

# Hotel Reservation Cancelation



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# Introduction



#### Motivation:

• As Hotel we will overbook once we know our safe margin.

### Objective:

Try to predict the ones who will cancel reservations?

#### Goals:

 Try to find best algorithm and predict the which reservations will be canceled.



#### Data:

Hotel booking demand datasets: Kaggle

#### Tools:

- Database : Postgresql
- Data Cleaning : pandas, numpy
- Visualizations: matplotlib, seaborn
- ML: sklearn, K-fold
- Development Environment : Jupyter

#### Roadmap

- Data Fetch & Clean (Postgresql -> dataframe)
- Eda
- Feature Engineering
- Model Selection with K-Fold
- Model tuning



#### Feature Data Types:

- 1 x Category
- 4 x float64
- 1 x int32
- 23 x int64
- 8 x object

#### Reservation Info:

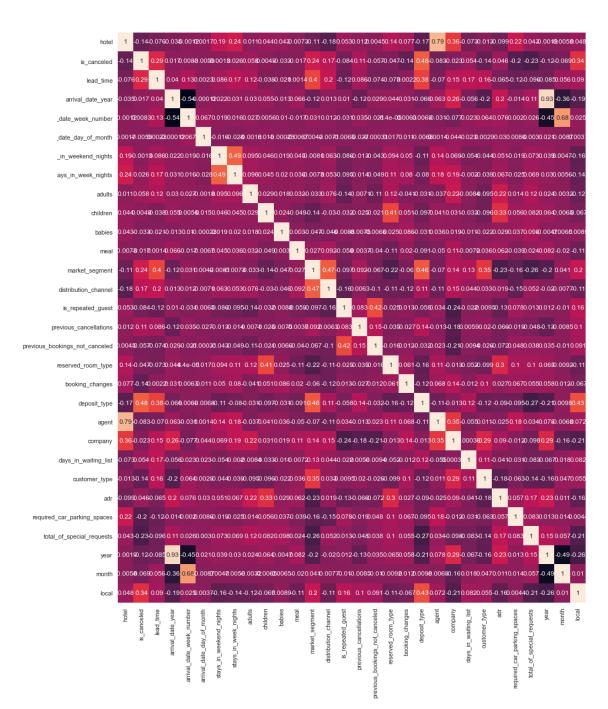
Lead time, Reservation date, Booking length, Weekday stay length, weekend stay length, Deposit, parking spot request, room type, booking channel, Room type, daily rate, total number of special requests, agent, Number of booking changes, number of days in waiting list

#### **Customer Info**

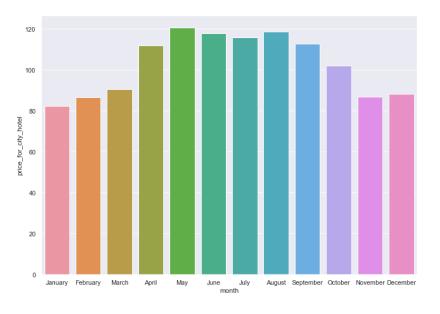
Previous cancelations, previous booking without cancel, repeated customer, number of adults, number of children, number of infants, business or individual,

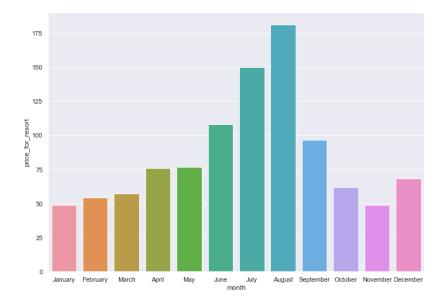
#### Hotel Info

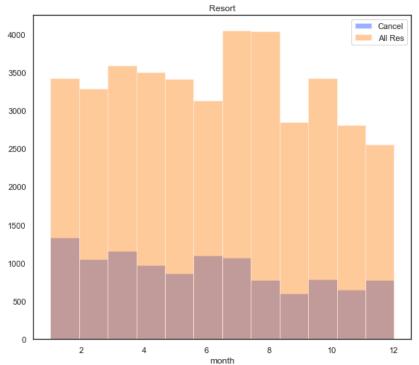
City or Resort, meal, parking space, average daily rate

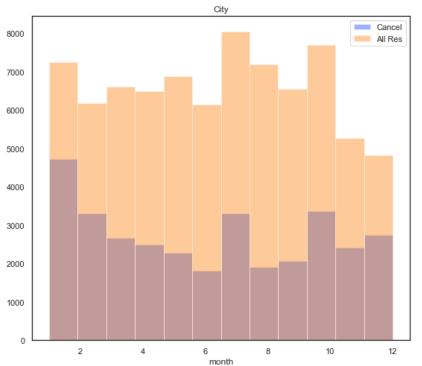


## **EDA**

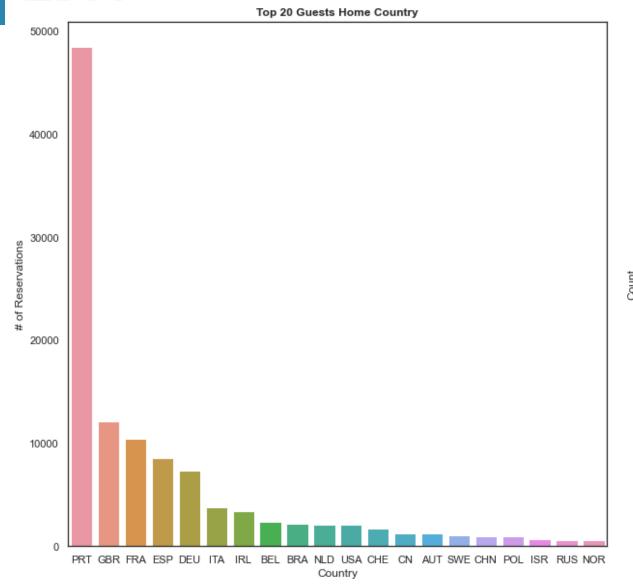


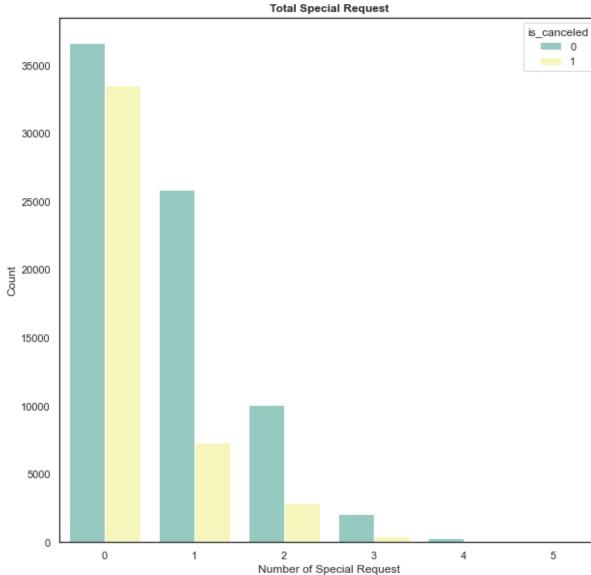






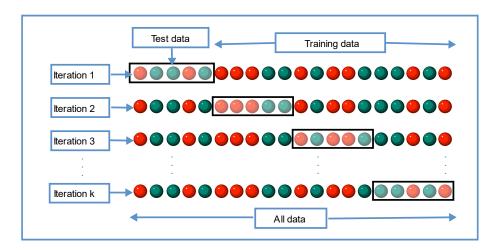






# **Model Selection with Cross Validation**

```
LogisticRegression() accuracy: : 0.773 (0.011) [0.75614462 0.77833235 0.77006963 0.78894388 0.77342505
LogisticRegression() precision: : 0.814 (0.015) [0.81814407 0.84099766 0.81167883 0.79842447 0.8009735
LogisticRegression() recall:
                              : 0.505 (0.044) [0.44385629 0.49129197 0.49988762 0.58104766 0.50829425]
RidgeClassifier() accuracy: : 0.849 (0.002) [0.8458183  0.85101921  0.85005453  0.8472863  0.85160641]
RidgeClassifier() precision: : 0.973 (0.001) [0.97325814 0.97512257 0.97214831 0.97282706 0.97360277]
                           : 0.610 (0.004) [0.60254533 0.61126921 0.61586873 0.60768885 0.61182931]
SGDClassifier() accuracy: : 0.613 (0.156) [0.41397534 0.78152001 0.74888852 0.43960238 0.68349971
SGDClassifier() precision: : 0.710 (0.263) [0.90264198 0.92813322 0.94332162 0.38684406 0.38974662]
                         : 0.449 (0.275) [0.29868228 0.3314741 0.30388851 0.99921313 0.31060519]
PassiveAggressiveClassifier() precision: : 0.820 (0.236) [0.98176718 0.36931746 0.95492289 0.99310777 0.80338308]
                                      : 0.541 (0.370) [0.10113752 0.38656801 0.96055293 0.2544964
PassiveAggressiveClassifier() recall:
KNeighborsClassifier() accuracy: : 0.815 (0.004) [0.80899253 0.81608087 0.81574532 0.81394178 0.81998155]
KNeighborsClassifier() precision: : 0.797 (0.004) [0.79234825 0.79372497 0.80320366 0.79559915 0.79935406]
KNeighborsClassifier() recall: : 0.672 (0.007) [0.66009686 0.67672168 0.67060013 0.67468525 0.67955611]
DecisionTreeClassifier() accuracy: : 0.920 (0.002) [0.91779213 0.91997316 0.91909236 0.91846322 0.92328664]
DecisionTreeClassifier() precision: : 0.886 (0.003) [0.89059522 0.88324421 0.88490399 0.88295053 0.88939154]
                                 : 0.899 (0.003) [0.8940196 0.90199203 0.9023376 0.89916817 0.8985242
ExtraTreeClassifier() accuracy: : 0.885 (0.004) [0.89044543 0.88176327 0.88461538 0.88780304 0.87870145]
ExtraTreeClassifier() precision: : 0.850 (0.006) [0.84004939 0.84635155 0.85808989 0.85482427 0.8495555
ExtraTreeClassifier() recall:
                               : 0.847 (0.008) [0.84243721 0.84712578 0.8409755 0.86207284 0.84429699]
LinearSVC() accuracy: : 0.761 (0.017) [0.78042949 0.75257948 0.74674943 0.78244275 0.74268098]
LinearSVC() precision: : 0.732 (0.294) [0.98410446 0.37096979 0.97516005 0.37417428 0.95551393]
                     : 0.763 (0.297) [0.99121523 0.27797382 0.55135986 0.99955036 0.99622469]
GaussianNB() accuracy: : 0.624 (0.007) [0.63182619 0.62339569 0.62020804 0.63014848 0.61265833
GaussianNB() precision: : 0.496 (0.007) [0.50335348 0.49369884 0.49490293 0.50254935 0.48424343
                      : 0.862 (0.006) [0.85369974 0.86511098 0.85659699 0.86420863 0.86843611
AdaBoostClassifier() accuracy: : 0.843 (0.002) [0.84376311 0.83977854 0.84468585 0.84380505 0.84334368]
AdaBoostClassifier() precision: : 0.849 (0.004) [0.84937636 0.84340849 0.85306511 0.85270049 0.84419692]
AdaBoostClassifier() recall:
                              BaggingClassifier() accuracy: : 0.943 (0.002) [0.94182535 0.94291586 0.94467746 0.94052512 0.9447194
BaggingClassifier() precision: : 0.956 (0.003) [0.95380103 0.95627214 0.95937122 0.95293406 0.95841584]
                             : 0.885 (0.004) [0.87847731 0.89038133 0.88783996 0.88399281 0.88605423]
RandomForestClassifier() accuracy: : 0.940 (0.001) [0.93943461 0.94077678 0.94044124 0.93809244 0.94245449]
RandomForestClassifier() precision: : 0.958 (0.002) [0.96074046 0.95650567 0.95936656 0.95440098 0.95938515]
RandomForestClassifier() recall:
                               : 0.879 (0.004) [0.87160716 0.88366534 0.87986064 0.87769784 0.88147809]
```



# **Optimize Decision Tree**

#### All Data

```
DecisionTreeClassifier() accuracy: 0.920 (0.002) [0.91779213 0.91997316 0.91909236 0.91846322 0.92328664]
DecisionTreeClassifier() precision: 0.886 (0.003) [0.89059522 0.88324421 0.88490399 0.88295053 0.88939154]
DecisionTreeClassifier() recall: 0.899 (0.003) [0.8940196 0.90199203 0.9023376 0.89916817 0.8985242 ]
```

#### Resort type Hotel

```
DecisionTreeClassifier() accuracy: 0.921 (0.005) [0.92846442 0.91822722 0.92421026 0.91909102 0.91497066] DecisionTreeClassifier() precision: 0.850 (0.008) [0.85932722 0.85733573 0.85065502 0.84492689 0.8380035 ] DecisionTreeClassifier() recall: : 0.866 (0.012) [0.88124157 0.84738956 0.87567568 0.86857663 0.85873938]
```

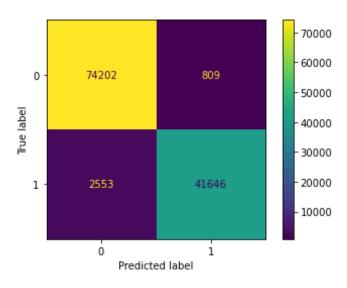
#### City type Hotel

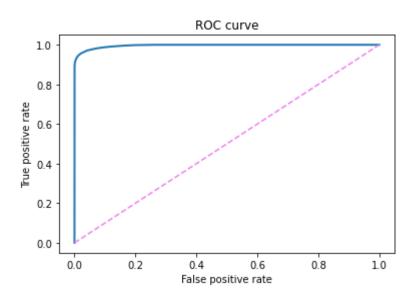
```
DecisionTreeClassifier() accuracy: 0.918 (0.003) [0.91745089 0.92174572 0.91302975 0.91668772 0.91883527]
DecisionTreeClassifier() precision:0.895 (0.007) [0.8968254 0.90344515 0.88404508 0.89929777 0.89347696]
DecisionTreeClassifier() recall: : 0.908 (0.003) [0.91107458 0.9065534 0.9089525 0.90272782 0.91076876]
```

```
recall for max depth= 2 0.3279
recall for max depth= 3 0.5913
recall for max depth= 4 0.3733
recall for max depth= 5 0.7946
recall for max depth= 6 0.7620
recall for max depth= 7 0.7870
recall for max depth= 8 0.7517
recall for max depth= 9 0.7616
recall for max depth= 10 0.7798
recall for max depth= 11 0.8102
recall for max depth= 12 0.8294
recall for max depth= 13 0.8413
recall for max depth= 14 0.8438
recall for max depth= 15 0.8586
recall for max depth= 16 0.8647
recall for max depth= 17 0.8690
recall for max depth= 18 0.8762
recall for max depth= 19 0.8806
recall for max depth= 20 0.8815
recall for max depth= 21 0.8871
recall for max depth= 22 0.8872
recall for max depth= 23 0.8919
recall for max depth= 24 0.8931
recall for max depth= 25 0.8947
recall for max depth= 26 0.8960
recall for max depth= 27 0.8970
recall for max depth= 28 0.8973
recall for max depth= 29 0.8983
recall for max depth= 30 0.8975
recall for max depth= 31 0.8980
recall for max depth= 32 0.8973
recall for max depth= 33 0.8982
recall for max depth= 34 0.8978
recall for max depth= 35 0.8987
recall for max depth= 36 0.8985
recall for max depth= 37 0.8989
recall for max_depth= 38 0.8992
recall for max depth= 39 0.8987
```

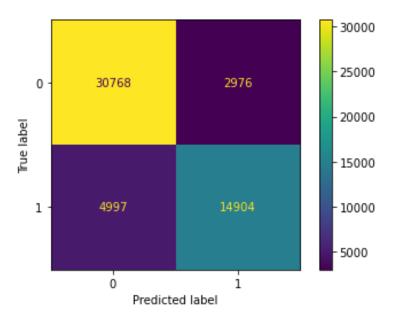
# **Decision Tree Results**

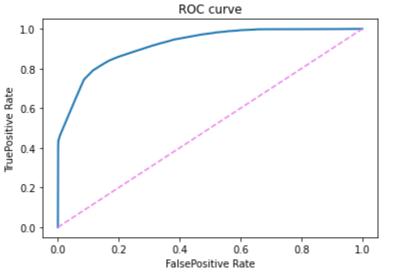
model.fit(X,y)





model.fit(X\_test,y\_test) Test Data Size: %0.45





# **Future Work**

- A web application for hotel staff
- Different models for different sessions
- Different models for different personas

