```
import numpy as np
In [1]:
        import pandas as pd
        import os
        for dirname, _, filenames in os.walk('/kaggle/input'):
            for filename in filenames:
                print(os.path.join(dirname, filename))
        /kaggle/input/hotel-reservations-classification-dataset/Hotel Reservations.csv
In [2]:
        import pandas as pd
        df = pd.read csv("/kaggle/input/hotel-reservations-classification-dataset/Hotel Re-
In [3]:
        df.info()
In [4]:
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 36275 entries, 0 to 36274
        Data columns (total 19 columns):
             Column
                                                  Non-Null Count Dtype
        ---
                                                  -----
         0
             Booking ID
                                                  36275 non-null object
         1
             no_of_adults
                                                  36275 non-null int64
                                                  36275 non-null int64
             no_of_children
         2
             no_of_weekend_nights
                                                  36275 non-null int64
             no_of_week_nights
                                                  36275 non-null int64
         5
                                                  36275 non-null object
             type_of_meal_plan
             required car parking space
                                                  36275 non-null int64
         7
             room type reserved
                                                  36275 non-null object
                                                  36275 non-null int64
             lead time
         8
         9
             arrival year
                                                  36275 non-null int64
         10 arrival_month
                                                  36275 non-null int64
         11 arrival date
                                                  36275 non-null int64
                                                  36275 non-null object
         12 market_segment_type
         13 repeated_guest
                                                  36275 non-null int64
         14 no_of_previous_cancellations
                                                  36275 non-null int64
         15 no_of_previous_bookings_not_canceled 36275 non-null int64
         16 avg price per room
                                                  36275 non-null float64
         17 no of special requests
                                                  36275 non-null int64
                                                  36275 non-null object
         18 booking_status
        dtypes: float64(1), int64(13), object(5)
        memory usage: 5.3+ MB
        df.isnull().sum()
In [5]:
```

```
3/5/23, 4:56 PM
                                                           mid-term
                Booking_ID
                                                          0
      Out[5]:
                no_of_adults
                                                          0
                no_of_children
                                                          0
                no of weekend nights
                                                          0
                no of week nights
                                                          0
                                                          0
                type_of_meal_plan
                required_car_parking_space
                                                          0
                room type reserved
                                                          0
                lead_time
                                                          0
                                                          0
                arrival_year
                arrival_month
                                                          0
                arrival_date
                                                          0
                market_segment_type
                                                          0
                repeated_guest
                                                          0
                no of previous cancellations
                                                          0
                no_of_previous_bookings_not_canceled
                                                          0
                                                          0
                avg_price_per_room
                                                          0
                no_of_special_requests
                booking_status
                                                          0
                dtype: int64
                df.head()
      In [6]:
      Out[6]:
                              no_of_adults no_of_children no_of_weekend_nights no_of_week_nights type_of_mo
                   Booking_ID
                                        2
                                                      0
                                                                                            2
                0
                    INN00001
                                                                          1
                                                                                                     Me
                1
                    INN00002
                                        2
                                                      0
                                                                          2
                                                                                            3
                                                                                                    Not:
                2
                                                                          2
                                                                                            1
                    INN00003
                                        1
                                                      0
                                                                                                     Me
                3
                    INN00004
                                        2
                                                      0
                                                                          0
                                                                                            2
                                                                                                     Me
                                        2
                    INN00005
                                                      0
                                                                          1
                                                                                            1
                                                                                                    Not:
    4
                from sklearn.preprocessing import LabelEncoder
                11=LabelEncoder()
                df["Booking_ID"]=l1.fit_transform(df["Booking_ID"])
      In [8]:
                df["type_of_meal_plan"]=l1.fit_transform(df["type_of_meal_plan"])
                df["room_type_reserved"]=l1.fit_transform(df["room_type_reserved"])
                df["market_segment_type"]=l1.fit_transform(df["market_segment_type"])
                df["booking_status"]=l1.fit_transform(df["booking_status"])
```

df

In [9]:

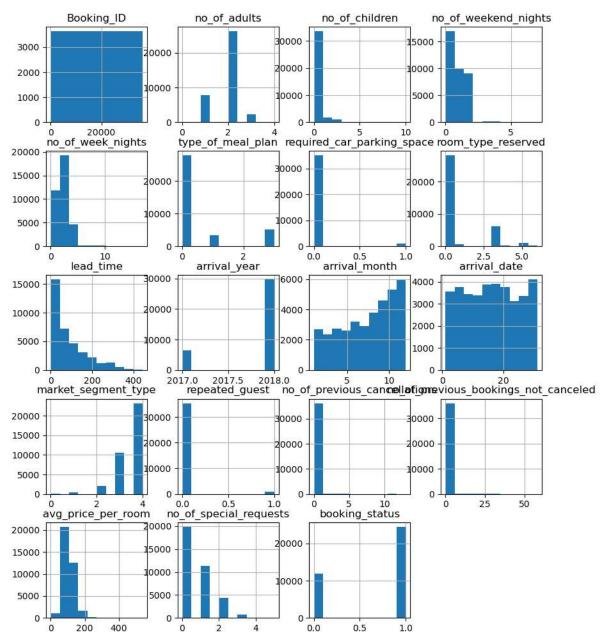
Out[9]:		Booking_ID	no_of_adults	no_of_children	no_of_weekend_nights	no_of_week_nights	type_(
	0	0	2	0	1	2	
	1	1	2	0	2	3	
	2	2	1	0	2	1	
	3	3	2	0	0	2	
	4	4	2	0	1	1	
	•••						
	36270	36270	3	0	2	6	
	36271	36271	2	0	1	3	
	36272	36272	2	0	2	6	
	36273	36273	2	0	0	3	
	36274	36274	2	0	1	2	

36275 rows × 19 columns

4					>
In [10]:	df.corr()				
Out[10]:		Booking_ID	no_of_adults	no_of_children	no_of_weekend_ni
	Booking_ID	1.000000	-0.009994	0.002153	-0.00(
	no_of_adults	-0.009994	1.000000	-0.019787	0.103
	no_of_children	0.002153	-0.019787	1.000000	0.029
	no_of_weekend_nights	-0.000083	0.103316	0.029478	1.000
	no_of_week_nights	0.001107	0.105622	0.024398	0.179
	type_of_meal_plan	-0.012476	0.025555	-0.086764	-0.02
	required_car_parking_space	-0.006579	0.011429	0.034244	-0.03
	room_type_reserved	-0.002375	0.270348	0.364073	0.057
	lead_time	-0.000535	0.097287	-0.047091	0.046
	arrival_year	-0.006980	0.076719	0.045983	0.05!
	arrival_month	0.003942	0.021841	-0.003076	-0.009
	arrival_date	0.005462	0.026338	0.025482	0.027
	market_segment_type	-0.051493	0.314103	0.130618	0.129
	repeated_guest	-0.001076	-0.192277	-0.036348	-0.067
	no_of_previous_cancellations	0.005227	-0.047426	-0.016390	-0.020
	no_of_previous_bookings_not_canceled	0.001465	-0.119166	-0.021189	-0.026
	avg_price_per_room	-0.002687	0.296886	0.337728	-0.004
	no_of_special_requests	-0.014795	0.189401	0.124486	0.060
	booking_status	-0.006237	-0.086920	-0.033078	-0.06
4					>

```
import seaborn as sns
In [11]:
               sns.heatmap(data=df.corr(),annot=True)
               <AxesSubplot:>
Out[11]:
                                                                                                                                        1.0
                                               Booking_ID - 1 0.00 08 239-051.001 00 6050 240 3060 0 33905 5050 00 0 0 5390 1050 2401 60
                                              no of adults -0.0 1 0.020.0.001026010.2709.00.0020202030.109047120.30.109.08
                                            no_of_children -0.0020 1 .00900408030436043040031225403.00606020304.402.03
                                                                                                                                       - 0.8
                                  no_of_weekend_nights -3e-050.02 1 ).18.00.70306.704.705.009507103.06.700.30260456106
                                       no_of_week_nights -0.00_101.020418 1 .088.304.990415-008.303.0093 10.-D.603034.90.304609
                                       type_of_meal_plan -.00.292698.792.708 1 .0-1320.06070.0809 4827.06.90.293.896.992202
                                                                                                                                       - 0.6
                            required car parking space -000000.DB490.D0991 1 .0B996.6D601769.0B037000706496.DB898
                                     room type reserved -002240.3506.0994.0103 1 0.1 D-0.00.633316.0260.098.40.40.02
                                                                                                                                       - 0.4
                                                 lead time -000.59704.0471-5.050-051 1 0.19.04-0659590404-607.863.40.4
                                               arrival year -.000000006556500016.0.14 1 0.8401945.01808926180581
                                            - 0.2
                                               arrival date -0.00562.602.502.700.93.4789.005.6005509941.0-0.0290.601.601.001050.80-0891
                                   market segment type -.05130.10.10.110.02.0003-060005-95000631 1 0.3340-73420.38.3-0.1
                                                                                                                                        0.0
                                          repeated guest -000.10906.6060-0.063301.026-0400.080304063 1 089.5-9.-070.021
                           no of previous cancellations -0.05.24.70.60.20.05.20.00.794363.90.90.30.7943 1 1.47.05600333
               no_of_previous_bookings_not_canceled -0.0-0.50202.02.094.003.00.04.00500.00.50.50.41 1 0.010200
                                                                                                                                        -0.2
                                     avg price per room -00 2730-340 04020 3006 960 477.0 63030 05.40 1083-9 .107.0 6031 1 ).1-9.1
                                  no_of_special_requests -.0 051 9.10206.04.002088150.0.058101010830.001003030718 1 ).2
                                           booking_status -000028.793.396.299.302.0960.20349. D80-010-D114.1010.3340-6.144.25
                                                                                                                             booking_status
                                                                                          lead time
                                                                                                    arrival date
                                                                                                                      avg_price_per_room
                                                                  no of adults
                                                                     no of children
                                                                        no_of_weekend_nights
                                                                            no_of_week_nights
                                                                               type_of_meal_plan
                                                                                   required_car_parking_space
                                                                                              arrival year
                                                                                                 arrival_month
                                                                                                        market_segment_type
                                                                                                           repeated_guest
                                                                                                               no_of_previous_cancellations
                                                                                                                  no of previous bookings not canceled
                                                                                      room_type_reserved
               import matplotlib.pyplot as plt
               df.hist(figsize=(10,12))
```

```
plt.show()
```



```
In [13]: x=df.iloc[:,:-1].values
    y=df.iloc[:,-1].values
```

In []:

```
In [14]: from imblearn.over_sampling import SMOTE
sm= SMOTE()
x_data,y_data = sm.fit_resample(x,y)
```

- In [15]: from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test=train_test_split(x_data,y_data,test_size=0.18,random_
- In [16]: from sklearn.preprocessing import StandardScaler
 std=StandardScaler()
 x=std.fit_transform(x_train)
- In [17]: #from sklearn.tree import DecisionTreeClassifier
 #dt = DecisionTreeClassifier(criterion='entropy', max_depth=5, random_state=12)
 #dt.fit(x_train,y_train)

 #from sklearn.linear_model import LogisticRegression

```
#lg = LogisticRegression()
          #lg.fit(x_train,y_train)
          #from sklearn.svm import SVC
          #svc=SVC(kernel='linear',random_state=11)
          #svc.fit(x_train,y_train)
         #from sklearn.neighbors import KNeighborsClassifier
          #knn = KNeighborsClassifier()
          #knn.fit(x_train,y_train)
 In [ ]:
In [18]:
         from sklearn.tree import DecisionTreeClassifier
          dt = DecisionTreeClassifier(criterion='entropy', max depth=5, random state=40)
          from sklearn.linear_model import LogisticRegression
         lg = LogisticRegression()
          from sklearn.naive bayes import GaussianNB
          nvb = GaussianNB()
         from sklearn.neighbors import KNeighborsClassifier
         knn = KNeighborsClassifier()
         from sklearn.ensemble import VotingClassifier
         vot = VotingClassifier(estimators=[("DT",dt),("LG",lg),("NVB",nvb),("KNN",knn)])
          vot.fit(x_train,y_train)
         from sklearn.metrics import accuracy score
         y_pred_vot = vot.predict(x_test)
         from sklearn.metrics import accuracy score
          ac vot=accuracy score(y test,y pred vot)
          ac vot
         /opt/conda/lib/python3.7/site-packages/sklearn/linear model/ logistic.py:818: Conv
         ergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
           extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,
         0.7593668147135861
Out[18]:
         from sklearn.tree import DecisionTreeClassifier
In [19]:
          dt = DecisionTreeClassifier(criterion='entropy', max_depth=5, random_state=12)
          dt.fit(x_train,y_train)
Out[19]: DecisionTreeClassifier(criterion='entropy', max_depth=5, random_state=12)
         from sklearn.linear_model import LogisticRegression
In [20]:
         lg = LogisticRegression()
```

```
lg.fit(x_train,y_train)
         /opt/conda/lib/python3.7/site-packages/sklearn/linear_model/_logistic.py:818: Conv
         ergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
           extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,
         LogisticRegression()
Out[20]:
In [21]: #from sklearn.svm import SVC
         #svc=SVC(kernel='linear',random state=11)
         #svc.fit(x train,y train)
         from sklearn.neighbors import KNeighborsClassifier
In [22]:
         knn = KNeighborsClassifier()
         knn.fit(x_train,y_train)
         KNeighborsClassifier()
Out[22]:
         y_pred_dt=dt.predict(x_test)
In [23]:
         from sklearn.metrics import accuracy score
         ac_dt=accuracy_score(y_test,y_pred_dt)*100
         print("ac_dt = ",ac_dt)
         print(f'----')
         y_pred_lg=lg.predict(x_test)
         from sklearn.metrics import accuracy_score
         ac_lg=accuracy_score(y_test,y_pred_lg)*100
         print("ac_lg = ",ac_lg)
print(f'----')
         #y_pred_svm=svc.predict(x_test)
         #from sklearn.metrics import accuracy_score
         ##ac_svm=accuracy_score(y_test,y_pred_svm)*100
         #print("ac_svm = ",ac_svm)
         #print(f'----')
         y_pred_knn=knn.predict(x_test)
         from sklearn.metrics import accuracy_score
         ac_knn=accuracy_score(y_test,y_pred_knn)*100
         print("ac_knn = ",ac_knn)
         print(f'----')
         ac_dt = 85.05864935656531
         -----
         ac_{lg} = 71.27889761986106
         ______
         ac knn = 73.19211934859355
In [24]:
         from sklearn.ensemble import RandomForestClassifier
         ran = RandomForestClassifier(n_estimators=40, max_features=4)
         ran.fit(x_train,y_train)
         from sklearn.metrics import accuracy_score
```

```
y_pred_ran = ran.predict(x_test)
        from sklearn.metrics import accuracy_score
        ac_ran=accuracy_score(y_test,y_pred_ran)
        ac ran
        0.9221045439016058
Out[24]:
        from sklearn.ensemble import BaggingClassifier
In [25]:
         bag = BaggingClassifier(base estimator=dt,n estimators=5,random state=40)
        bag.fit(x_train,y_train)
        from sklearn.metrics import accuracy_score
        y_pred_bag = bag.predict(x_test)
        from sklearn.metrics import accuracy score
        ac_bag=accuracy_score(y_test,y_pred_bag)
        ac_bag
        0.8509281403029267
Out[25]:
        print("ac_vot = ",ac_vot*100)
In [26]:
        print(f'----')
         print("ac_ran = ",ac_ran*100)
        print(f'----')
        print("ac_bag = ",ac_bag*100)
print(f'----')
        print("ac_dt = ",ac_dt)
        print(f'----')
        print("ac_lg = ",ac_lg)
        print(f'----')
        print("ac_knn = ",ac_knn)
        print(f'----')
        ac_vot = 75.93668147135861
        -----
        ac_ran = 92.21045439016058
        -----
        ac_bag = 85.09281403029267
        -----
        ac_dt = 85.05864935656531
        -----
        ac lg = 71.27889761986106
        ac_knn = 73.19211934859355
        **Report for random forest classification.
In [29]: from sklearn.metrics import classification report
         print(classification_report(y_test,y_pred_ran))
```

	precision	recall	f1-score	support
0 1	0.93 0.91	0.91 0.94	0.92 0.92	4383 4398
accuracy			0.92	8781
macro avg	0.92	0.92	0.92	8781
weighted avg	0.92	0.92	0.92	8781

```
In [30]: from sklearn.metrics import confusion_matrix
  confusion_matrix = confusion_matrix(y_test,y_pred_ran)
  print(confusion_matrix)
```

[[3982 401] [283 4115]]

Inference

Using DecisionTreeClassifier, the model's accuracy was around 92%; precision, recall, and f1-score were92%,92%, and 92% after the model had been trained.

In []: