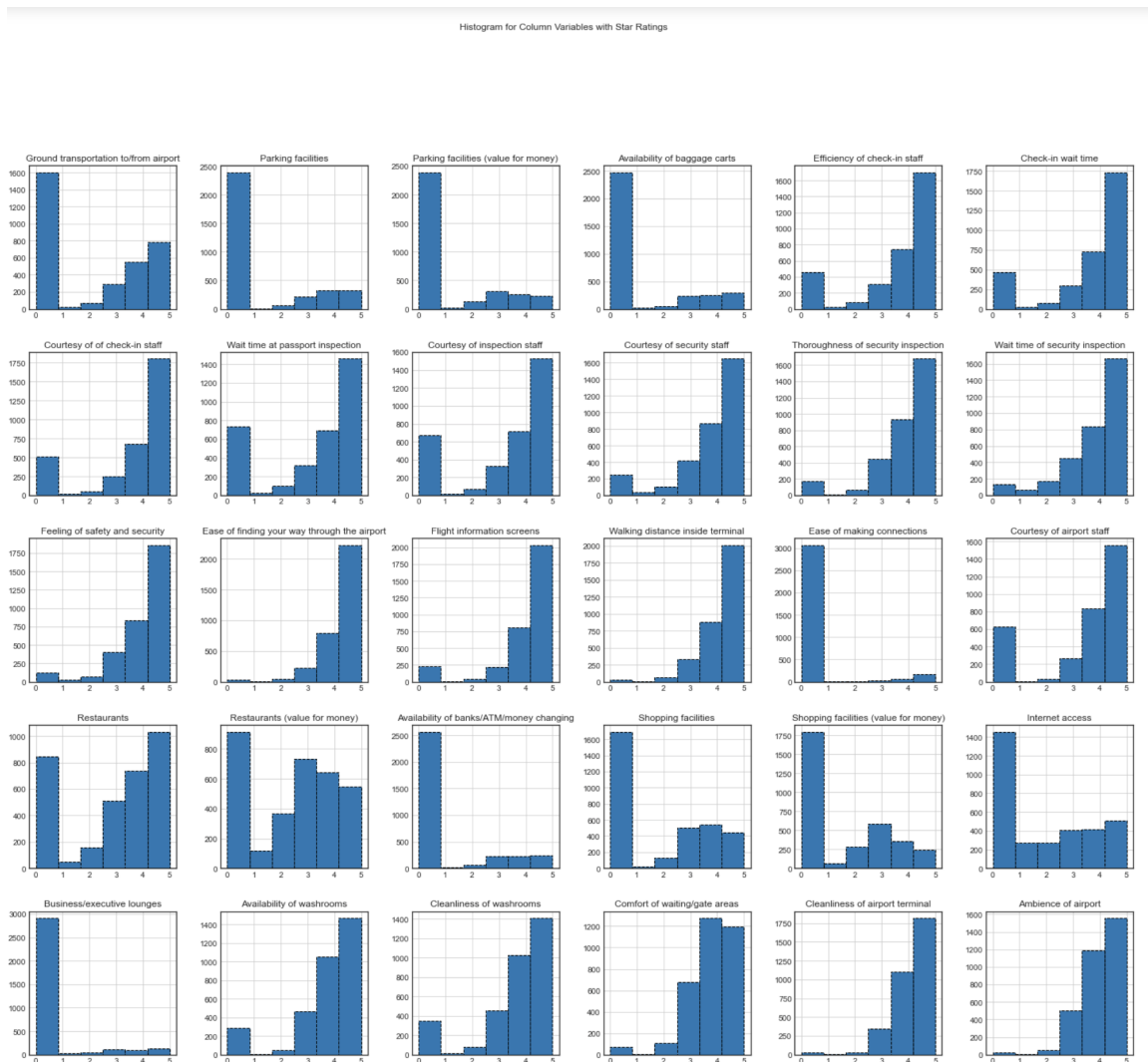
For both the data sets, we made a histogram and a regplot as that goes with our problem statements. A histogram best explains that for the attributes in both data sets with ratings given from 0 to 5, how much is the count of each star in each attribute. This also gives an idea which attributes were ranked higher in stars than the others and which attributes did poorly. The histogram of these attributes are below:

### Data set 1

Histogram for Column Variables with Star Ratings



## Data set 2:



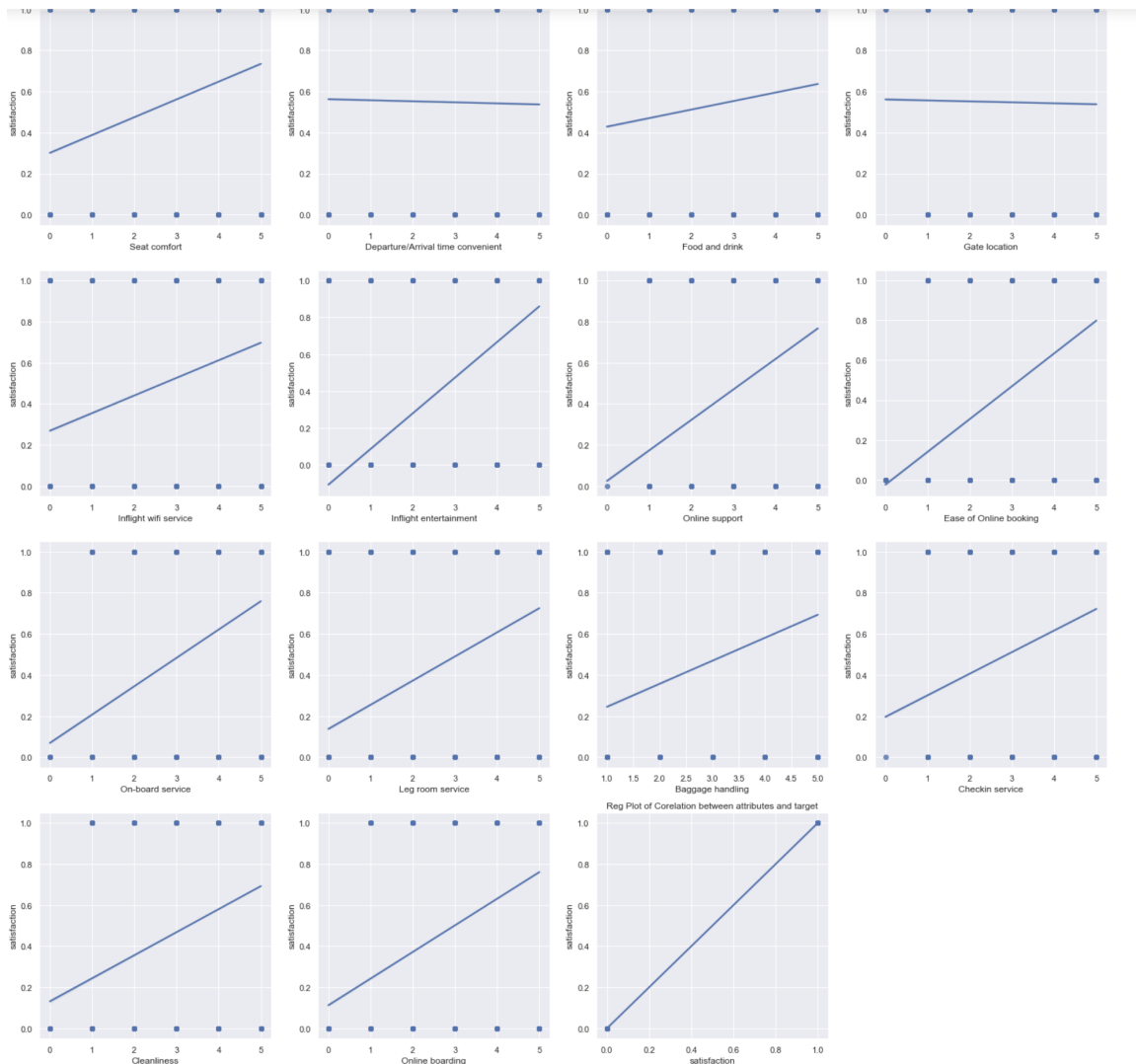
Secondly, we calculated the correlation score of these attributes with satisfaction and plotted a regplot to visualize this information as well as to see what attributes have most impacted the satisfaction score and what type of relationship do they share with the target variable.

## Data set 1:



Seat comfort	0.242371
Departure/Arrival time convenient	-0.015624
Food and drink	0.120568
Gate location	-0.012272
Inflight wifi service	0.227010
Inflight entertainment	0.523364
Online support	0.389890
Ease of Online booking	0.432017
On-board service	0.352283
Leg room service	0.305115
Baggage handling	0.260398
Checkin service	0.266089
Cleanliness	0.259504
Online boarding	0.338118

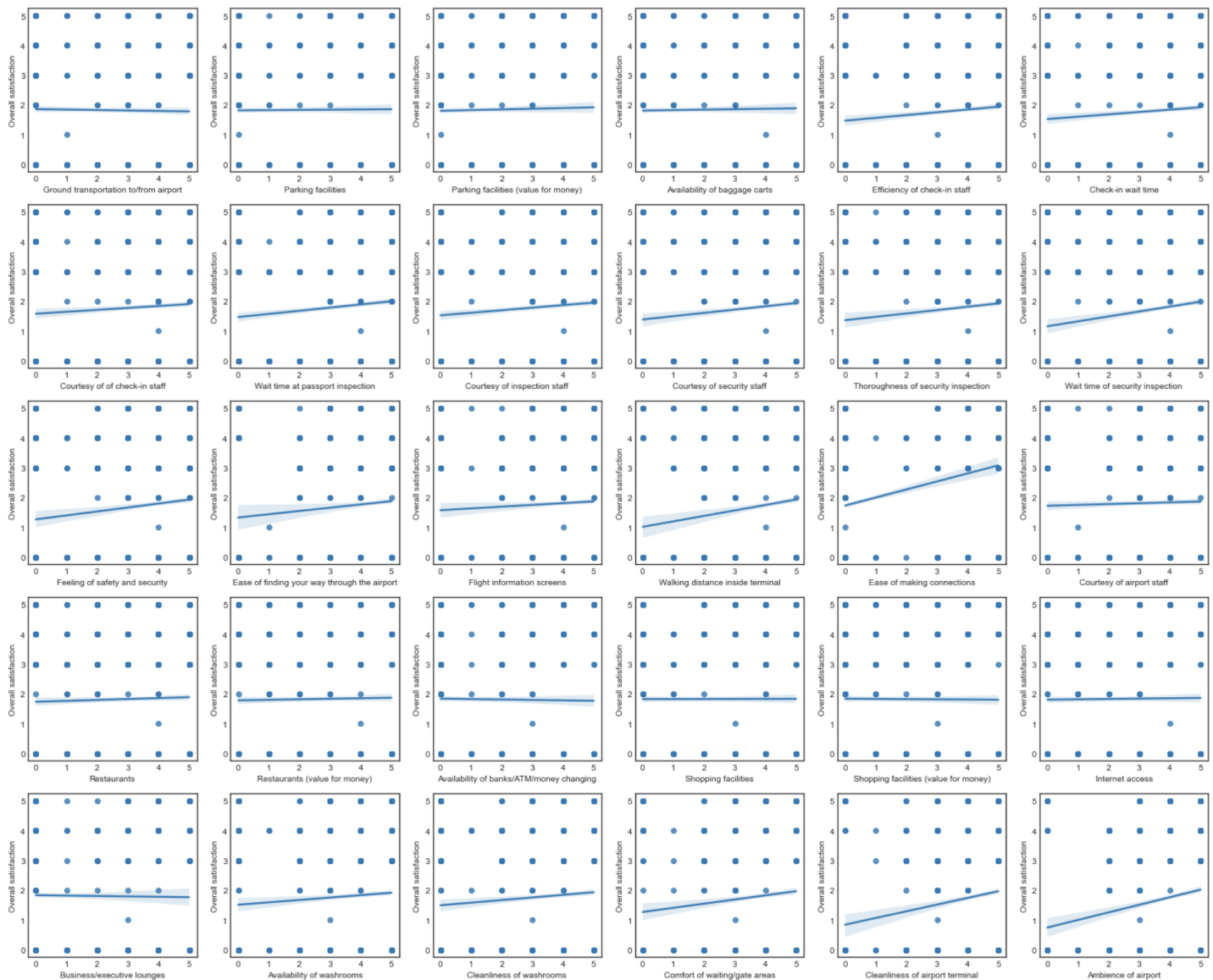
dtype: float64



Data set 2:

Ground transportation to/from airport	-0.014919
Parking facilities	0.006306
Parking facilities (value for money)	0.017704
Availability of baggage carts	0.012070
Efficiency of check-in staff	0.072653
Check-in wait time	0.061926
Courtesy of of check-in staff	0.052286
Wait time at passport inspection	0.094178
Courtesy of inspection staff	0.074165
Courtesy of security staff	0.071371
Thoroughness of security inspection	0.064964
Wait time of security inspection	0.096904
Feeling of safety and security	0.071428
Ease of finding your way through the airport	0.042013
Flight information screens	0.035525
Walking distance inside terminal	0.074720
Ease of making connections	0.150044
Courtesy of airport staff	0.023830
Restaurants	0.027231
Restaurants (value for money)	0.014550
Availability of banks/ATM/money changing	-0.010806
Shopping facilities	0.001176
Shopping facilities (value for money)	-0.005498
Internet access	0.010482
Business/executive lounges	-0.008013
Availability of washrooms	0.052074
Cleanliness of washrooms	0.060536
Comfort of waiting/gate areas	0.064303
Cleanliness of airport terminal	0.084277
Ambience of airport	0.101082
Arrivals passport and visa inspection	-0.877192
Speed of baggage delivery	0.597898
Customs inspection	-0.517455

dtype: float64



## Define your classification or prediction variables

In the first data set, the classification variable is the first column with the name 'satisfaction'. This column tells whether each customer was satisfied or not with the airline based on the attributes mentioned in the first dataset.

For the second dataset, our classification variable is the column 'Overall satisfaction' where we have two labels; satisfied and non satisfied. This column initially had rating values from 0 to 5. As mentioned above, users that gave a rating of 4 or 5 are classified as satisfied; all the others have been classified as not satisfied.

