Dr Tony Diana DATA 602 Introduction to Machine Learning How to Use R in Regression Analysis

Load R and then RStudio to work with R

```
head(cars) # display the first 6 observations scatter.smooth(x=carspeacespeed, y=carspeacedist, main="Dist ~ Speed") # scatterplot par(mfrow=c(1, 2)) # divide graph area in 2 columns
```

boxplot(cars\$speed, main="Speed", sub=paste("Outlier rows: ", boxplot.stats(cars\$speed)\$out)) # box plot for 'speed'

boxplot(cars\$dist, main="Distance", sub=paste("Outlier rows: ", boxplot.stats(cars\$dist)\$out)) # box plot for 'distance'

library(e1071) # for skewness function par(mfrow=c(1, 2)) # divide graph area in 2 columns

plot(density(cars\$speed), main="Density Plot: Speed", ylab="Frequency", sub=paste("Skewness:", round(e1071::skewness(cars\$speed), 2))) # density plot for 'speed'

polygon(density(cars\$speed), col="red")

plot(density(cars\$dist), main="Density Plot: Distance", ylab="Frequency", sub=paste("Skewness:", round(e1071::skewness(cars\$dist), 2))) # density plot for 'dist'

polygon(density(cars\$dist), col="red")

cor(cars\$speed, cars\$dist) # calculate correlation between speed and distance

linearMod <- lm(dist ~ speed, data=cars) # build linear regression model on full data print(linearMod)

summary(linearMod) # model summary

capture model summary as an object modelSummary <- summary(linearMod)

model coefficients modelCoeffs <- modelSummary\$coefficients

```
# get beta estimate for speed
beta.estimate <- modelCoeffs["speed", "Estimate"]
# get std.error for speed
std.error <- modelCoeffs["speed", "Std. Error"]
# calc t statistic
t value <- beta.estimate/std.error
# calc p Value
p_value <- 2*pt(-abs(t_value), df=nrow(cars)-ncol(cars))
# fstatistic
f_statistic <- linearMod$fstatistic[1]</pre>
# parameters for model p-value calc
f <- summary(linearMod)$fstatistic
model_p <- pf(f[1], f[2], f[3], lower=FALSE)
AIC(linearMod)
BIC(linearMod)
# Create Training and Test data -
set.seed(100) # setting seed to reproduce results of random sampling
trainingRowIndex <- sample(1:nrow(cars), 0.8*nrow(cars)) # row indices for training
data
trainingData <- cars[trainingRowIndex, ] # model training data
testData <- cars[-trainingRowIndex,] # test data
# Build the model on training data
ImMod <- Im(dist ~ speed, data=trainingData) # build the model
distPred <- predict(ImMod, testData) # predict distance
summary (lmMod) # model summary
actuals_preds <- data.frame(cbind(actuals=testData$dist, predicteds=distPred)) #
make actuals predicteds dataframe.
correlation_accuracy <- cor(actuals_preds) # 82.7%
head(actuals preds)
# Min-Max Accuracy Calculation
```

min_max_accuracy <- mean(apply(actuals_preds, 1, min) / apply(actuals_preds, 1, max))

MAPE Calculation mape <- mean(abs((actuals_preds\$predicteds actuals_preds\$actuals))/actuals_preds\$actuals)

DMwR::regr.eval(actuals_preds\$actuals, actuals_preds\$predicteds) # different way to load a library and to run the regression

library(DAAG)

cvResults <- suppressWarnings(CVIm(df=cars, form.Im=dist ~ speed, m=5, dots=FALSE, seed=29, legend.pos="topleft", printit=FALSE, main="Small symbols are predicted values while bigger ones are actuals.")); # performs the CV attr(cvResults, 'ms')