Predicting Chances of booking cancellation using Logistics Regression

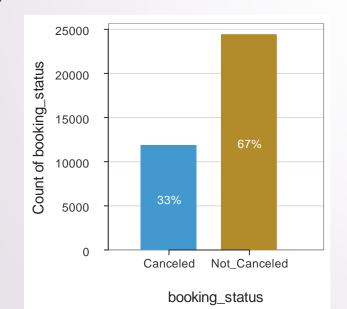
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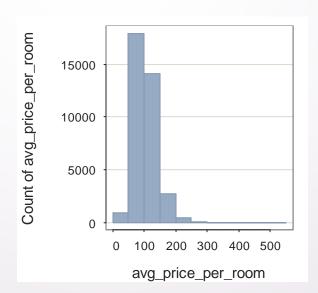
Deepshi Sachan

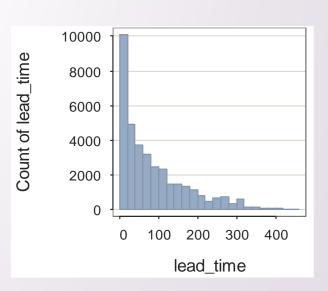
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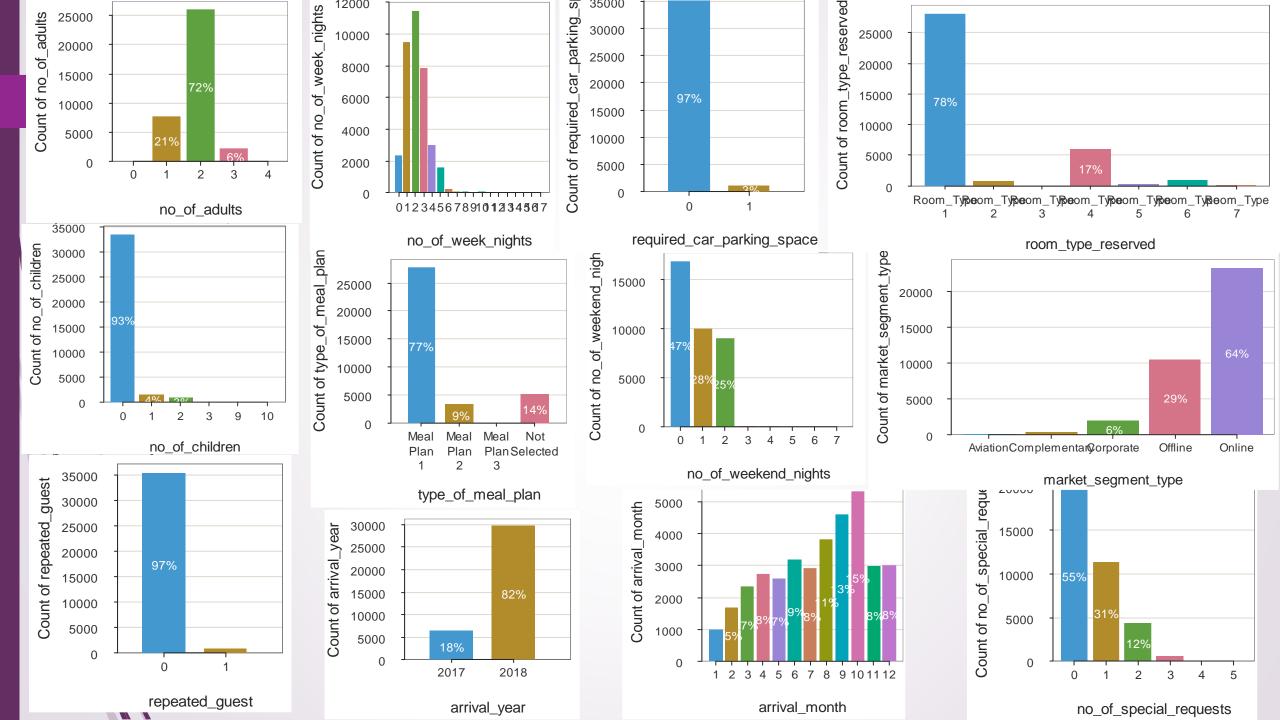
ABOUT DATASETS

- Source of this dataset is Kaggle.com website. (Hotel_reservations.csv)
- This dataset contains 36725 observations and 19 variables.
- Contains different attributes of hotel reservations like type of room reserved by the customer, Number of weekend nights, type of meal plan booked, etc.
- Used Logistic Regression model to forecast the chances of booking cancellation.
- ► Transformed Categorical variable (booking status) to Numerical. (0- Not Cancelled, 1- Cancelled)

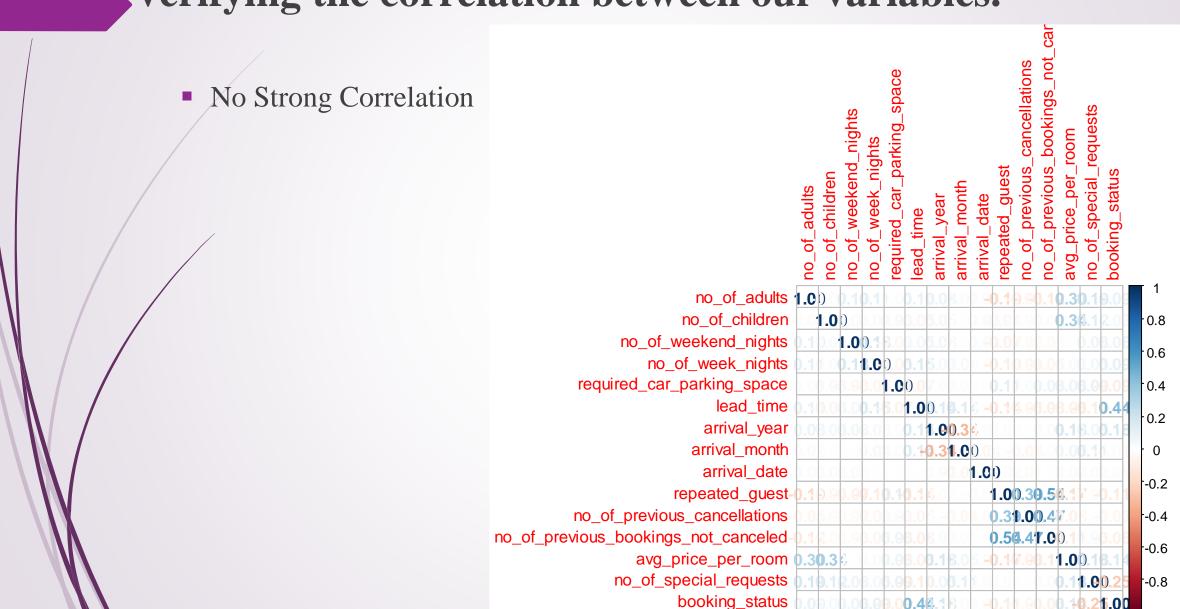






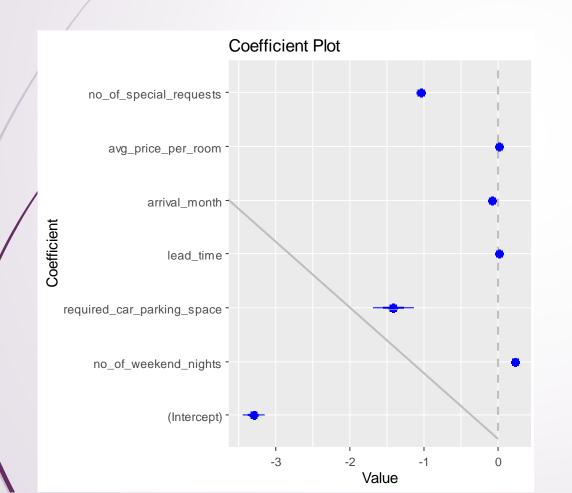


Verifying the correlation between our variables.



Logistic Regression Model

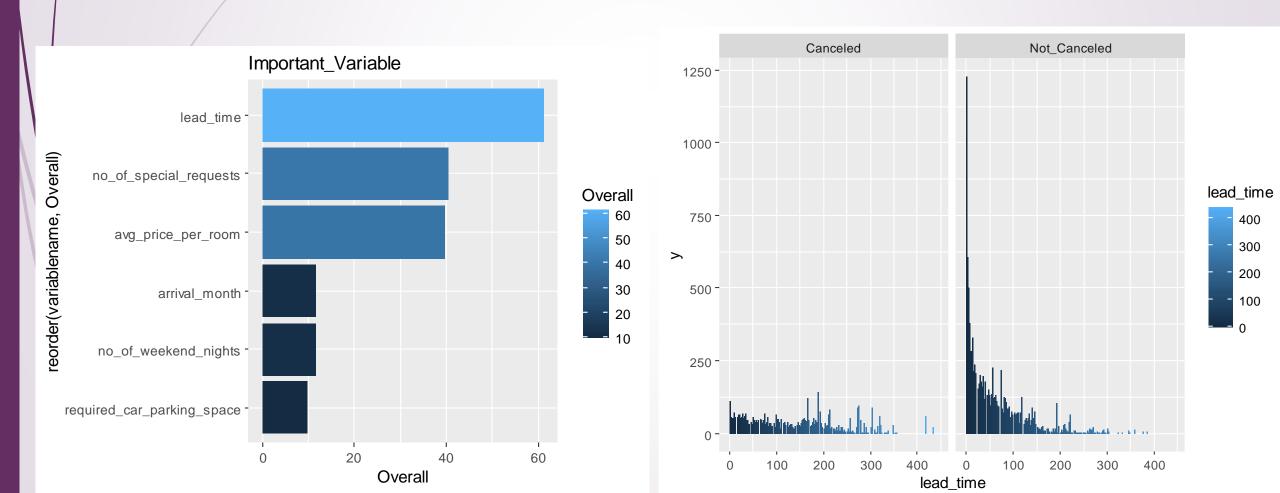
- Ratio of training to test dataset is 7:3.
- Model trained on the train dataset.
- Model tested on test dataset and results are shown below.



Confusion Matrix and Statistics Reference Prediction 0 6554 779 1 1405 2200 Accuracy : 0.8003 95% CI: (0.7927, 0.8078) No Information Rate: 0.7276 P-Value [Acc > NIR] : < 0.0000000000000022 Kappa : 0.5273 Mcnemar's Test P-Value : < 0.00000000000000022 Sensitivity: 0.8235 Specificity: 0.7385 Pos Pred Value: 0.8938 Neg Pred Value: 0.6103 Prevalence: 0.7276 Detection Rate: 0.5992 Detection Prevalence: 0.6704 Balanced Accuracy: 0.7810 'Positive' Class: 0

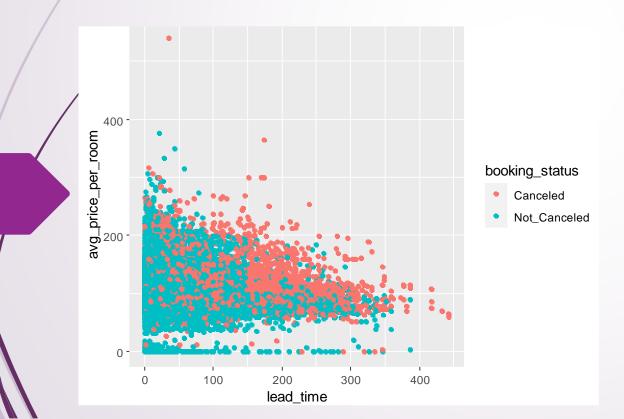
Checking the most important variables of the model

■ Lead_time is by far the most important predictor variable, followed by no_of_special_request, and average price per room.

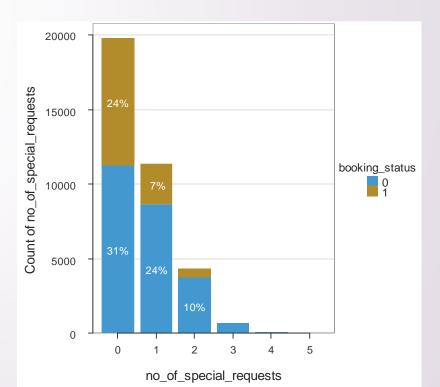


CONCLUSION

- Model Accuracy is 80.03%.
- ☐ Most important variable for predicting is Lead_time, followed by Average price per room and no. of special guest.



(0 = Not Cancelled, 1 = Cancelled)



Thank You

APPENDIX

Data Dictionary

- Booking_ID: unique identifier of each booking
- no_of_adults: Number of adults
- no_of_children: Number of Children
- no_of_weekend_nights: Number of weekend nights (Saturday or Sunday) the guest stayed or booked to stay at the hotel
- no_of_week_nights: Number of weeknights (Monday to Friday) the guest stayed or booked to stay at the hotel
- type_of_meal_plan: Type of meal plan booked by the customer:
- required_car_parking_space: Does the customer require a car parking space? (0 No, 1- Yes)
- room_type_reserved: Type of room reserved by the customer. The values are ciphered (encoded) by INN Hotels.
- lead_time: Number of days between the date of booking and the arrival date
- arrival_year: Year of arrival date
- arrival month: Month of arrival date
- arrival_date: Date of the month
- market_segment_type: Market segment designation.
- repeated_guest: Is the customer a repeated guest? (0 No, 1- Yes)
- no_of_previous_cancellations: Number of previous bookings that were canceled by the customer prior to the current booking
- no_of_previous_bookings_not_canceled: Number of previous bookings not canceled by the customer prior to the current booking
- avg_price_per_room: Average price per day of the reservation; prices of the rooms are dynamic. (in euros)
- no_of_special_requests: Total number of special requests made by the customer (e.g. high floor, view from the room, etc)
- **b**ooking_status: Flag indicating if the booking was canceled or not.