

# **AGENDA**







# **Background**

Background Concepts

Kaggle



# BACKGROUND



# How many analytics terms can you think of that are synonymous with "HR Analytics"?



# BACKGROUND

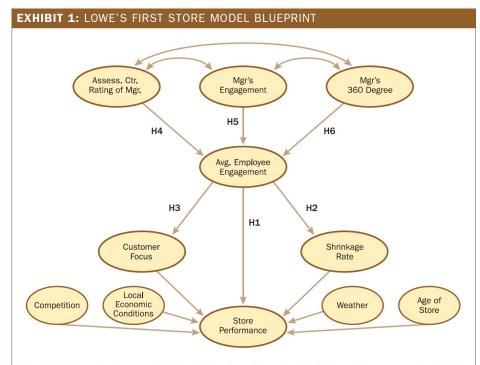


HR Analytics
Workforce Analytics
People Analytics
Talent Analytics
Human Capital Management



# **BACKGROUND**





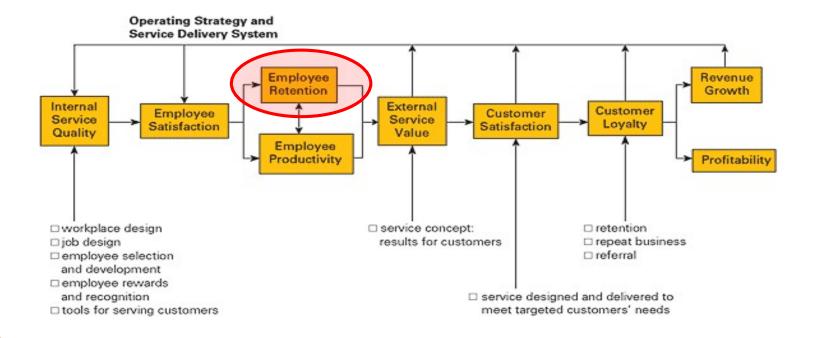
The initial Lowe's store model blueprint and roadmap for the final models represented initial hypotheses from the key stakeholders about how the data would interact in the model.



# **BACKGROUND**



### The Links in the Service-Profit Chain





# **Code and Concepts**

Background Concepts

Kaggle



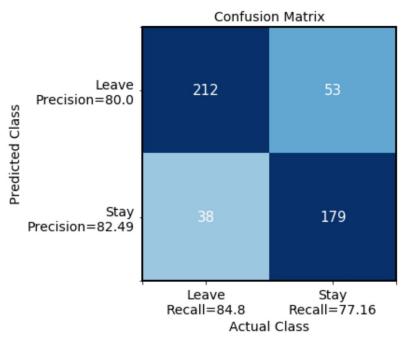
# **CODE & CONCEPTS**

Confusion Matrix
Feature Importance
Loading the Hilton Data (task, DV, and variables)
Data Splitting
Decision Tree Visual
Grid Search
XGBoost
ROC Curve
ROC Variance



# **CODE & CONCEPTS - CONFUSION MATRIX**

```
def displayConfusionMatrix(confusionMatrix, precisionNegative, precisionPositive, recallNegative, recallPositive, ti
         # Set font size for the plots. You can ignore this line.
         PLOT_FONT_SIZE = 14
         # Set plot size. Please ignore this line
         plt.rcParams['figure.figsize'] = [5, 5]
         # Transpose of confusion matrix to align the plot with the actual precision recall values. Please ignore this as
         confusionMatrix = np.transpose(confusionMatrix)
         # Plotting the confusion matrix
         plt.imshow(confusionMatrix, interpolation='nearest',cmap=plt.cm.Blues, vmin=0, vmax=100)
         # Setting plot properties. You should ignore everything from here on.
         xticks = np.array([-0.5, 0, 1, 1.5])
         plt.gca().set xticks(xticks)
         plt.gca().set yticks(xticks)
         plt.gca().set xticklabels(["", "Leave\nRecall=" + str(recallNegative) , "Stay\nRecall=" + str(recallPositive), "
         plt.gca().set_yticklabels(["", "Leave\nPrecision=" + str(precisionNegative) , "Stay\nPrecision=" + str(precision=" + str
         plt.ylabel("Predicted Class", fontsize=PLOT FONT SIZE)
         plt.xlabel("Actual Class", fontsize=PLOT_FONT_SIZE)
         plt.title(title, fontsize=PLOT_FONT_SIZE)
         # Add text in heatmap boxes
         for i in range(2):
                   for j in range(2):
                             text = plt.text(j, i, confusionMatrix[i][j], ha="center", va="center", color="white", size=15) ### size
         plt.show()
```





# **CODE & CONCEPTS - LOADING THE HILTON DATA**

### 4.1 Data Loading with Pandas

```
[7]: # Read data into a data frame
data = pd.DataFrame(pd.read_csv("Data/HiltonPredictionData_Train.csv", ","))

# Delete null values
data = data.dropna()

# Delete columns for dependent variables
data = data.drop(columns=['Engagement', 'JobSatisfaction', 'PaySatisfaction', 'RecommendToWork', 'RecommendFriendsToStay'])

# Get column names. We will use these for visualization purposes
columns = list(data.columns)

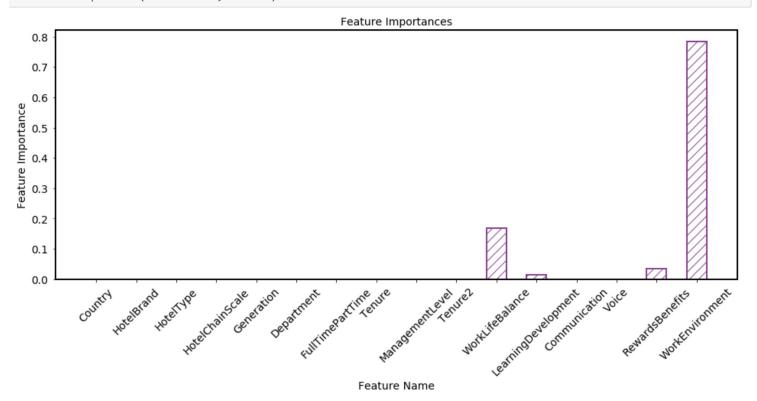
# Display the data frame as a table
display(data)
```

	AnonymousID	HotelInncode	Country	HotelBrand	HotelType	HotelChainScale	Generation	Department	FullTimePartTime	Tenure	Manageme
0	2	100038	1	4	2	3	2	20	1	4	
1	3	100038	1	4	2	3	1	16	1	2	
2	4	100038	1	4	2	3	1	16	1	1	
3	5	100038	1	4	2	3	1	16	1	2	
4	7	100038	1	4	2	3	2	20	1	4	



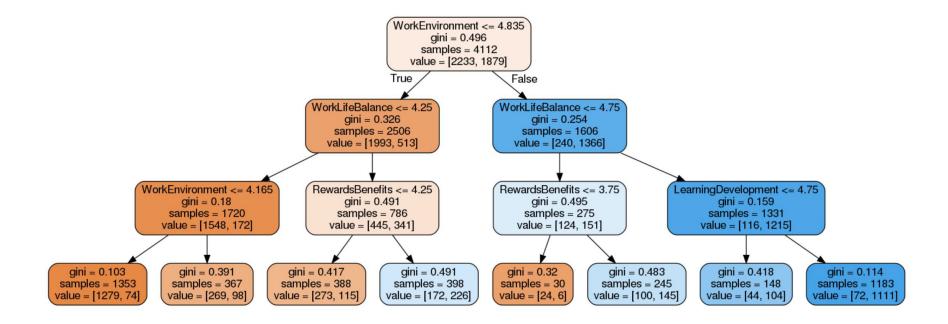
# **CODE & CONCEPTS - FEATURE IMPORTANCE**

# Calculate feature importance
showFeatureImportance(classifierDt, columns)





# **CODE & CONCEPTS - DECISION TREE VISUAL**





# **CODE & CONCEPTS - DATA SPLIT**

### 4.3 Data Splitting

```
[32]: # Test data percentage (0.1 = 10%)
TEST_DATA_PERCENTAGE = 0.1

# Validation data percentage
VALIDATION_DATA_PERCENTAGE = 0.05

# Split into train and test
trainData, testData, trainLabels, testLabels = train_test_split(features, labels, test_size=TEST_DATA_PERCENTAGE)
```

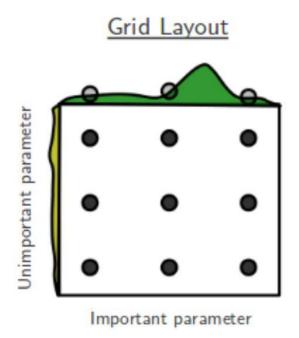
Training Data (used to build the model)

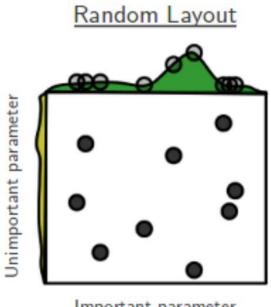
Validation Data
(this is called test data in the notebook)

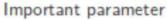
Test Data (this is the Kaggle dataset: Section 5.6)



# **CODE & CONCEPTS - GRID SEARCH**









# CODE & CONCEPTS - GRID SEARCH

[40]: ###### We will repeat the same process we did in the previous DecisionTreeClassifier cell but we'll now wrap our cla
# Create an instance of the decision tree algorithm. You can consider "classifier" as a function that is going to ge
classifierDtGrid = GridSearchCV(DecisionTreeClassifier(), parameters)

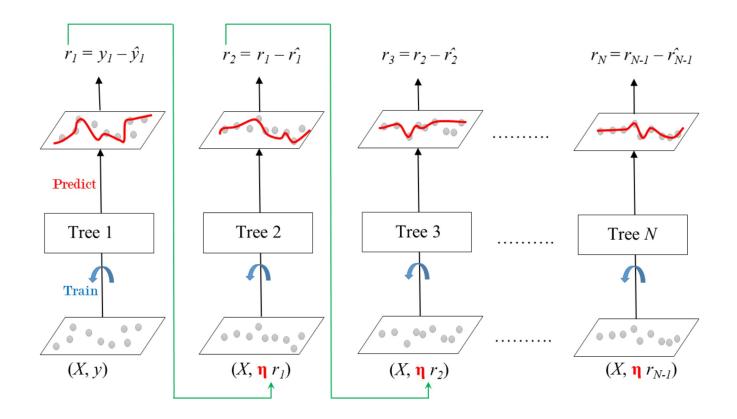
# Train the decision tree model using the function .fit
classifierDtGrid.fit(trainData, trainLabels)

# Calculate training accuracy of the classifier
trainAccuracy = classifierDtGrid.score(trainData, trainLabels)

# Predict on test data
predictions = classifierDtGrid.predict(testData) # This will give binary labels e.g 0/1
predictionProbabilities = classifierDtGrid.predict\_proba(testData) # This will give prediction probabilities for bot



# **CODE & CONCEPTS - XGBOOST**





# **CODE & CONCEPTS - XGBOOST**

### 5.3.1 Grid Search on XgBoost

XgBoost has so many parameters that you can tune. Let us try to do a grid search on those parameters to get the best model possible. The list of all parameters is listed here:

https://scikit-learn.org/stable/modules/generated/sklearn.ensemble.GradientBoostingClassifier.html



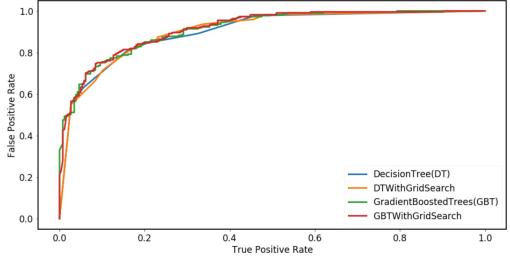
# **CODE & CONCEPTS - ROC CURVE**

### 5.4 ROC Curve

```
plt.rcParams['figure.figsize'] = [16, 8]
FONT_SIZE = 17

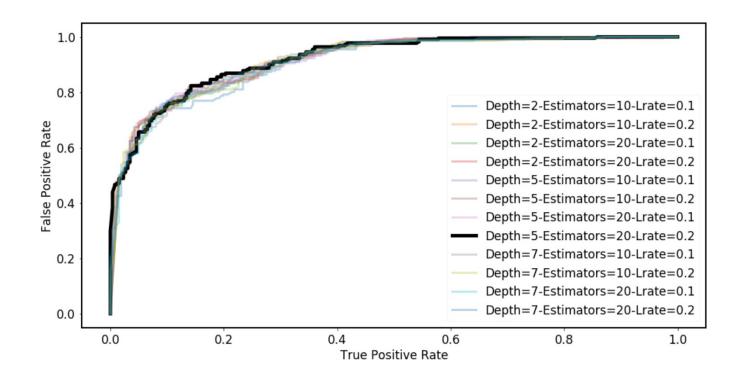
plt.plot(falsePositiveRateDt, truePositiveRateDt, linewidth = 3, label = "DecisionTree(DT)")
plt.plot(falsePositiveRateDtWithGridSearch, truePositiveRateDtWithGridSearch, linewidth = 3, label = "DTWithGridSearch")
plt.plot(falsePositiveRateGb, truePositiveRateGb, linewidth = 3, label = "GradientBoostedTrees(GBT)")
plt.plot(falsePositiveRateGbWithGridSearch, truePositiveRateGbWithGridSearch, linewidth = 3, label = "GBTWithGridSearch")

plt.legend(fontsize=FONT_SIZE)
plt.xicks(fontsize=FONT_SIZE)
plt.yticks(fontsize=FONT_SIZE)
plt.xlabel("True Positive Rate", fontsize=FONT_SIZE)
plt.ylabel("False Positive Rate", fontsize=FONT_SIZE)
plt.show()
```





# **CODE & CONCEPTS - ROC VARIANCE**





# **SUBMISSION**

```
resultsFile = open("Results/dtPredictions.csv", "w")
resultsFile.write("Id,Prediction\n")
for predictionProb, i in zip(classifierDt.predict proba(testFeaturesForKaggle), np.arange(len(testFeaturesForKaggle))):
    resultsFile.write(str(i + 1) + "," + str(float(predictionProb[1])) + "\n")
resultsFile.close()
resultsFile = open("Results/dtPredictionsGrid.csv", "w")
resultsFile.write("Id,Prediction\n")
for predictionProb, i in zip(classifierDtGrid.predict proba(testFeaturesForKaggle), np.arange(len(testFeaturesForKaggle))
    resultsFile.write(str(i + 1) + "," + str(float(predictionProb[1])) + "\n")
resultsFile.close()
resultsFile = open("Results/gbPredictions.csv", "w")
resultsFile.write("Id, Prediction\n")
for predictionProb, i in zip(gbClassifier.predict proba(testFeaturesForKaggle), np.arange(len(testFeaturesForKaggle))):
    resultsFile.write(str(i + 1) + "," + str(float(predictionProb[1])) + "\n")
resultsFile.close()
resultsFile = open("Results/gbPredictionsGrid.csv", "w")
resultsFile.write("Id,Prediction\n")
for predictionProb, i in zip(gbClassifierWithGridSearch.predict proba(testFeaturesForKaggle), np.arange(len(testFeature
    resultsFile.write(str(i + 1) + "," + str(float(predictionProb[1])) + "\n")
resultsFile.close()
print("All predictions have been placed in the results folder...")
```



