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| Assignment\_17.1#1. Use the below given data set |
|  | #Data Set |
|  | #2. Perform the below given activities: |
|  | #a. Create classification model using logistic regression model |
|  |  |
|  | #using dataset cs2m |
|  | #reading the dataset |
|  | cs2m <-read.csv("D:\\BIG DATA\\DATA ANALYTICS WITH R, EXCEL & TABLEAU\\17 ENSEMBLE MODELS\\cs2m.csv") |
|  | View(cs2m) |
|  |  |
|  | #logistic regression |
|  | model<-glm(classe~cvtd\_timestamp+total\_accel\_belt+yaw\_dumbbell+roll\_forearm+accel\_forearm\_y, data= cs2m ,family= binomial(link='logit')) |
|  | model |
|  | summary(model) |
|  |  |
|  | #classification |
|  | library(caTools) |
|  | library(tree) |
|  | #splitting |
|  | set.seed(1) |
|  | split<-sample.split(cs2m$classe,SplitRatio=0.70) |
|  | cs2mTrain <-subset(cs2m,split ==TRUE) |
|  | cs2mTest<-subset(cs2m, split ==FALSE) |
|  |  |
|  | modelClassTree<-tree(classe~cvtd\_timestamp+total\_accel\_belt+yaw\_dumbbell+roll\_forearm+accel\_forearm\_y,data= cs2mTrain) |
|  | plot(modelClassTree) |
|  |  |
|  | text(modelClassTree,pretty=0 ,cex=0.75) |
|  | pred<-predict(modelClassTree,newdata= cs2mTest) |
|  |  |
|  | predict<- predict(model,type="response") |
|  | head(predict,3) |
|  | cs2m$predict <- predict |
|  | cs2m$predictROUND<-round(predict,digits=0) |
|  | #confusion matrix |
|  | table(cs2m$classe,predict>=0.5) |
|  |  |
|  | sum<- sum(table(cs2m$classe,predict>=0.5)) |
|  |  |
|  | #b. verify model goodness of fit |
|  | #c. Report the accuracy measures |
|  | #f. Interpret the results |
|  |  |
|  | #Answer for b & c & f |
|  |  |
|  |  |
|  | #interpretation, Accuracy and model goodness of our model |
|  | summary(model) |
|  |  |
|  | #accuracy of our model |
|  | accuracy<- (1185+679)/(2266) |
|  | accuracy |
|  | #0.8225949 |
|  |  |
|  | library(verification) |
|  | library(AUC) |
|  |  |
|  | predictTrain<- predict(model,cs2m,type="response") |
|  | table(cs2m$classe,predictTrain >=0.5) |
|  | head(predictTrain,3) |
|  | auc(cs2m$classe,predictTrain) |
|  |  |
|  | #model goodness |
|  | #\*\*\*\*NOTE\*\*\*\* |
|  | #Area under the curve: 0.9333333 |
|  | #also our AIC is less which is measure of good model |
|  | #NULL deviance is also less which is good for model |
|  | #Residual deviance is also less model |
|  | #by this all things we conclude that our model is good and fit |
|  |  |
|  |  |
|  |  |
|  |  |
|  | #e. Report the unimportant variables |
|  |  |
|  | library(MASS) |
|  | step\_fit<-stepAIC(model,method="backward") |
|  | summary(step\_fit) |
|  | confint(step\_fit) |
|  | #thus by this method we get our best model and variable cvtd\_timestamp is not as much important y this method |
|  |  |
|  | #some test |
|  | #ANOVA on base model |
|  | anova(model,test='Chisq') |
|  | #ANOVA from reduced model after applying the Step AIC |
|  | anova(step\_fit,test='Chisq') |
|  |  |
|  | #check for multicollinearity |
|  | library(car) |
|  | vif(model) |
|  | vif(step\_fit) |
|  |  |
|  | #d. Report the variable importance |
|  |  |
|  | library(caret) |
|  | varImp(step\_fit) |
|  |  |
|  |  |
|  | #g. Visualize the results |
|  |  |
|  | #plot the fitted model |
|  | plot(model$fitted.values) |
|  |  |
|  | #plot glm |
|  | library(ggplot2) |
|  | ggplot(cs2mTrain, aes(x=yaw\_dumbbell, y=classe)) +geom\_point() + |
|  | stat\_smooth(method="glm", family="binomial", se=FALSE) |