

MD2201 Data Science Course Project

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1. Project Title: Campus Recruitment Analysis

2. Data Set Name: Campus Recruitment

3. Data set Link: https://www.kaggle.com/datasets/niki188/campus-recruitment

- **4. Objective:** The objective of using Machine Learning is to discover patterns in the chosen Dataset "Campus Recruitment" and make the prediction based on these intricate patterns for answering the placed and Not placed applicants analysis. with the help of suitable machine learning algorithms analyzing the data as well as identifying trends and status of the applicant.
- **5. Project Description:** Machine learning uses programmed algorithms that receive and analyze input data to predict output values within an acceptable range. As new data is fed to these algorithms, they learn and optimize their operations to improve performance, developing 'intelligence' over time. The dataset chosen for the analysis is "Campus Recruitment". The algorithms like Logistic regression, Random Forest are used for Prediction of

'placed' and 'Not placed' Applicants. Logistic regression is commonly used for prediction and classification problems Logistic regression is a classification algorithm. It is used to predict a binary outcome based on a set of independent variables. A binary outcome is one where there are only two possible scenarios—either the event happens (1) or it does not happen (0). Here for this dataset Placed and Not placed is considered. Independent variables are those variables or factors which may influence the outcome (or dependent variable).

Second Algorithm implemented on the dataset is "Random Forest", It is a classifier that contains a number of decision trees on various subsets of the given dataset and takes the average to improve the predictive accuracy of that dataset."

In Addition the next Algorithm used is SVM it is a support vector machines are a set of supervised learning methods used for classification, regression, and outliers detection, SVMs are different from other classification algorithms because of the way they choose the decision boundary that maximizes the distance from the nearest data points of all the classes. The decision boundary created by SVMs is called the maximum margin classifier or the maximum margin hyper plane.

6. Code:

f<- read.csv("Placement_Data_Full_Class.csv")

f <- f[,-15]

f < -f[,-1]



```
f$gender = as.factor(f$gender)
f$ssc b = as.factor(f$ssc b)
f$hsc s = as.factor(f$hsc s)
f$hsc b = as.factor(f$hsc b)
f$degree t = as.factor(f$degree t)
f$workex = as.factor(f$workex)
f$specialisation = as.factor(f$specialisation)
f$status = as.factor(f$status)
library("caret")
library("ROSE")
f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = "both", p = 0.5, seed = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, method = f \leftarrow ovun.sample(status \sim ., data = f, meth
222)$data
set.seed(123)
a <- sample(2, nrow(f), replace = TRUE, prob = c(0.75, 0.25))
traning <- f[a==1,]
test <- f[a==2,]
#Logistic Regression----
cat("\n Logistic Regression \n")
library(nnet)
lr <- multinom(status~., data = traning)</pre>
p1 <- predict(lr, test)</pre>
confusionMatrix(p1, test$status)
```



```
#Random Forest----
cat("\n Random Forest \n")
library(randomForest)
rf <- randomForest(status~., data = traning)
f1 <- predict(rf, test)
confusionMatrix(f1, test$status)
#Support Vector Machine----
cat("\n Support Vector Machine \n")
library("e1071")
svm <- svm(status~.,data = traning)
k1 <- predict(svm, test)
confusionMatrix(k1, test$status)
```

7. Results: Quantitative findings and Plots.

Confusion Matrices for Logistic Regression, Random Forest & SVM



Logistic Regression

> confusionMatrix(p1, test\$status)

Confusion Matrix and Statistics

Reference

Prediction Placed Not Placed Placed 2330 331 Not Placed 280 2319

Accuracy : 0.8838

95% CI: (0.8749, 0.8924)

No Information Rate: 0.5038 P-Value [Acc > NIR]: <2e-16

Kappa: 0.7677

Mcnemar's Test P-Value : 0.0431

Sensitivity: 0.8927 Specificity: 0.8751

Pos Pred Value : 0.8756 Neg Pred Value : 0.8923

Prevalence : 0.4962

Detection Rate: 0.4430 Detection Prevalence: 0.5059

Balanced Accuracy: 0.8839

'Positive' Class : Placed



Random Forest

confusionMatrix(f1, test\$status)

Confusion Matrix and Statistics

Reference

Prediction Placed Not Placed

Placed 2610 C

Not Placed 0 2650

Accuracy: 1

95% CI: (0.9993, 1)

No Information Rate : 0.5038

P-Value [Acc > NIR] : < 2.2e-16

Kappa: 1

Mcnemar's Test P-Value : NA

Sensitivity : 1.0000

Specificity: 1.0000

Pos Pred Value: 1.0000

Neg Pred Value: 1.0000

Prevalence: 0.4962

Detection Rate: 0.4962

Detection Prevalence: 0.4962

Balanced Accuracy: 1.0000

'Positive' Class : Placed



Support Vector Machine

> confusionMatrix(k1, test\$status)

Confusion Matrix and Statistics

Reference

Prediction Placed Not Placed Placed 2598 83 Not Placed 12 2567

Accuracy : 0.9819

95% CI: (0.978, 0.9854)

No Information Rate : 0.5038 P-Value [Acc > NIR] : < 2.2e-16

Kappa: 0.9639

Mcnemar's Test P-Value: 6.878e-13

Sensitivity: 0.9954

Specificity: 0.9687

Pos Pred Value : 0.9690

Neg Pred Value : 0.9953

Prevalence: 0.4962

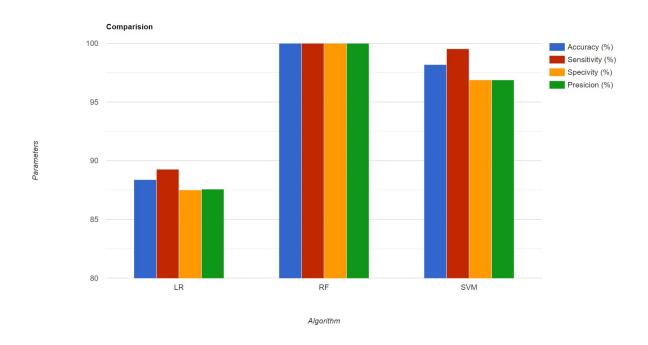
Detection Rate: 0.4939 Detection Prevalence: 0.5097

Balanced Accuracy : 0.9820

'Positive' Class : Placed

Bar Graph

		Algorithm			
		LR	RF	SVM	
	Accuracy (%)	88.38	100	98.19	
	Sensitivity (%)	89.27	100	99.54	
	Specivity (%)	87.51	100	96.87	
Parameters	Precision (%)	87.56	100	96.9	





8. Conclusions:

Machine learning is a subset of AI that leverages algorithms to analyze vast amounts of data. The algorithms like Logistic Regression, Random Forest, and SVM are used to make a prediction from "placement Analysis" data. The most suitable algorithm found is Random forest then SVM and lastly Logistic regression on the basis of parameters like Accuracy, sensitivity, precision, specivity.