### Importing Libraries

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

Loading the dataset

```
df = pd.read_csv('dataset.csv')
```

EDA Exploratory Data Analysis and Data Cleaning

```
df.head()
{"type": "dataframe", "variable name": "df"}
df.tail()
{"type": "dataframe"}
df.shape
(119390, 32)
df.columns
'arrival_date_day_of_month', 'stays_in_weekend_nights',
       'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',
       'country', 'market_segment', 'distribution_channel',
       'is_repeated_guest', 'previous_cancellations',
       'previous bookings not canceled', 'reserved room type',
       'assigned_room_type', 'booking_changes', 'deposit_type',
'agent',
       'company', 'days in waiting list', 'customer type', 'adr',
       'required_car_parking_spaces', 'total_of_special_requests',
       'reservation status', 'reservation status date'],
     dtype='object')
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 119390 entries, 0 to 119389
Data columns (total 32 columns):
#
    Column
                                   Non-Null Count
                                                   Dtype
    hotel
                                   119390 non-null object
```

```
is canceled
                                     119390 non-null int64
 2
     lead time
                                     119390 non-null int64
 3
     arrival_date_year
                                     119390 non-null int64
 4
     arrival date month
                                     119390 non-null object
 5
     arrival_date_week_number
                                     119390 non-null int64
     arrival_date_day_of_month
 6
                                     119390 non-null int64
 7
     stays in weekend nights
                                     119390 non-null int64
 8
     stays in week nights
                                     119390 non-null int64
 9
                                     119390 non-null int64
     adults
 10 children
                                     119386 non-null float64
 11 babies
                                     119390 non-null int64
 12 meal
                                     119390 non-null object
 13 country
                                     118902 non-null object
 14 market_segment
                                     119390 non-null
                                                      object
 15 distribution_channel
                                     119390 non-null object
                                     119390 non-null int64
 16 is_repeated_guest
 17 previous_cancellations
                                     119390 non-null int64
 18 previous_bookings_not_canceled 119390 non-null int64
 19 reserved_room_type
                                     119390 non-null object
 20 assigned_room_type
                                     119390 non-null object
 21 booking_changes
                                     119390 non-null int64
 22 deposit type
                                     119390 non-null object
 23 agent
                                     103050 non-null float64
 24 company
                                     6797 non-null
                                                      float64
 25 days in waiting list
                                     119390 non-null int64
                                     119390 non-null object
 26 customer type
 27 adr
                                     119390 non-null float64
 28 required car parking spaces
                                     119390 non-null int64
 29 total_of_special_requests
                                     119390 non-null int64
 30 reservation_status
                                     119390 non-null object
     reservation_status_date
                                     119390 non-null datetime64[ns]
 31
dtypes: datetime64[ns](1), float64(4), int64(16), object(11)
memory usage: 29.1+ MB
# Convert 'reservation status date' to datetime
df['reservation status date'] =
pd.to datetime(df['reservation status date'], format='%d/%m/%Y')
df.describe(include = 'object')
{"summary":"{\n \"name\": \"df\",\n \"rows\": 4,\n \"fields\": [\n
{\n \"column\": \"hotel\",\n \"properties\": {\n
\"dtype\": \"string\",\n \"num unique values\": 4,\n
                          2,\n
                                       \"79330\",\n
\"samples\": [\n
                                \"semantic_type\": \"\",\n
\"119390\"\n
                    ],\n
\"description\": \"\"\n }\n {\n \"column\":
\"arrival_date_month\",\n \"properties\": {\n \"dtype\":
\"string\",\n \"num_unique_values\": 4,\n \"samples\":
                           \"13877\",\n\\"119390\"\
[\n
             12,\n
                     \"semantic_type\": \"\",\n
n
         ],\n
```

```
\"num_unique_values\": 4,\n \"samples\": [\n 5,\n
],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"country\",\n \"properties\":
_\"5647<del>7</del>\",\n
                                                                                                                                                                           \"119390\"\
 [\n
                              8,\n \"56477\",\n ],\n \"semantic_type\": \"\",\n
5,\n \"9787\",\n \"semantic_type\": \"\",\n
\"column\":
                                                                                                                                                                                              \"dtype\":
 \"string\",\n \"num_unique_values\": 4,\n \"samples\":
 [\n 10,\n
                                                                         \"85994\",\n
                                                                                                                                                                               \"119390\"\
                              ],\n \"semantic_type\": \"\",\n
\"column\":
                                                                                                                                                                                                  \"dtvpe\":
 \"string\",\n \"num_unique_values\": 4,\n
                                                                                                                                                                                                  \"samples\":
 [\n
                                                                          \"74053\",\n
                                                                                                                                                                                \"119390\"\
                       12,\n
n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n },\n {\n
                                                                                                                                                                                       \"column\":
 \"deposit_type\",\n \"properties\": {\n
                                                                                                                                                                                      \"dtype\":
 \"string\",\n
                                                                          \"num_unique_values\": 4,\n
                                                                                                                                                                                     \"samples\":
                         3,\n \"104641\",\n ],\n \"semantic_type\": \"\",\n
                                                                                                                                                                                \"119390\"\
 [\n
 \ensuremath{\mbox{"description}}: \ensuremath{\mbox{"\n}} \ensuremath{\mbox{n}} \ensuremath{\mbox{\mbox{$\backslash$}}}, \ensuremath{\mbox{$\backslash$}} \ensuremath{
                                                                                                                                                                                       \"column\":
 \"customer_type\",\n \"properties\": {\n
                                                                                                                                                                                         \"dtype\":
                                                                                                                                                                                 \"samples\":
 \"string\",\n
                                                                         \"num_unique_values\": 4,\n
                                             4,\n \"89013\ ,\"
\"semantic_type\": \"\",\n
                                                                       \"8961<del>3</del>\",\n
                                                                                                                                                                           \"119390\"\
 [\n
                               ],\n
\"description\": \"\"\n }\n },\n {\n \"column\":
\"reservation_status\",\n \"properties\": {\n \"dtype\":
\"string\",\n \"num_unique_values\": 4,\n \"samples\":
                              3,\n \"75166\",\n \"semantic_type\": \"\",\n
                                                                                                                                                                           \"119390\"\
 [\n
 \label{localization} $$ \color= 1.00 \colo
 for col in df.describe(include = 'object').columns:
               print(col)
               print(df[col].unique())
               print('-'*50)
```

```
hotel
['Resort Hotel' 'City Hotel']
arrival date month
['July' 'August' 'September' 'October' 'November' 'December' 'January'
'February' 'March' 'April' 'May' 'June']
meal
['BB' 'FB' 'HB' 'SC' 'Undefined']
country
['PRT' 'GBR' 'USA' 'ESP' 'IRL' 'FRA' nan 'ROU' 'NOR' 'OMN' 'ARG' 'POL'
 'DEU' 'BEL' 'CHE' 'CN' 'GRC' 'ITA' 'NLD' 'DNK' 'RUS' 'SWE' 'AUS'
'EST'
 'CZE' 'BRA' 'FIN' 'MOZ' 'BWA' 'LUX' 'SVN' 'ALB' 'IND' 'CHN' 'MEX'
'MAR'
 'UKR' 'SMR' 'LVA' 'PRI' 'SRB' 'CHL' 'AUT' 'BLR' 'LTU' 'TUR' 'ZAF'
' AGO '
 'ISR' 'CYM' 'ZMB' 'CPV' 'ZWE' 'DZA' 'KOR' 'CRI' 'HUN' 'ARE' 'TUN'
'JAM'
 'HRV' 'HKG' 'IRN' 'GEO' 'AND' 'GIB' 'URY' 'JEY' 'CAF' 'CYP' 'COL'
'GGY'
 'KWT' 'NGA' 'MDV' 'VEN' 'SVK' 'FJI' 'KAZ' 'PAK' 'IDN' 'LBN' 'PHL'
'SEN'
'SYC' 'AZE' 'BHR' 'NZL' 'THA' 'DOM' 'MKD' 'MYS' 'ARM' 'JPN' 'LKA'
'CUB'
 'CMR' 'BIH' 'MUS' 'COM' 'SUR' 'UGA' 'BGR' 'CIV' 'JOR' 'SYR' 'SGP'
'BDI'
 'SAU' 'VNM' 'PLW' 'QAT' 'EGY' 'PER' 'MLT' 'MWI' 'ECU' 'MDG' 'ISL'
'117B'
 'NPL' 'BHS' 'MAC' 'TGO' 'TWN' 'DJI' 'STP' 'KNA' 'ETH' 'IRQ' 'HND'
'RWA'
 'KHM' 'MCO' 'BGD' 'IMN' 'TJK' 'NIC' 'BEN' 'VGB' 'TZA' 'GAB' 'GHA'
'TMP'
'GLP' 'KEN' 'LIE' 'GNB' 'MNE' 'UMI' 'MYT' 'FRO' 'MMR' 'PAN' 'BFA'
'I RY'
'MLI' 'NAM' 'BOL' 'PRY' 'BRB' 'ABW' 'AIA' 'SLV' 'DMA' 'PYF' 'GUY'
'LCA'
'ATA' 'GTM' 'ASM' 'MRT' 'NCL' 'KIR' 'SDN' 'ATF' 'SLE' 'LAO']
market segment
['Direct' 'Corporate' 'Online TA' 'Offline TA/TO' 'Complementary'
'Groups'
'Undefined' 'Aviation']
distribution_channel
['Direct' 'Corporate' 'TA/TO' 'Undefined' 'GDS']
reserved room type
['C' 'A' 'D' 'E' 'G' 'F' 'H' 'L' 'P' 'B']
```

```
assigned_room_type
['C' 'A' 'D' 'E' 'G' 'F' 'I' 'B' 'H' 'P' 'L' 'K']
deposit type
['No Deposit' 'Refundable' 'Non Refund']
customer type
['Transient' 'Contract' 'Transient-Party' 'Group']
reservation_status
['Check-Out' 'Canceled' 'No-Show']
df.isnull().sum()
hotel
is canceled
                                       0
                                       0
lead time
                                       0
arrival date year
arrival date month
                                       0
arrival date week number
                                       0
arrival date day of month
                                       0
stays_in_weekend_nights
                                       0
stays in_week_nights
                                       0
                                       0
adults
children
                                       4
babies
                                       0
                                       0
meal
country
                                     488
market segment
                                       0
                                       0
distribution channel
is repeated guest
                                       0
previous cancellations
previous bookings not canceled
                                       0
reserved_room_type
                                       0
assigned_room_type
                                       0
                                       0
booking_changes
deposit_type
                                  16340
agent
                                  112593
company
days in waiting list
                                       0
customer type
                                       0
                                       0
adr
required car parking spaces
                                       0
total_of_special_requests
                                       0
reservation status
                                       0
reservation status date
                                       0
dtype: int64
```

```
df.dropna().isnull().sum()
hotel
                                0
                                0
is canceled
lead time
                                0
                                0
arrival date year
arrival_date_month
                                0
arrival date week number
                                0
arrival date day of month
                                0
stays in weekend nights
                                0
                                0
stays in week nights
adults
                                0
                                0
children
                                0
babies
meal
                                0
                                0
country
market segment
                                0
                                0
distribution channel
is repeated quest
                                0
previous cancellations
                                0
previous bookings_not_canceled
                                0
reserved room type
                                0
assigned room type
                                0
booking changes
                                0
                                0
deposit type
                                0
days_in_waiting_list
                                0
customer type
                                0
adr
required_car_parking_spaces
                                0
total of special requests
                                0
reservation status
                                0
                                0
reservation status date
dtype: int64
df.describe()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 8,\n \"fields\": [\n
{\n \"column\": \"is_canceled\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 42210.59518981826,\n
\"min\": 0.0,\n \"max\": 119390.0,\n
\"num_unique_values\": 5,\n
0.37041628277075134,\n
0.
                                \"samples\": [\n
                             0.48291822659316763,\n
                                                             0.0\n
          \"semantic type\": \"\",\n \"description\": \"\"\n
],\n
\"std\":
                                                \"max\": 119390.0,\
      \"num_unique_values\": 8,\n \"samples\": [\n
                                              119390.0\n
104.01141636652986,\n
                          160.0,\n
                                                                ],\
        \"semantic_type\": \"\",\n \"description\": \"\"\n
n
              {\n \"column\": \"arrival_date_year\",\n
}\n
      },\n
```

```
\"properties\": {\n \"dtype\": \"number\",\n \"std\": 41605.6829984143,\n \"min\": 0.7074759445202401,\n
                                \"num_unique_values\": 6,\n
\"max\": 119390.0,\n
\"samples\": [\n 119390.0,\n 2016.156554150264,\n 0.7074759445202401\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n \"column\": \"arrival_date_week_number\",\n \"properties\": {\n \"description\": \"\"
                                                              2016.156554150264,\n
\"min\": 1.0,\n \"max\": 119390.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n 27.16517296255968,\n 38.0,\n 119390.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                                    ],\n
n },\n {\n \"column\": \"arrival_date_day_of_month\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 42205.50879379506,\n \"min\": 1.0,\n \"max\": 119390.0,\
n \"num_unique_values\": 8,\n \"samples\": [\n 15.798241058715135,\n 23.0,\n 119390.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"stays_in_weekend_nights\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 42209.53133083881,\n \"min\": 0.0,\n \"max\": 119390.0,\
n \"num_unique_values\": 7,\n \"samples\": [\n 119390.0,\n 0.9275986263506156,\n 19.0\n
                                                                                     ],\n
42207.539491149604,\n \"min\": 0.0,\n \"max\": 119390.0,\n \"num_unique_values\": 6,\n \"samples\": [\n 119390.0,\n 1.8564033838679956,\n
0.0,\n \"max\": 119386.0,\n \"num_unique_values\": 5,\n
{\n \"dtype\": \"number\",\n \"std\":
42210.22904846642,\n \"min\": 0.0,\n \"max\": 119390.0,\
n \"num_unique_values\": 5,\n \"samples\": [\n
0.007948739425412514,\n 0.09743619130130332,\n
                                                                                     0.0\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
```

```
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n \"num_unique_values\": 5,\n \"samples\": [\n
0.087117849066<del>0</del>8594,\n 0.8443363841518928,\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\":
\"previous_bookings_not_canceled\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 42207.02766341039,\n
\"min\": 0.0,\n \"max\": 119390.0,\n
\"num_unique_values\": 5,\n \"samples\": [\n 0.13709690928888515,\n 1.4974368477089461,\n 0.0\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\": \"booking_changes\",\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\":
42209.63515931599,\n \"min\": 0.0,\n \"max\": 119390.0,\
n \"num_unique_values\": 5,\n \"samples\": [\n
0.221124047240<del>1</del>3737,\n 0.6523055726747069,\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n     },\n     {\n          \"column\": \"days_in_waiting_list\",\n
\"properties\": {\n          \"dtype\": \"number\",\n          \"std\":
42190.20344235136,\n         \"min\": 0.0,\n          \"max\": 119390.0,\
n \"num_unique_values\": 5,\n \"samples\": [\n 2.321149174972778,\n 17.594720878752014,\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\": \"total_of_special_requests\",\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\":
42210.36738912644,\n \"min\": 0.0,\n \"max\": 119390.0,\
n \"num_unique_values\": 6,\n \"samples\": [\n
119390.0,\n 0.5713627607002262,\n
0.7927984228089581\n ],\n \"semantic type\": \"\",\n
```

```
\"dtype\": \"date\",\n \"min\": \"1970-01-01 \\
00:00:00:00.000119390\",\n \"max\": \"2017-09-14 00:00:00\",\n \\"num_unique_values\": 7,\n \"samples\": [\n
n}","type":"dataframe"}
df = df[df['adr'] < 5000]
df.describe()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 8,\n \"fields\": [\n
{\n \"column\": \"is_canceled\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 40869.56721485206,\n
\"min\": 0.0,\n \"max\": 115597.0,\n
\"num_unique_values\": 5,\n \"samples\": [\n 0.36986253968528593,\n 0.48276936275999693,\n
                 \"semantic_type\": \"\",\n \"description\": \"\"\n
1,\n
}\n    },\n    {\n     \"column\": \"lead_time\",\n
\"properties\": {\n          \"dtype\": \"number\",\n          \"std\":
40809.80185564527,\n         \"min\": 0.0,\n         \"max\": 115597.0,\
n \"num_unique_values\": 8,\n \"samples\": [\n 105.01959393409864,\n 162.0,\n 115597.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\": \"arrival_date_year\",\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\": 40264.85470157284,\n \"min\": 0.7069896291119789,\n \"max\": 115597.0,\n \"num_unique_values\": 6,\n \"samples\": 6,\n \"15507.0,\n \"num_unique_values\": 6,\n \"samples\": 6
\"min\": 1.0,\n \"max\": 115597.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
27.045788385511734,\n 38.0,\n 115597.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
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n \"num_unique_values\": 8,\n \"samples\": [\n 15.770089189165809,\n 23.0,\n 115597.0\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
                                                                                                                                                                                ],\n
n },\n {\n \"column\": \"stays_in_weekend_nights\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 40868.50372950941,\n \"min\": 0.0,\n \"max\": 115597.0,\
n \"num_unique_values\": 7,\n \"samples\": [\n
```

```
115597.0,\n 0.9207678399958477,\n
n },\n {\n \"column\": \"stays_in_week_nights\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 40866.66477184999,\n \"min\": 0.0,\n \"max\": 115597.0,\
n \"num_unique_values\": 8,\n \"samples\": [\n 2.4814657819839616,\n 3.0,\n 115597.0\n ],\r\"semantic_type\": \"\",\n \"description\": \"\"\n }\n \,\n \"column\": \"adults\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\":
40866.512138446815,\n \"min\": 0.0,\n \"max\": 115597.0,\n \"num_unique_values\": 6,\n \"samples\": [\n 115597.0,\n 1.8460340666280266,\n
0.0,\n \"max\": 115593.0,\n \"num_unique_values\": 5,\n \"samples\": [\n 0.08267801683492945,\n 0.3499607735485756,\n 0.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\
40869.20118364985,\n\\"min\": 0.0,\n\\"max\": 115597.0,\
n \"num_unique_values\": 5,\n \"samples\": [\n
0.0074915438981980505,\n 0.09510277181586718,\n
                                                                   0.0
n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n }\n {\n \"column\": \"is_repeated_guest\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 40869.65018514639,\n \"min\":
0.0,\n \"max\": 115597.0,\n \"num_unique_values\": 5,\n
n \"num_unique_values\": 5,\n \"samples\": [\n
0.089803368599\overline{5}3113, \n 0.857844399643005, \n 0.0\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\":
\"previous_bookings_not_canceled\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 40865.9985148412,\n
\"min\": 0.0,\n \"max\": 115597.0,\n
\"num_unique_values\": 5,\n \"samples\": [\n 0.1408254539477668,\n 1.521073870276689,\n
                                                              0.0\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
```

```
40868.60763556607,\n\\"min\": 0.0,\n
                                          \"max\": 115597.0,\
       \"num unique values\": 5,\n \"samples\": [\n
0.21700390148533266,\n
                             0.6472000688286631,\n
                                                          0.0\n
           \"semantic type\": \"\",\n
                                          \"description\": \"\"\n
      },\n
             {\n \"column\": \"days_in_waiting_list\",\n
}\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\": 40849.165604258305,\n \"min\": 0.0,\n \"max\": 115597.0,\n \"num_unique_values\": 5,\n \"samples\": [\n
2.3876398176423264,\n 17.863820777442267,\n
     \"semantic_type\": \"\",\n \"description\": \"\"\n
],\n
             }\n
        \"dtype\": \"number\",\n \"std\": 40838.2599250576,\n
\"min\": -6.38,\n\\"max\": 115597.0,\n
\"num_unique_values\": 8,\n \"samples\": [\n
97.03664299246519,\n
                           121.37,\n
                                            115597.0\n
       \"semantic type\": \"\",\n \"description\": \"\"\n
n \"num_unique_values\": 5,\n \"samples\": [\n
                              0.2401089616347926,\n
0.059672828879642204,\n
           \"semantic_type\": \"\",\n \"description\": \"\"\n
],\n
             {\n \"column\": \"total_of_special_requests\",\n
}\n
      },\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\": 40869.340246054584,\n \"min\": 0.0,\n \"max\":
115597.0,\n \"num_unique_values\": 6,\n 115597.0,\n 0.5616754760071628,\n
                                              \"samples\": [\n
0.7853326807198735\n
                                    \"semantic type\": \"\",\n
                          ],\n
\"description\": \"\"\n
                         }\n
                                 },\n {\n \"column\":
\"reservation_status_date\",\n \"properties\": {\n
\"dtype\": \"date\",\n \\"min\": \"1970-01-01 \\"max\": \"2017-09-14 00:00:00\",\n \\"num_unique_values\": 7,\n \\"samples\": [\n
\"115597\",\n\\"2016-07-24\13:40:20.128550144\",\n
\"2017-02-02 00:00:00\"\n ],\n \"semantic_type\": \"\",\
n \"description\": \"\"n }\n
                                         }\n ]\
n}","type":"dataframe"}
```

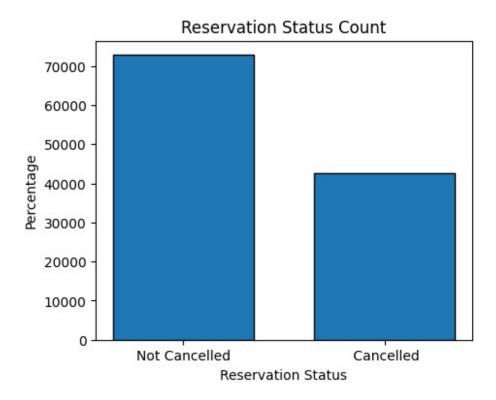
Data Analysis and Visualizations (Analysis And Findings)

```
cancelled_perc = df['is_canceled'].value_counts(normalize = True)
print(cancelled_perc)

plt.figure(figsize = (5,4))
plt.title('Reservation Status Count')
plt.bar(['Not Cancelled','
Cancelled'],df['is_canceled'].value_counts(),edgecolor = 'k',width = 0.7)
plt.xlabel('Reservation Status')
```

```
plt.ylabel('Percentage')
plt.show()

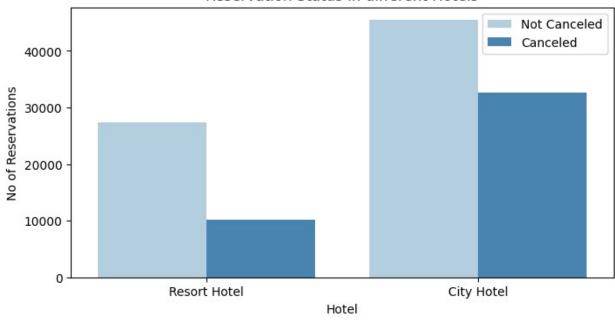
is_canceled
0    0.630137
1    0.369863
Name: proportion, dtype: float64
```



The accompanying bar graph shows the percentage of reservations that are canceled and those that are not. It is obvious that there are still a significant number of reservations that have not been canceled. There are still 37% of clients who canceled their reservation, which has a significant impact on the hotel's earning.

```
plt.figure(figsize = (8,4))
ax1=sns.countplot(x='hotel',hue='is_canceled',data=df,palette='Blues')
legend_labels,_ = ax1.get_legend_handles_labels()
plt.title('Reservation Status in different Hotels')
plt.xlabel('Hotel')
plt.ylabel('No of Reservations')
plt.legend(['Not Canceled', 'Canceled'])
plt.show()
```

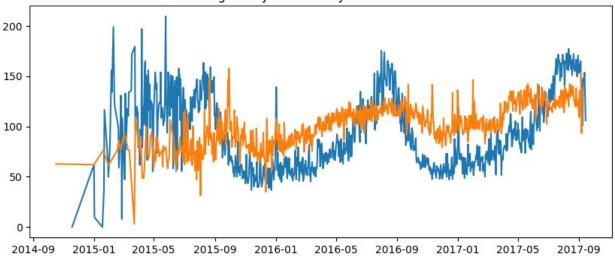
## Reservation Status in different Hotels



In comparison to resort hotels, city hotels have more bookings. It's possible that resort hotel's are more expensive than thse in cities.

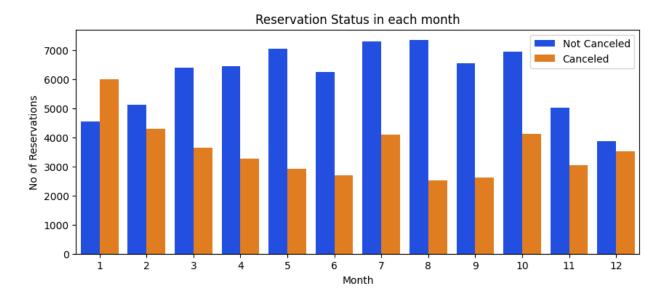
```
resort hotel = df[df['hotel']=='Resort Hotel']
resort hotel['is canceled'].value counts(normalize = True)
is canceled
     0.729877
     0.270123
1
Name: proportion, dtype: float64
city hotel = df[df['hotel']=='City Hotel']
city hotel['is canceled'].value counts(normalize = True)
is canceled
0
     0.582137
     0.417863
Name: proportion, dtype: float64
resort hotel = resort hotel.groupby('reservation status date')
['adr'].mean()
city hotel = city hotel.groupby('reservation status date')
['adr'].mean()
plt.figure(figsize = (10,4))
plt.title('Average Daily Rate in City and Resort Hotels')
plt.plot(resort hotel.index, resort hotel, label = 'Resort Hotel')
plt.plot(city hotel.index, city hotel, label = 'City Hotel')
plt.show()
```

#### Average Daily Rate in City and Resort Hotels



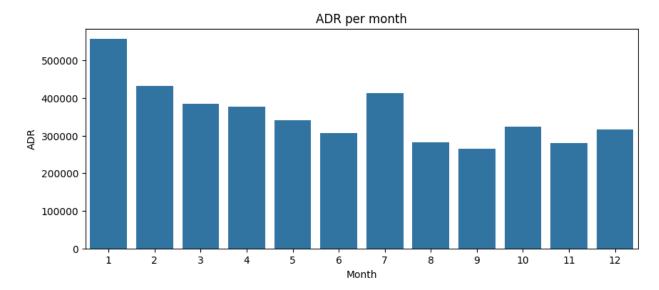
The line graph above shows that, on certain days, the average daily rate for a city hotel is less than that of a resort hotel, and on other days, it is even less. It goes without saying that weekends and hoildays may see a rise in resort hotel rates.

```
df['month'] = df['reservation_status_date'].dt.month
plt.figure(figsize=(10,4))
ax1 =
sns.countplot(x='month', hue='is_canceled', data=df, palette='bright')
legend_labels,_ = ax1.get_legend_handles_labels()
ax1.legend(bbox_to_anchor=(1,1))
plt.title('Reservation Status in each month')
plt.xlabel('Month')
plt.ylabel('No of Reservations')
plt.legend(['Not Canceled', 'Canceled'])
plt.show()
```



We have developed the grouped bar graph to analyze the months with the highest and lowest reservation levels according to reservations and the number of canceled reservations are largest in the month of august whereas January is the moth with the most canceled reservations.

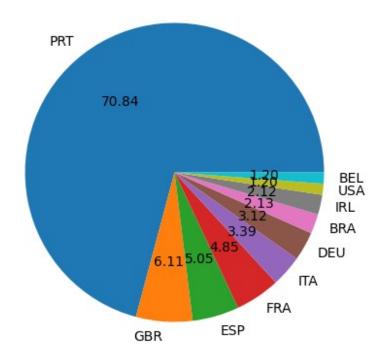
```
plt.figure(figsize = (10,4))
plt.title('ADR per month')
sns.barplot(x='month', y='adr', data=df[df['is_canceled'] ==
1].groupby('month')[['adr']].sum().reset_index())
plt.xlabel('Month')
plt.ylabel('ADR')
plt.show()
```



This bar graph demonstrates that cancellations are most common when prices are greatest and are least common when they are lowest. Therefore, the cost of the accommodation is solely responsible for the cancellation.

```
cancelled_data = df[df['is_canceled'] == 1]
top_10_country = cancelled_data['country'].value_counts()[:10]
plt.figure(figsize = (10,5))
plt.title('Top 10 Countries with Reservations Canceled')
plt.pie(top_10_country, labels = top_10_country.index, autopct =
'%.2f')
plt.show()
```

Top 10 Countries with Reservations Canceled



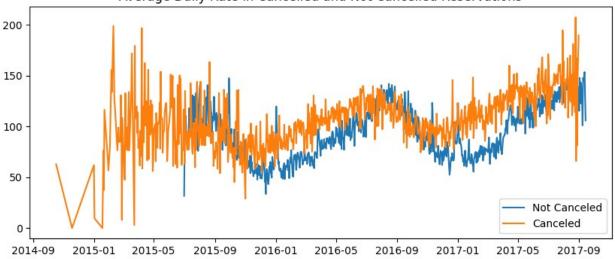
Now, let's see which country has the highest reservation canceled. The top country is Portugal with the highest number of cancellations.

Let's check the area from where guest are visiting the hotels and making reservations. Is it coming from Direct, Group, Online, Offline Travel Agents? Around 46% of the clients come from online travel agencies, whereas 27% come from groups. Only 4% of clients book hotels directly by visiting them and making reservations.

```
df['market segment'].value counts()
market_segment
Online TA
                 53786
Offline TA/TO
                 24111
                 19712
Groups
Direct
                 11717
                  5289
Corporate
Complementary
                   743
Aviation
                   237
Undefined
Name: count, dtype: int64
df['market segment'].value counts(normalize=True)
market segment
Online TA
                 0.465289
```

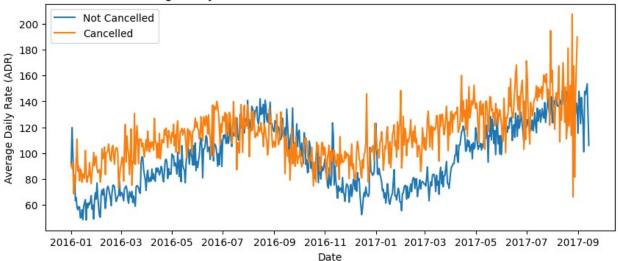
```
Offline TA/TO
                 0.208578
Groups
                 0.170523
Direct
                 0.101361
Corporate
                 0.045754
Complementary
                 0.006428
Aviation
                 0.002050
Undefined
                 0.000017
Name: proportion, dtype: float64
cancelled data['market segment'].value counts(normalize=True)
market segment
Online TA
                 0.456321
Groups
                 0.282049
Offline TA/TO
                 0.193919
Direct
                 0.041024
Corporate
                 0.023155
Complementary
                 0.002269
Aviation
                 0.001216
Undefined
                 0.000047
Name: proportion, dtype: float64
cancelled df adr = cancelled data.groupby('reservation status date')
['adr'].mean()
cancelled df adr = cancelled df adr.reset index()
cancelled df adr.sort values('reservation status date', inplace =
True)
not_cancelled_df_adr = df[df['is canceled'] == 0]
not cancelled df adr =
not cancelled df adr.groupby('reservation status date')['adr'].mean()
not cancelled df adr = not cancelled df adr.reset index()
not cancelled df adr.sort values('reservation status date', inplace =
True)
plt.figure(figsize = (10,4))
plt.title('Average Daily Rate in Cancelled and Not Cancelled
Reservations')
plt.plot(not cancelled df adr['reservation status date'],
not cancelled df adr['adr'],label = 'Not Canceled')
plt.plot(cancelled df adr['reservation status date'],
cancelled df adr['adr'],label = 'Canceled')
plt.legend()
plt.show()
```

#### Average Daily Rate in Cancelled and Not Cancelled Reservations



```
filtered df = df[(df['reservation status date'] >= '2016-01-01') &
(df['reservation status date'] <= '2017-09-30')]</pre>
plt.figure(figsize=(10, 4))
plt.title('Average Daily Rate for Cancelled and Not Cancelled
Reservations')
# Filter data for the specified date range
cancelled df filtered =
cancelled_df_adr[(cancelled df adr['reservation status date'] >=
'2016-01-01') & (cancelled df adr['reservation status date'] <= '2017-
09-30')]
not cancelled df filtered =
not cancelled df adr[(not cancelled df adr['reservation status date']
>= '2016-01-01') & (not cancelled df adr['reservation status date'] <=
'2017-09-30')1
plt.plot(not_cancelled_df_filtered['reservation_status_date'],
not cancelled df filtered['adr'], label='Not Cancelled')
plt.plot(cancelled df filtered['reservation status date'],
cancelled df filtered['adr'], label='Cancelled')
plt.xlabel('Date')
plt.ylabel('Average Daily Rate (ADR)')
plt.legend()
plt.show()
```

Average Daily Rate for Cancelled and Not Cancelled Reservations



As seen in thhe graph, reservations are cancelled when the average daily rate is higher than when it is not canceled. It clearly proves all the above analysis, that the higher price leads to higher cancellation.

# Suggestions

- 1. Cancellation rates rise as the price does. In order to prvent cancellations of reservations, hotel could work on their pricing strategies and try to lower the rates for specific hotels based on locations. They can also provide some discounts to the consumers.
- 2.As the ratio of the cancellation and not cancellation of the resort hotel in higher the resort hotel than the city hotels. So the hotels should provide a reasonable discount on the room prices on weekends or on hoildays.
- 3.In the month of January, hotels can start campaigns or making with a reasonable amount to increase their revenue as the cancellation is the highest in this month.
- 4.they can also increase the quality of their hotels and their services mainly in Portugal to reduce the cancellation rate.