



CAPSTONE PROJECT

# ML PROJECT

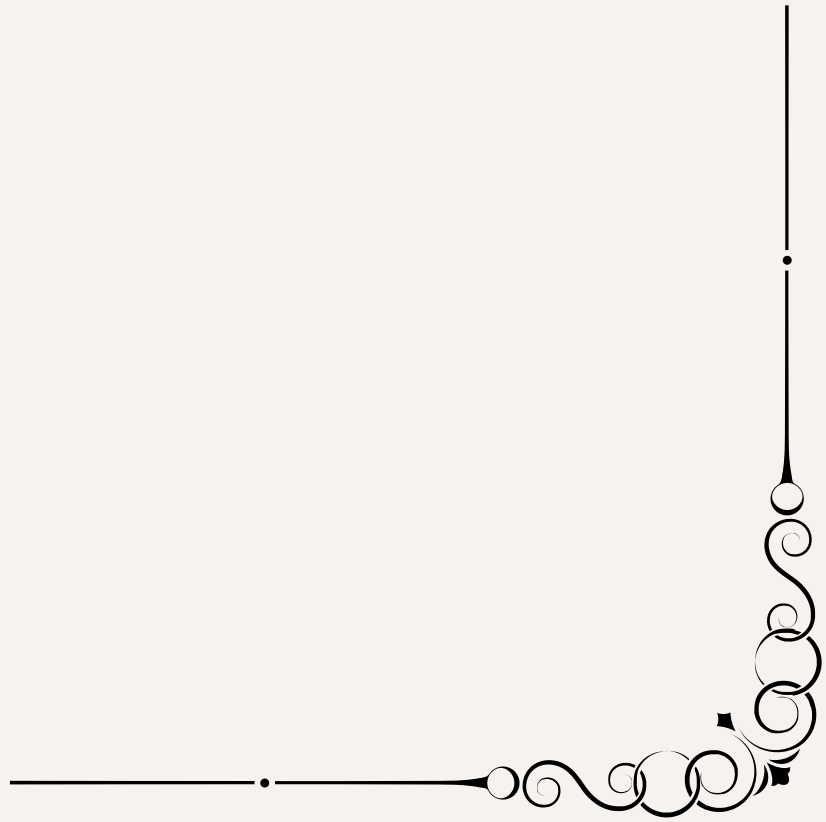
SAMSUNG INNOVATION CENTER

BY  
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# Agenda

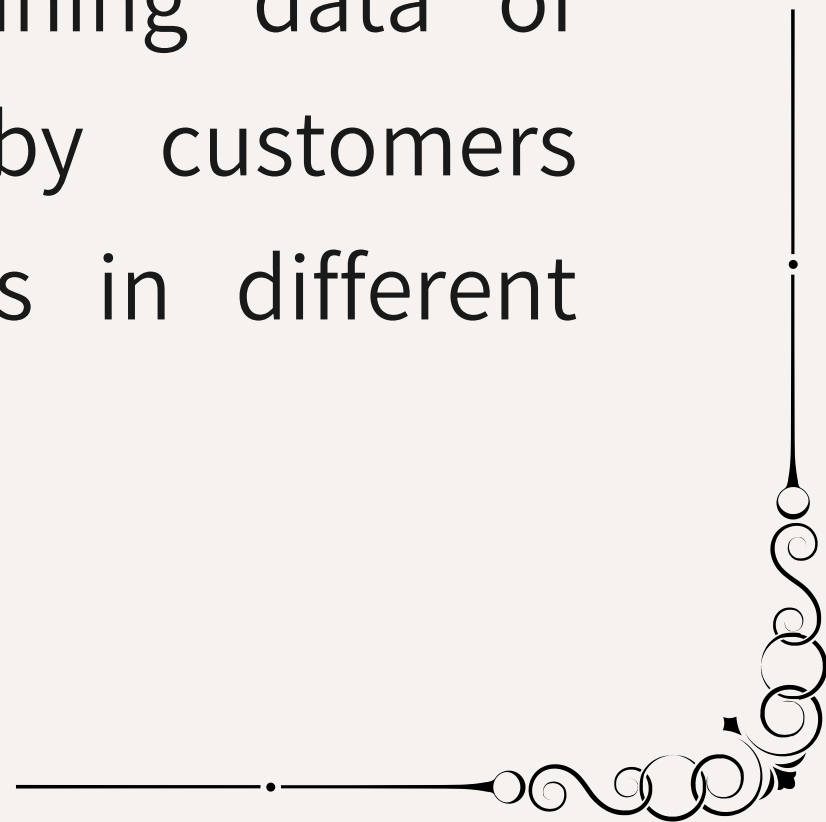
- 1 Problem Statement
  - 2 Data Given in a Dataset
  - 3 Visualization with inference
  - 4 Steps done
  - 5 Chosen Model
  - 6 Conclusion
- 



# Problem Statement

Hotel Reservation:

A machine learning model to predict whether the customer cancels there hotel reservation or not . By using given dataset containing data of reservations made by customers from different places in different hotels.





# Data given in Dataset

Given a Dataset Hotel\_Bookings.csv contains 119390 rows  $\times$  32 columns which describes all the features of the hotel and booking details including their Arrival Timings, booking stats from which country they are from, through which agent they have booked etc..

DATASET USED:

[https://raw.githubusercontent.com/Premalatha-success/Datasets/main/hotel\\_bookings.csv](https://raw.githubusercontent.com/Premalatha-success/Datasets/main/hotel_bookings.csv)




# Datatypes and DataShape in Data

```
[56] 1 #Explore the data-shape
      2 data.shape

(119390, 32)
```

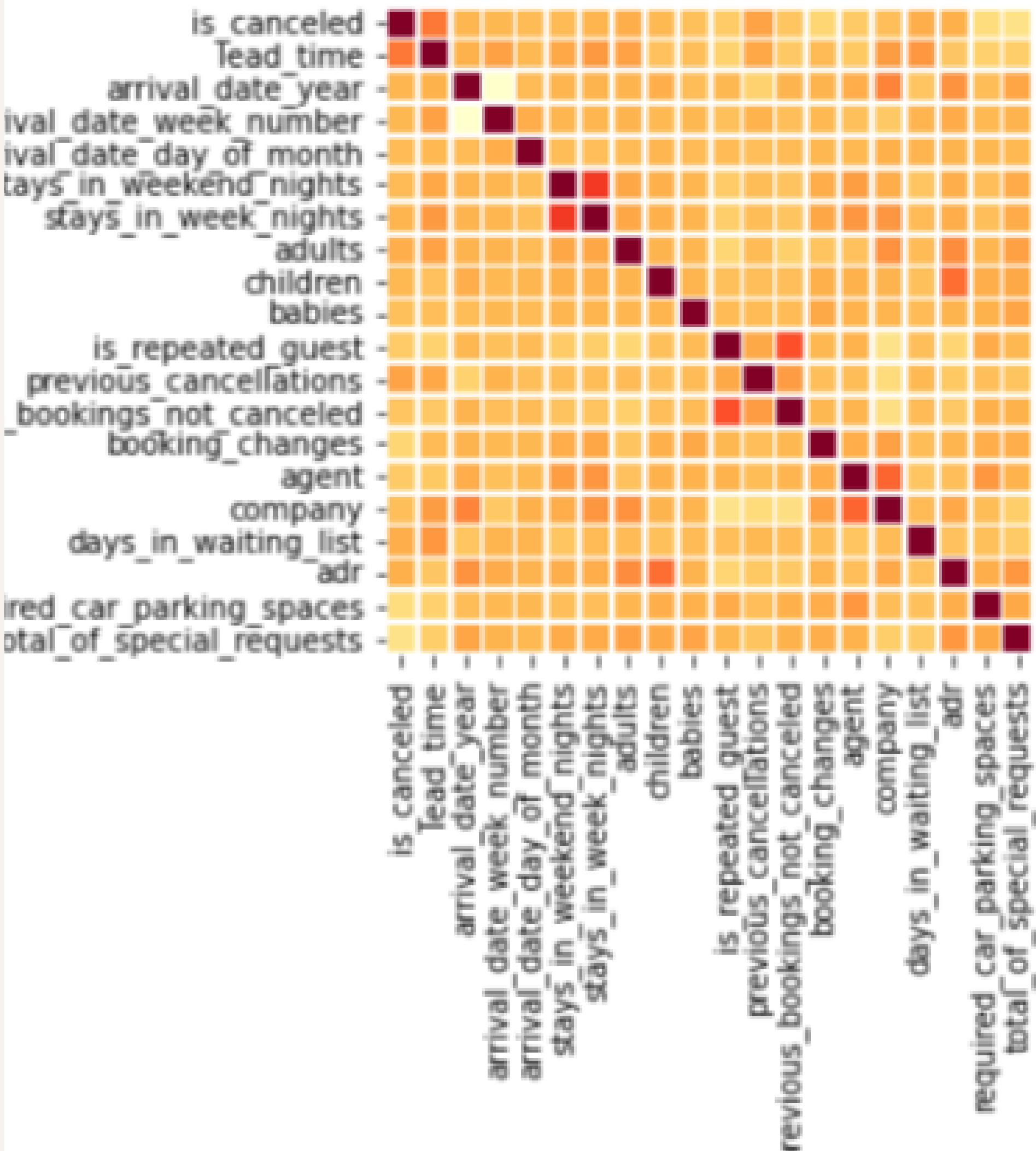
```
hotel          object
is_canceled    int64
lead_time      int64
arrival_date_year  int64
arrival_date_month  object
arrival_date_week_number  int64
arrival_date_day_of_month  int64
stays_in_weekend_nights  int64
stays_in_week_nights  int64
adults         int64
children       float64
babies         int64
meal           object
country        object
market_segment object
distribution_channel  object
is_repeated_guest  int64
previous_cancellations  int64
previous_bookings_not_canceled  int64
reserved_room_type  object
assigned_room_type  object
booking_changes  int64
deposit_type    object
agent          float64
company        float64
days_in_waiting_list  int64
customer_type   object
adr            float64
required_car_parking_spaces  int64
total_of_special_requests  int64
reservation_status  object
reservation_status_date  object
dtype: object
```



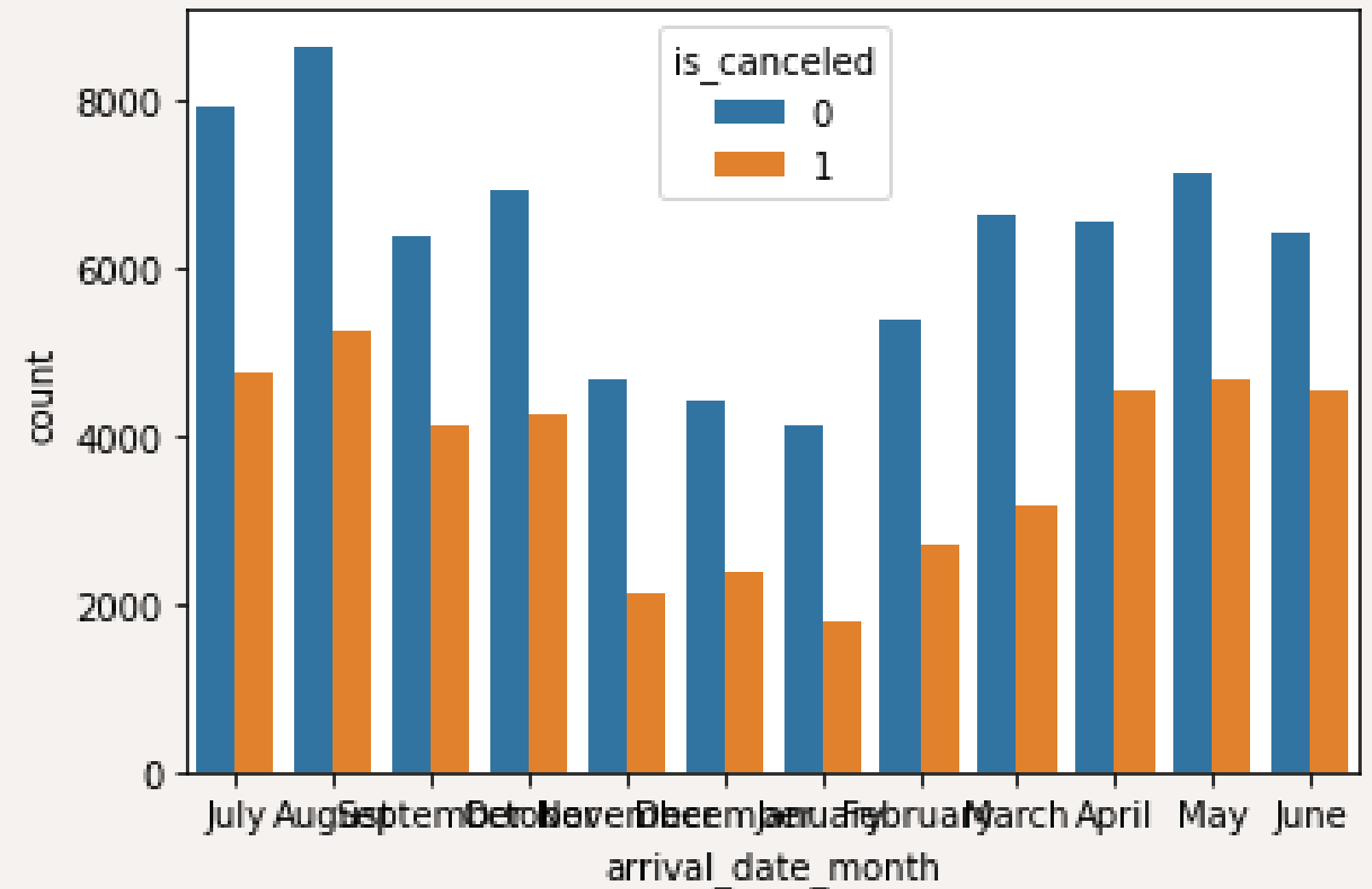
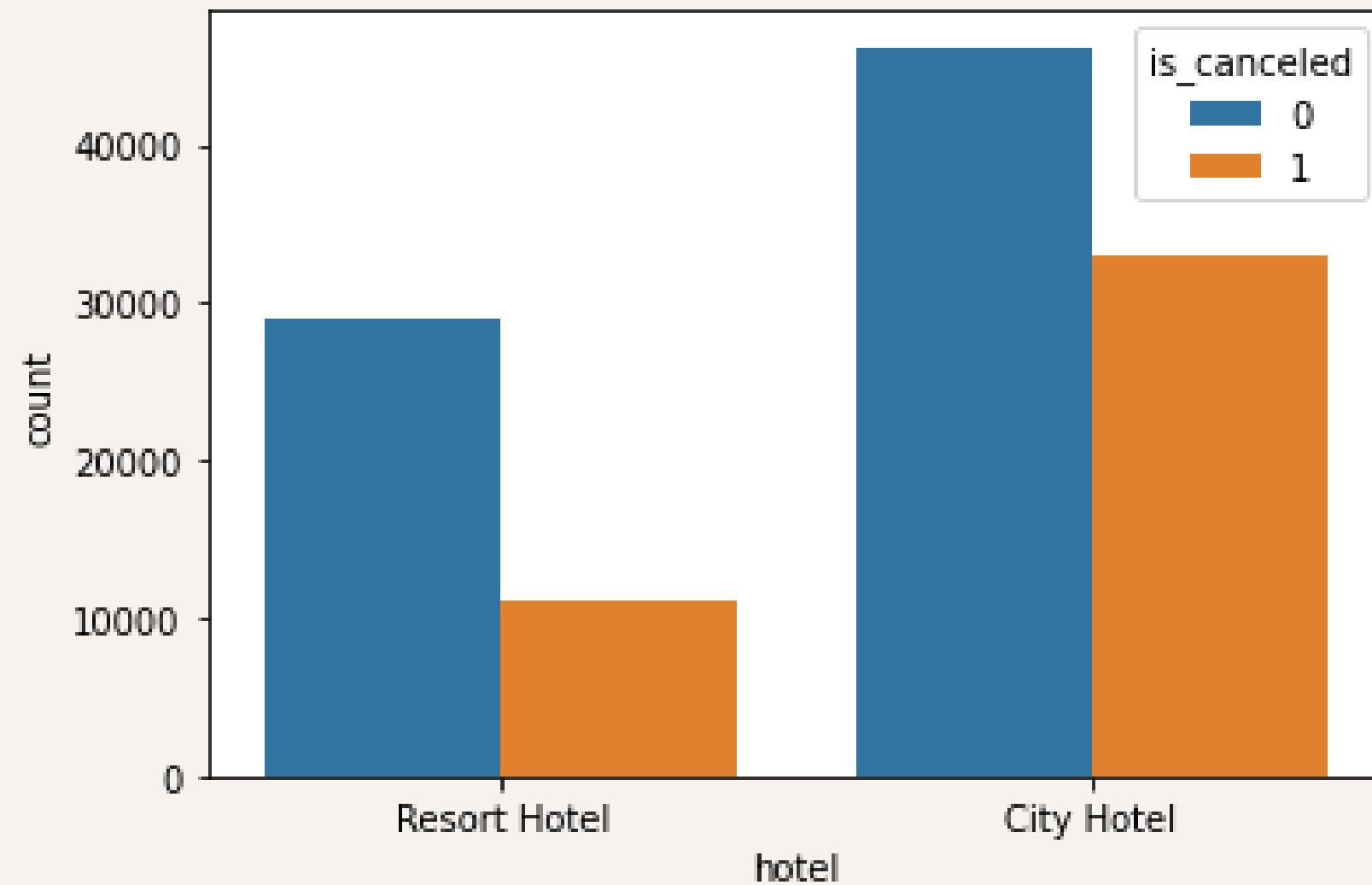
# Visualiz

# Correlation Visualization

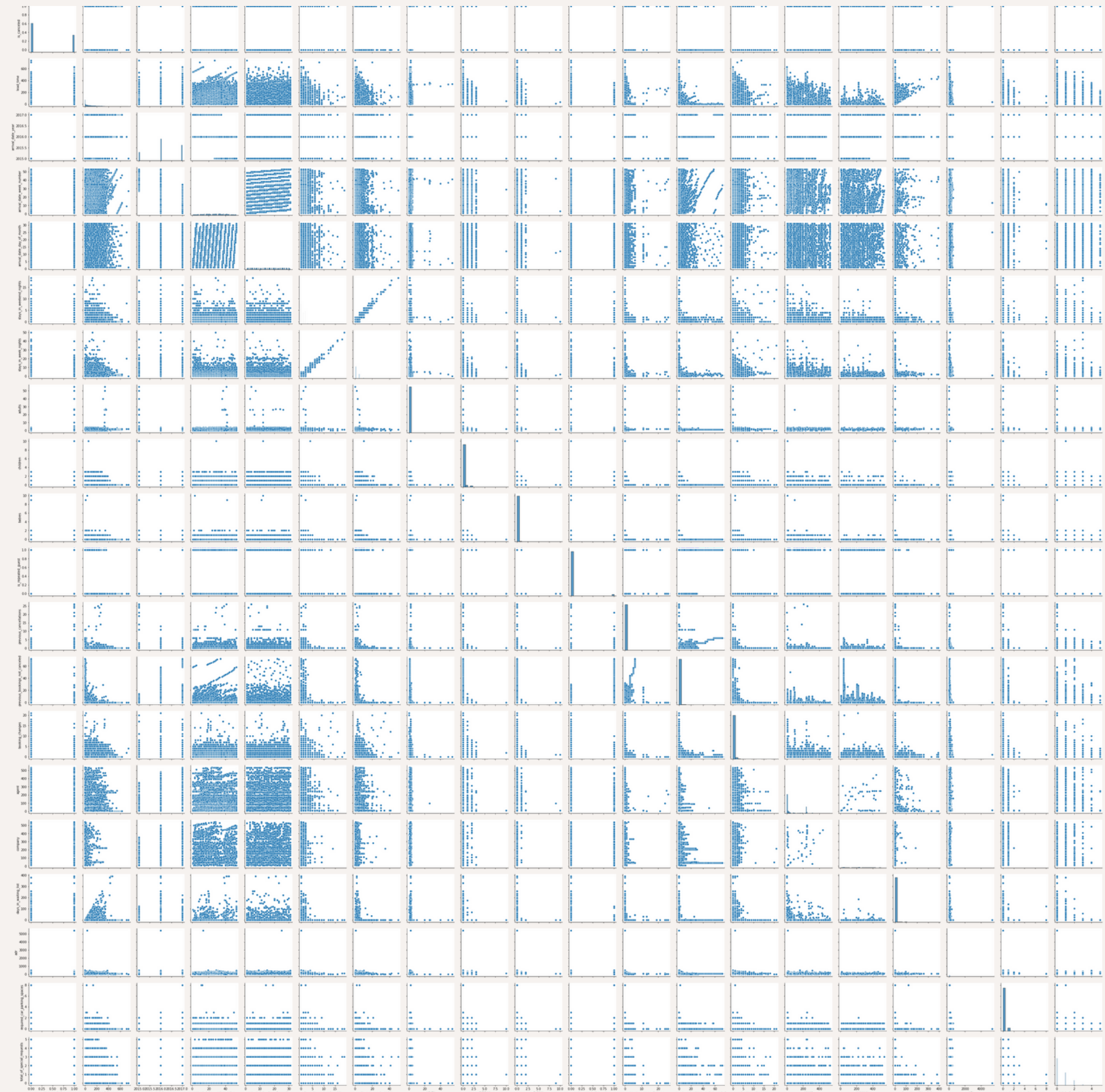
Correlation visualization summarizes the association between two variables. It will be ranging from -1 to +1



# Visualizing from feature to feature

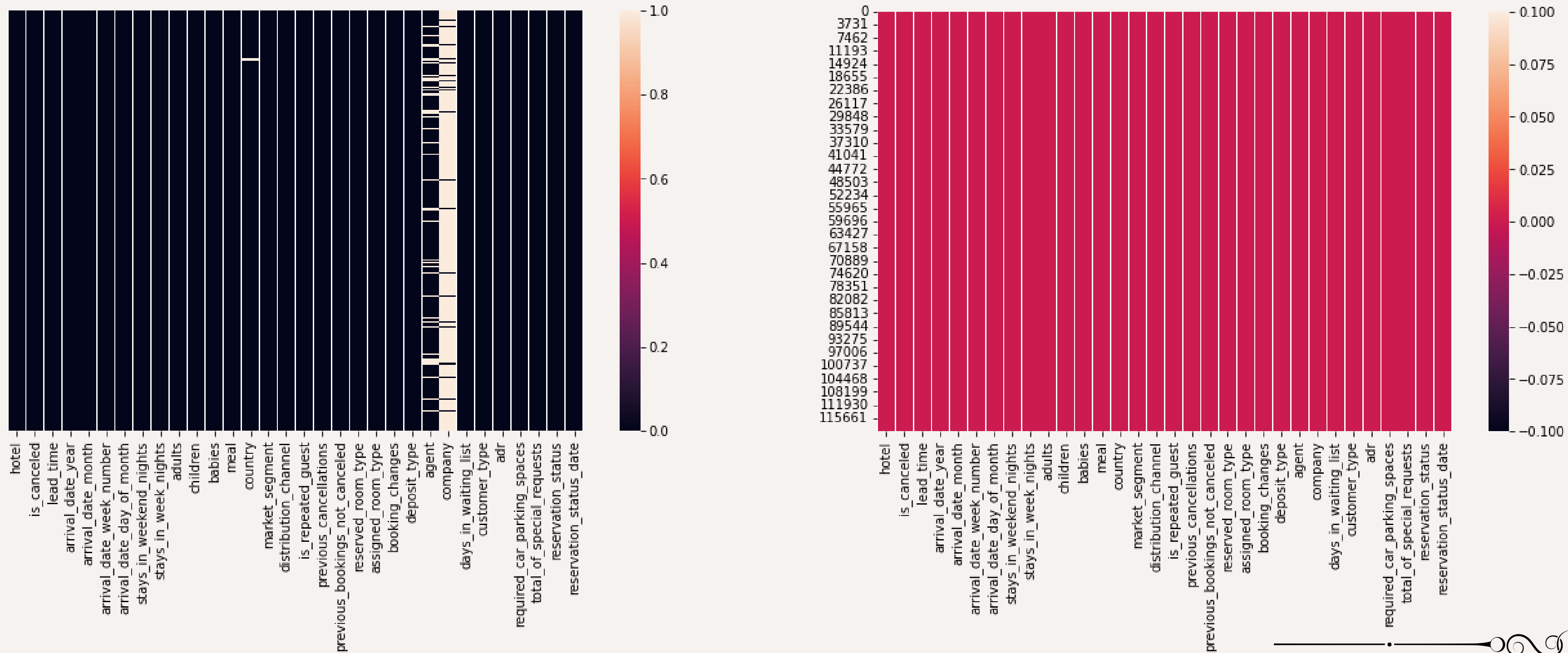


# Visualizing pairplot



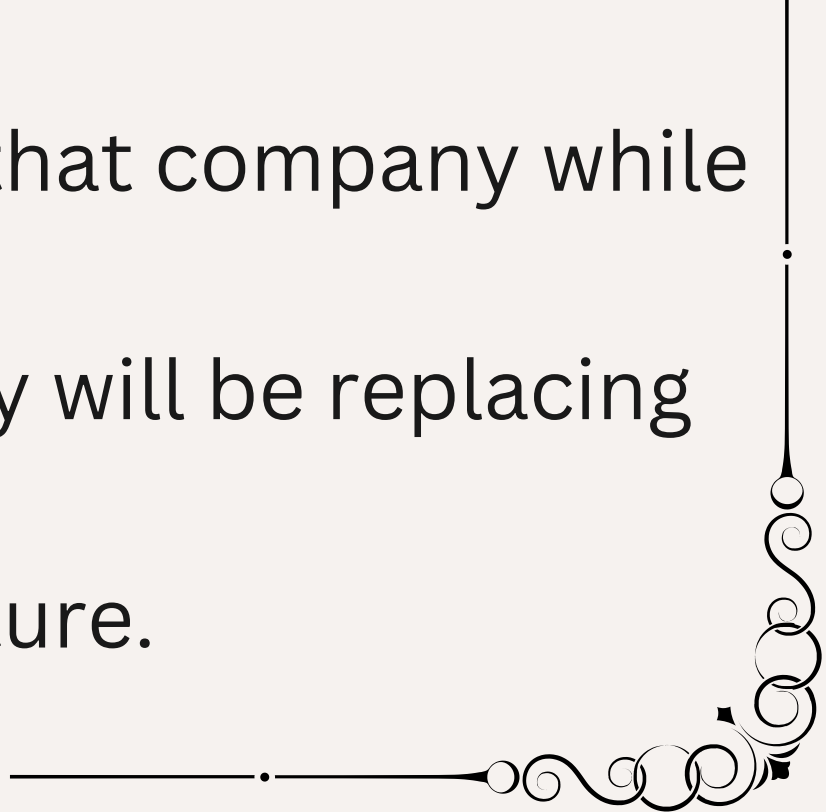


# Visualizing null values before and after





# Inference on the data

- Correlated all the features of the dataset in order to get the relationship between the features.
  - Then we have checked the null-values of all features and we had observed that the feature named 'company' has high null values. So, we can drop that company while training the data from the dataset.
  - For the remaining features which consists of null values, they will be replacing with median/mode/mean of that remaining non-null data in that feature.
- 



# Steps done

1) imported required libraries

- Basic and most important libraries
- Model evaluation tools
- Data processing functions

2) Importing dataset

- Exploring the data shape
- Exploring the Datatypes

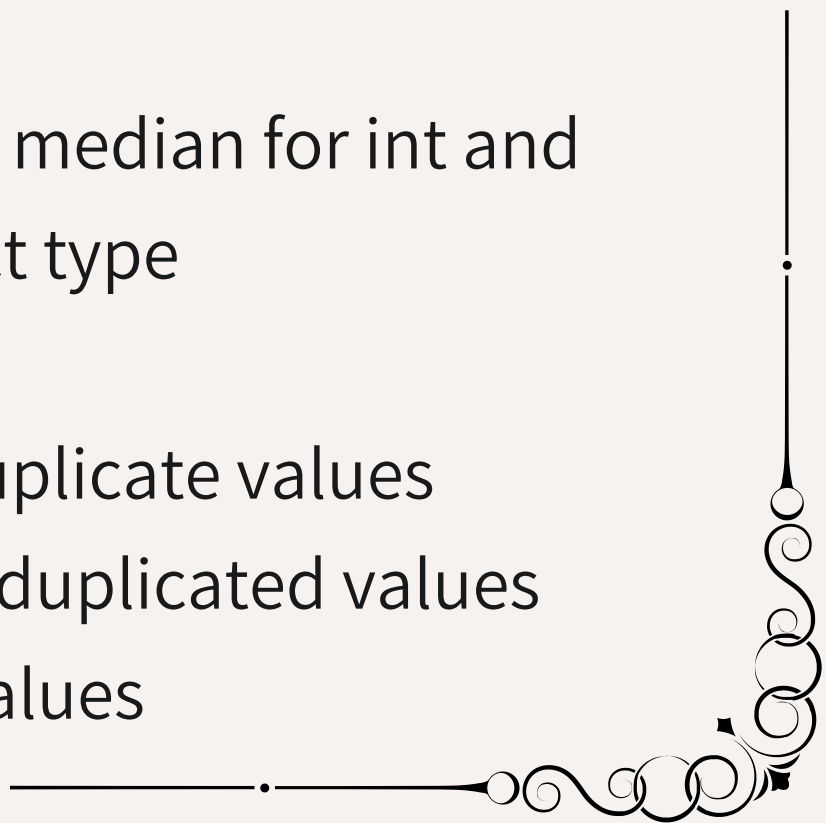
3) Correlation between all the Features

4) Describe Statistical Summary

5) Check for Null Values

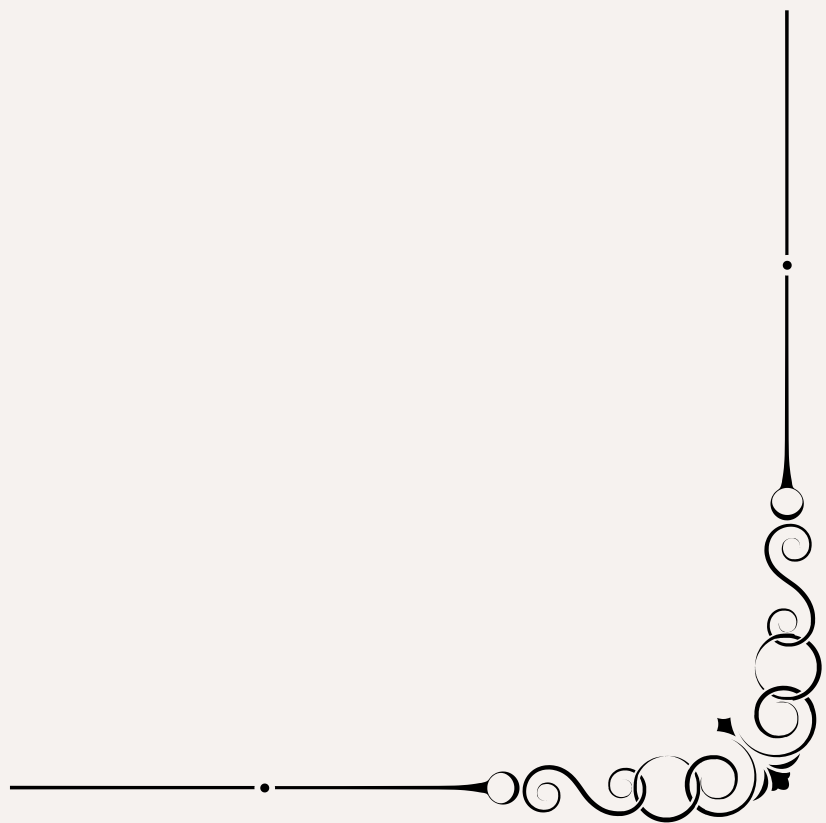
- Replace Null values with median for int and float and mode for object type
- verifying the null values

) identifying and removing duplicate values

- Identify total number of duplicated values
  - Dropping all duplicate values
- 



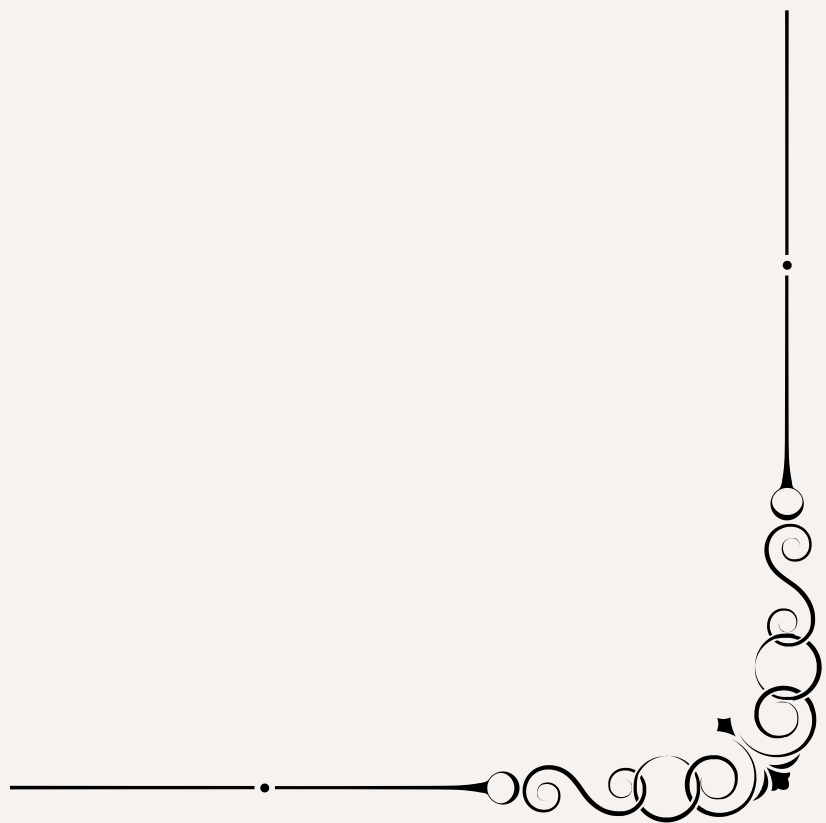
# Steps done

- Verifying duplicates are any
- 7)Encoding: Used label encoder
- 8)Evaluating a classification model
- Dividing data into Input X variables and Target Y variable.
  - Y with only 'is canceled' feature and X with the remaining features.
- 9)Applied following Algorithms to find best model
- logistic regression
  - KNN
  - SVM(Linear kernel)
  - Naive Bayes
  - Decision Tree
  - Bagging Classifier
- 



# Steps done

10) Classification report and confusion matrix for every model

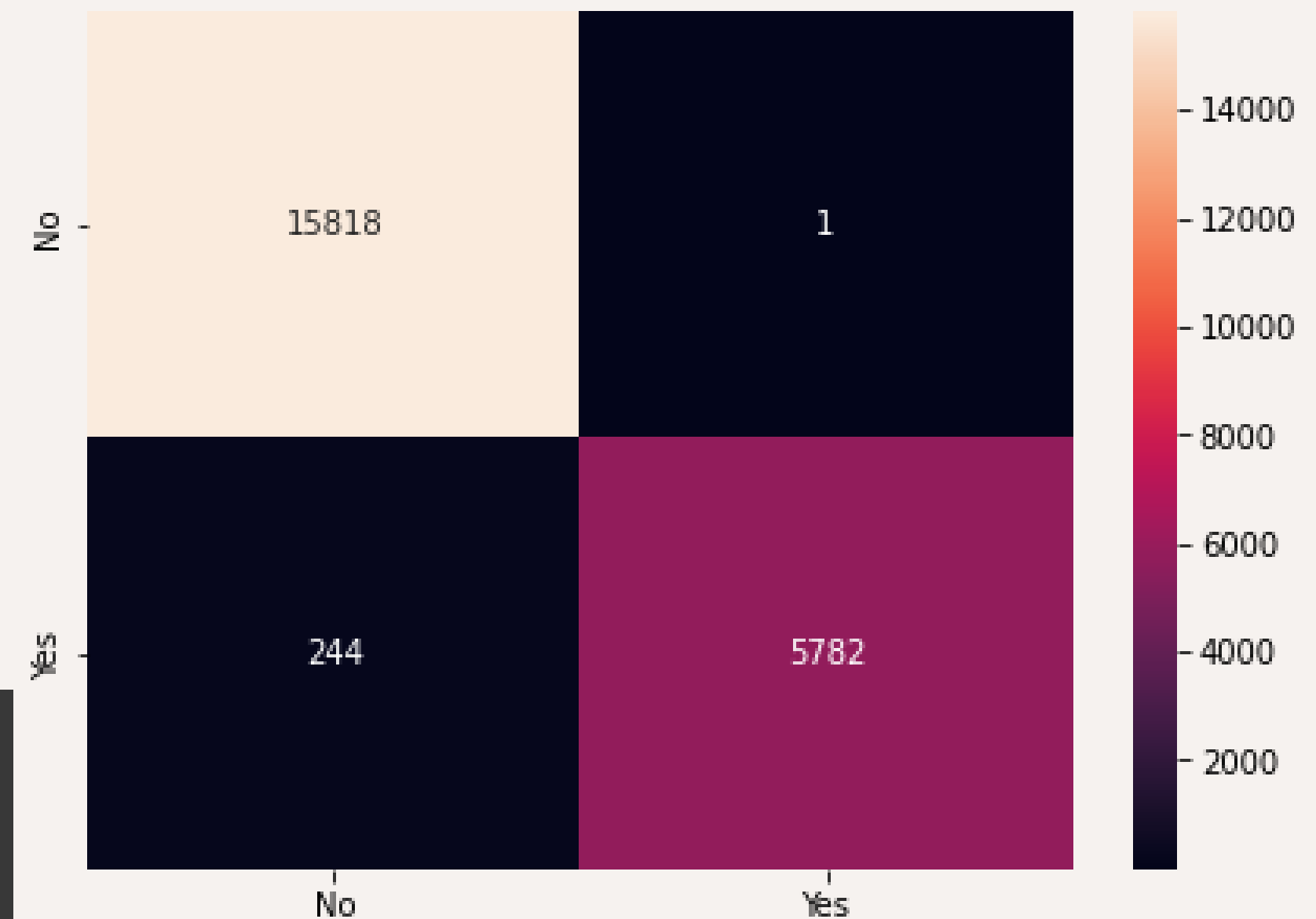


# Evaluating Models

Logistic Regression

Classification Report after training:

Classification Report				
	precision	recall	f1-score	support
1	1.00	0.96	0.98	6026
0	0.98	1.00	0.99	15819
accuracy			0.99	21845
macro avg	0.99	0.98	0.99	21845
weighted avg	0.99	0.99	0.99	21845



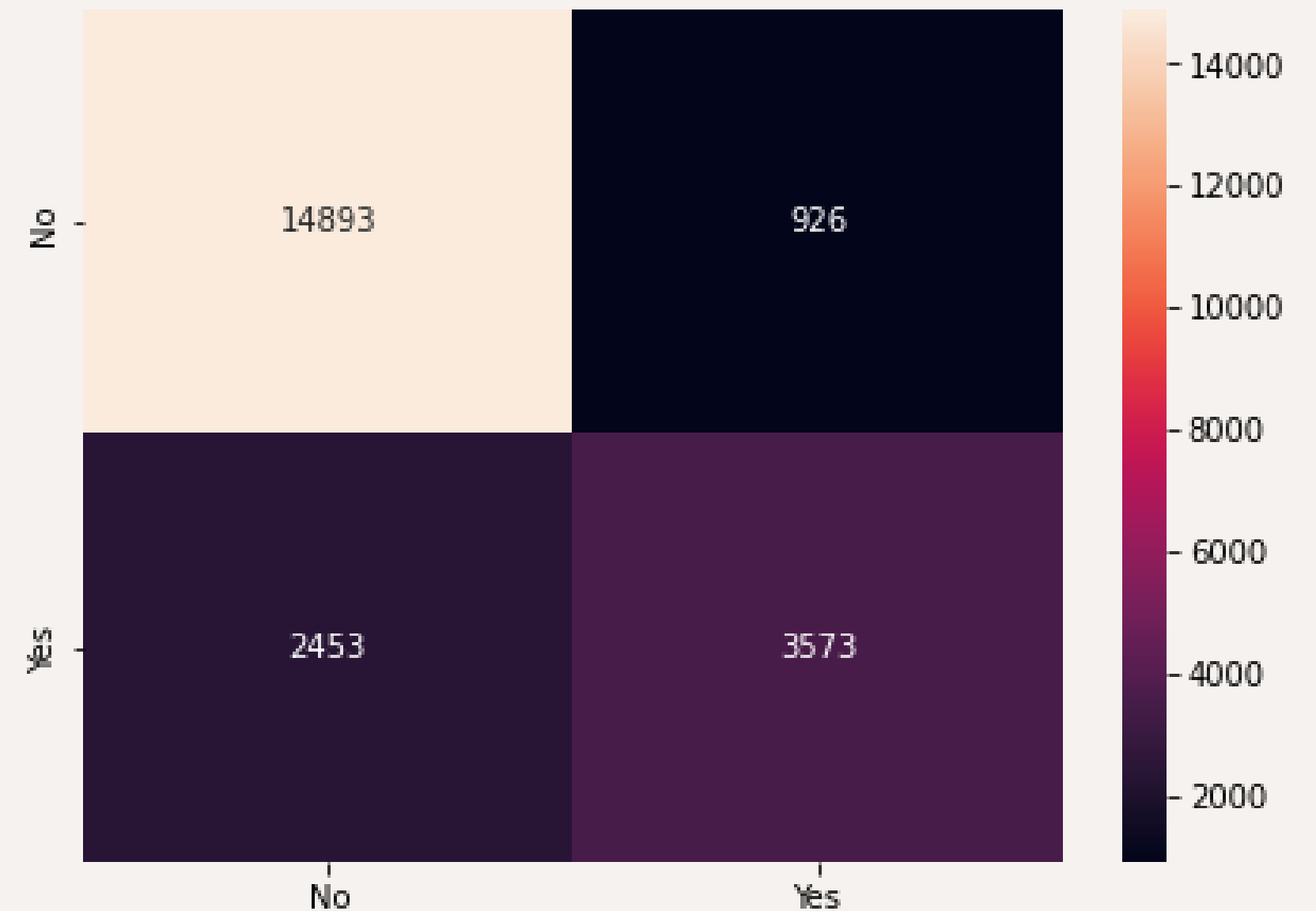
Confusion matrix  
for  
Logistic Regression

# Evaluating Models

KNN

Classification Report after training:

Classification Report				
	precision	recall	f1-score	support
1	0.79	0.59	0.68	6026
0	0.86	0.94	0.90	15819
accuracy			0.85	21845
macro avg	0.83	0.77	0.79	21845
weighted avg	0.84	0.85	0.84	21845



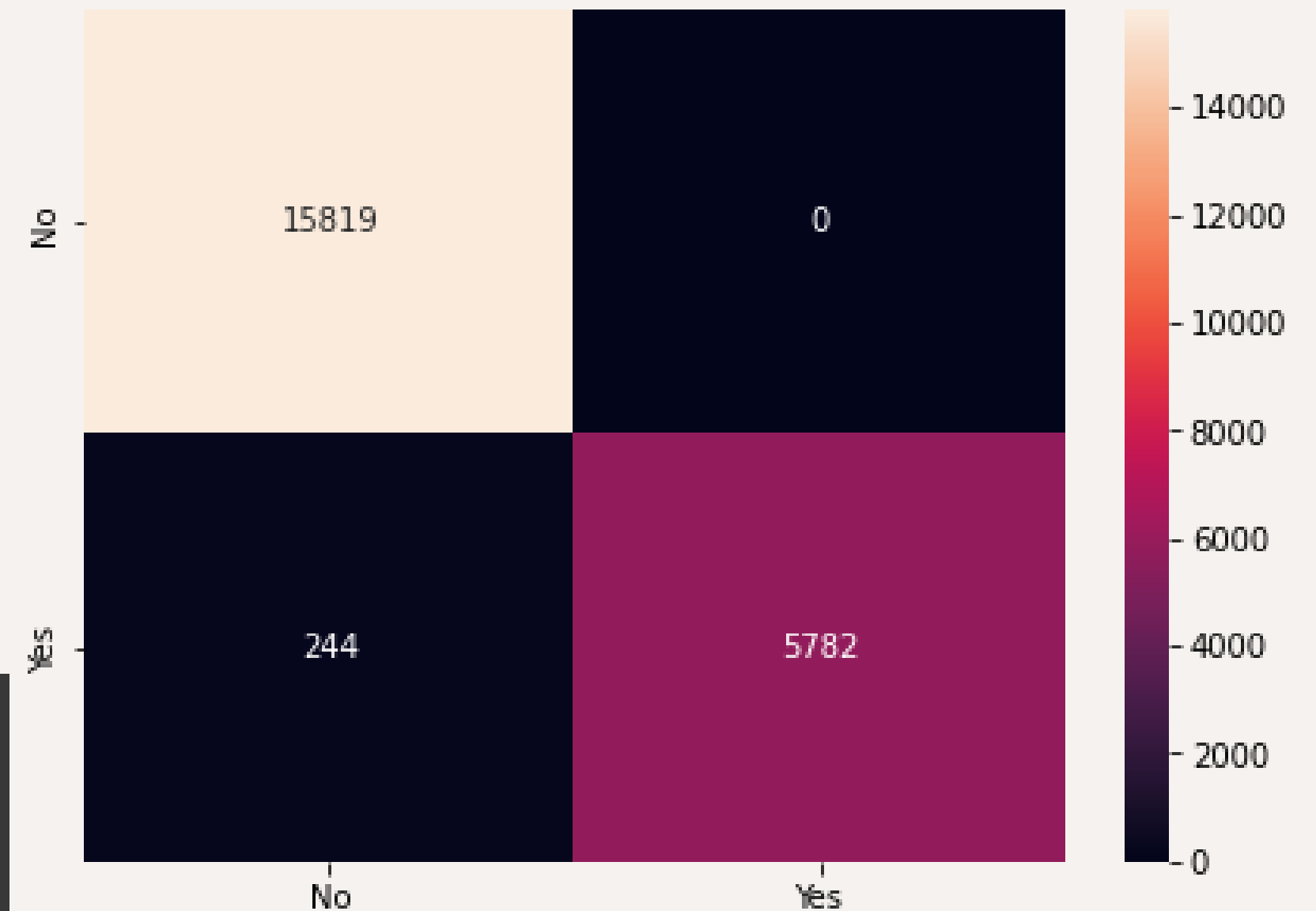
Confusion matrix  
for  
KNN

# Evaluating Models

SVM-Linear Kernel

Classification Report after training:

Classification Report				
	precision	recall	f1-score	support
1	1.00	0.96	0.98	6026
0	0.98	1.00	0.99	15819
accuracy			0.99	21845
macro avg	0.99	0.98	0.99	21845
weighted avg	0.99	0.99	0.99	21845



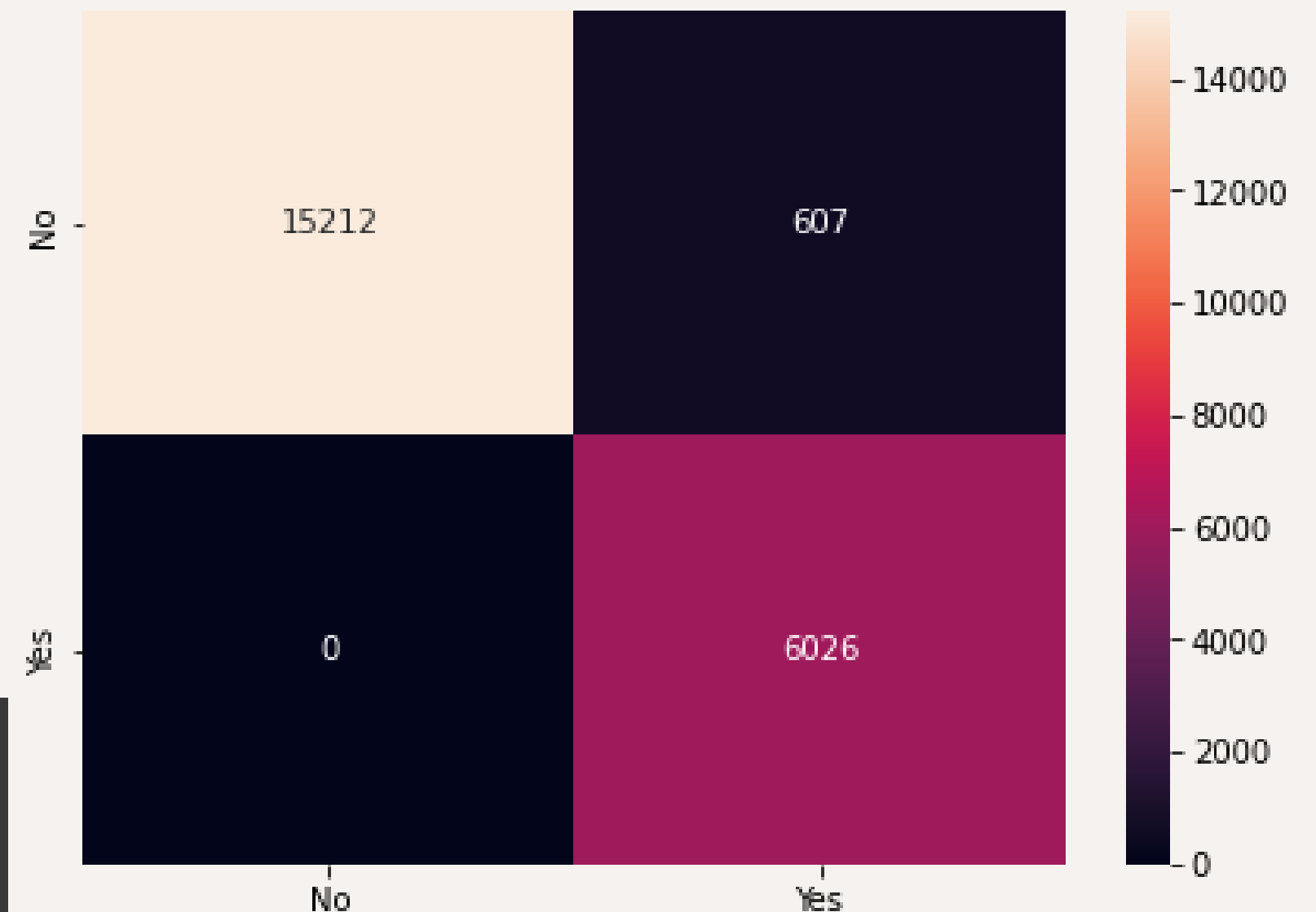


# Evaluating Models

Naive Bayes

Classification Report after training:

Classification Report					
	precision	recall	f1-score	support	
1	0.91	1.00	0.95	6026	
0	1.00	0.96	0.98	15819	
accuracy			0.97	21845	
macro avg	0.95	0.98	0.97	21845	
weighted avg	0.97	0.97	0.97	21845	



Confusion matrix  
for  
Naive Bayes

# Evaluating Models

Decision Tree with

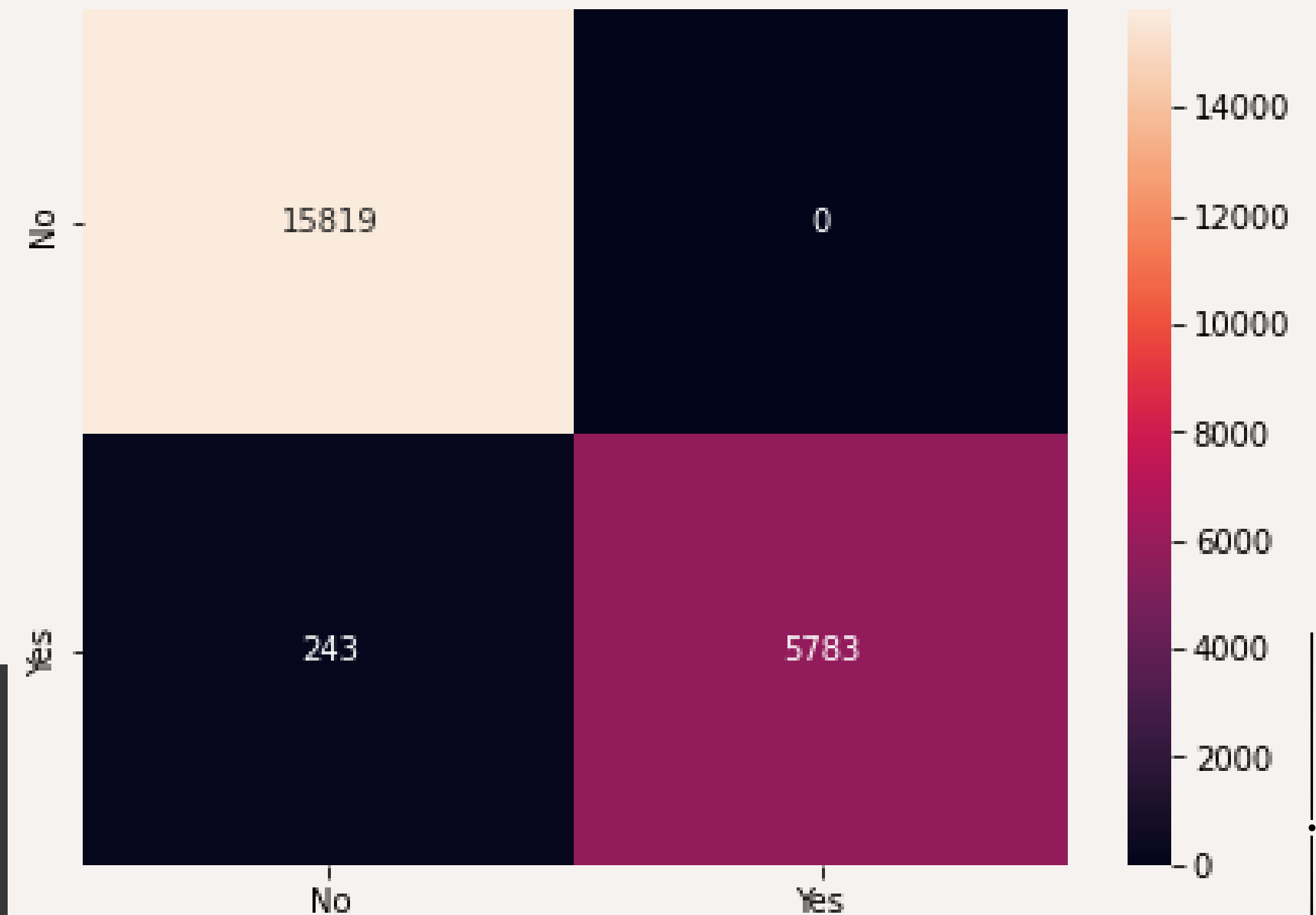
criterion = 'gini', max\_depth = 1

Classification Report after training:

```
Classification Report
              precision    recall  f1-score   support

     1         1.00      0.96      0.98         6026
     0         0.98      1.00      0.99        15819

 accuracy              0.99         21845
 macro avg              0.99         21845
 weighted avg           0.99         21845
```



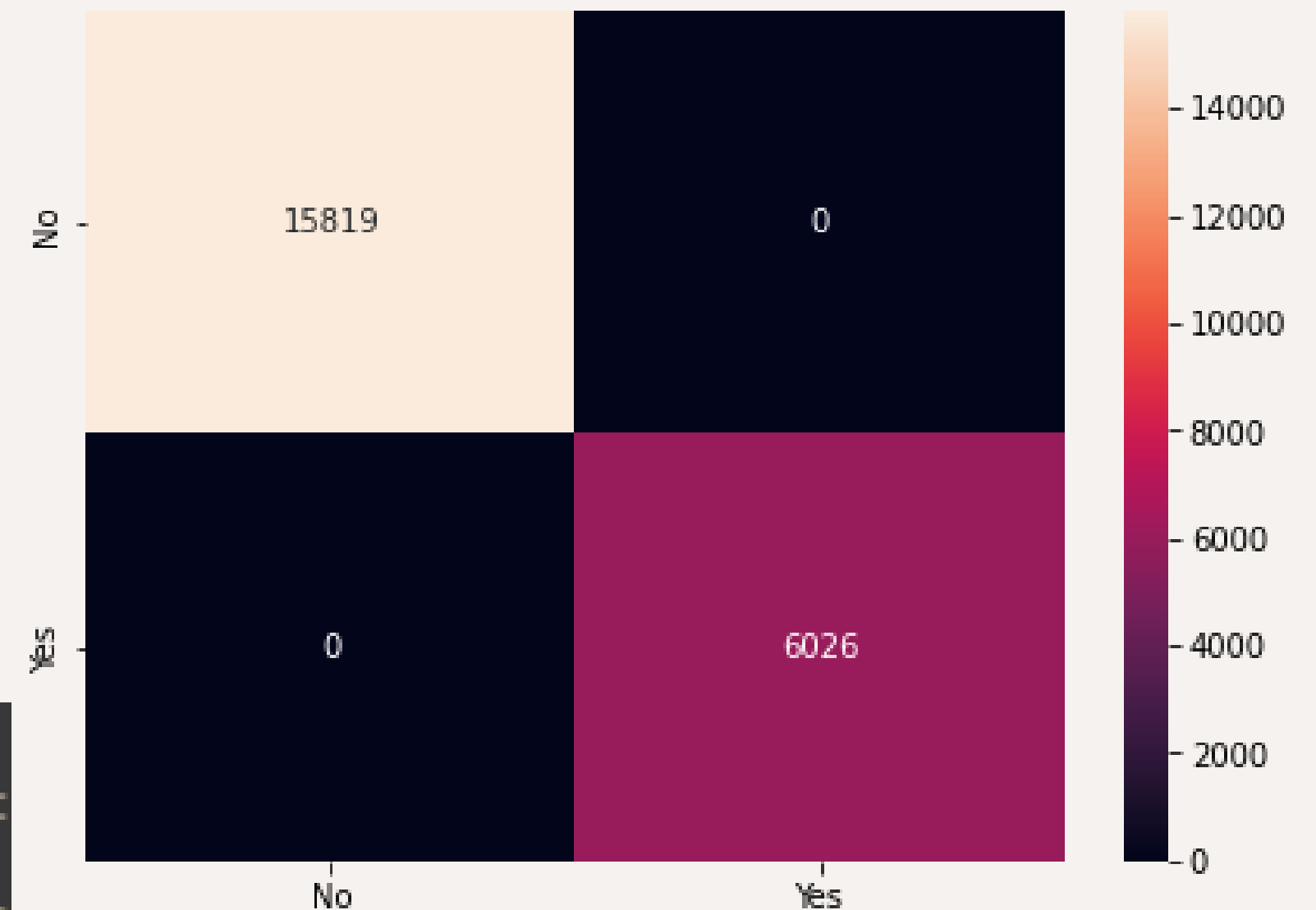
Confusion matrix  
for  
Decision Tree

# Chooosen Model

Bagging Classifier

Classification Report after training:

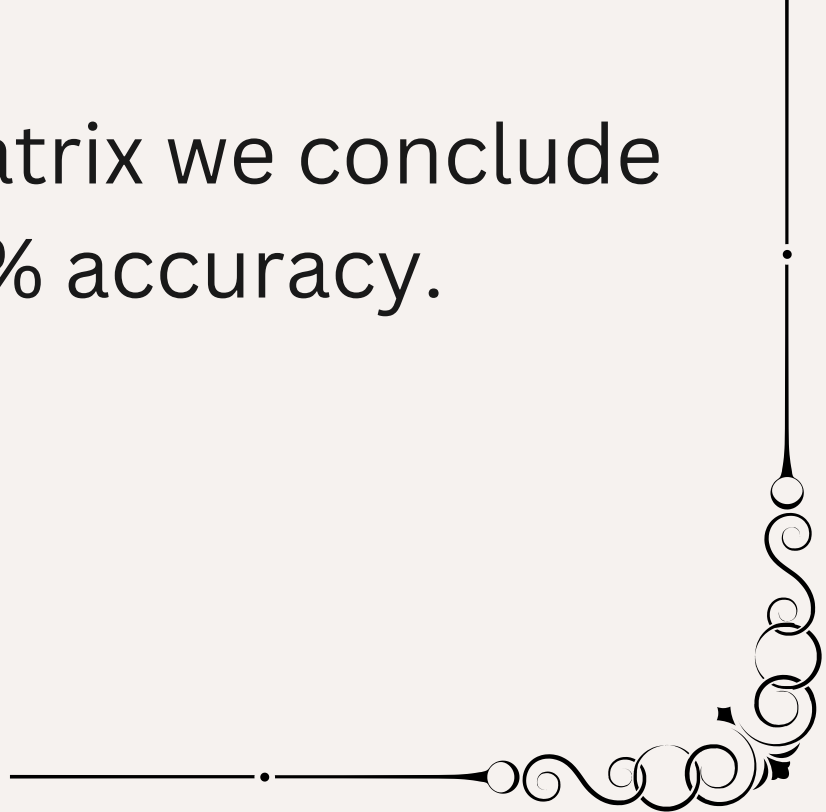
Classification Report				
	precision	recall	f1-score	support
1	1.00	1.00	1.00	6026
0	1.00	1.00	1.00	15819
accuracy			1.00	21845
macro avg	1.00	1.00	1.00	21845
weighted avg	1.00	1.00	1.00	21845



Confusion matrix  
for  
Bagging Classifier



# Conclusion

- In Bagging, each individual trees are independent of each other because they consider different subset of features and samples to predict a model.
  - In this model n\_estimators as 150 with default base\_estimator and random\_state=0 are used.
  - From, all the evaluation with accuracy score and confusion matrix we conclude with bagging classifier we got the best model which gives 100% accuracy.
- 



*Thank You*