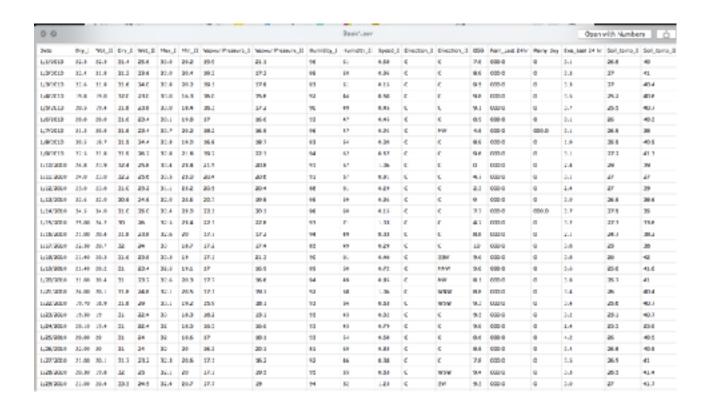
LAB EXAM: DATA ANALYSIS ON WEATHER DATA

Exploratory Data Analysis in R:

#Data



#Check dimensions (number of row & columns) in data set

>dim(data1)

[1] 455 20

#check the variables and their types in data

>str(data1)

```
data.frame':
                  455 obs. of 20 variables:
                     : Factor w/ 366 levels "","10/10/2010",...: 46 87 118 119 120 121 122 123 124 33 ...
                      : num 22.3 22.4 22.6 19.8 20.5 20.8 21.2 20.5 22.5 24.8 ...
 $ Dry_I
$ Wet_II
                      : num 22.3 21.8 21.8 19 19.4 20 20.8 19.7 21.8 23.9 ...
 $ Dry_I.1
                      : num 31.4 31.2 31.6 32 31.8 31.6 31.8 31.5 31.9 32.4 ...
                      : num 25.6 23.6 24 23 23.8 23.4 23.4 24.4 26.2 25.8 ...
 $ Wet_II.1
$ Max_I
                     : num 33 33 32.8 33 33 33.1 33.7 32.8 32.8 33.4 ...
                      : num 20.2 20.4 20.2 16.3 18.4 19.8 20.2 19 21.8 23.4 ...
$ Min_II
$ Vapour.Pressure_I : Factor w/ 87 levels "","1.2","14.3",...: 38 32 31 8 11 17 26 15 32 50 ... $ Vapour.Pressure_II: num 21.1 17.2 17.8 15.6 17.2 16.6 16.5 18.7 22.1 20.8 ...
 \mu = 1... $ Humidity_I : num 96 95 93 92 90 93 96 93 94 91 ...
$ Rain_Last.24hr : num 0 0 0 0 0 0 0 0 0 0 ...
$ Rainy.day : num 0 0 0 0 0 0 0 0 0 ...
$ Eva_Last.24.hr : num 3.1 3.2 3.3 3.5 3.7 3.1 3.1 1.9 3.1 2.4 ...
$ Soil_temp_I : Factor w/ 85 levels "","23","23.3",..: 28 30 30 14 16 20 25 16 32 46 ... $ Soil_temp_II : Factor w/ 138 levels "","24.6","25.5",..: 87 94 89 91 92 88 77 90 95 83 ...
```

#check if this data has missing values >table(is.na(data1))

```
FALSE TRUE
7835 1265
```

> pie(colSums(is.na(data1)))



#Column wise finding out the missing value

>colSums(is.na(data1))

```
Dry_I
         Date
           0
                       90
              Wet II
                            Dry_I.1
          90
                        90
      Wet_{II.1}
                       Max_I
          90
                        90
        Min_II Vapour.Pressure_I
          90
                        0
Vapour.Pressure_II
                        Humidity_I
          90
                        90
    Humidity_II
                       Speed_I
          90
                        93
    Direction I
                   Direction_II
           0
                        0
         BSS
                Rain_Last.24hr
          91
                        90
     Rainy.day
                  Eva_Last.24.hr
          90
                        91
    Soil_temp_I
                    Soil_temp_II
           0
                        0
```

#Full summary of each column in the dataset

> summary(train)

```
Date Dry_I Wet_II
: 90 Min. :19.50 Min. :18.00

10/10/2010: 1 1st Qu.:23.50 1st Qu.:23.00

10/11/2010: 1 Median :24.30 Median :23.80

10/1/2010: 1 Mean :24.59 Mean :23.54

10/12/2010: 1 3rd Qu.:26.00 3rd Qu.:24.50

10/13/2010: 1 Max. :30.20 Max. :28.00

(Other) :360 NA's :90 NA's :90
```

```
Dry_I.1 Wet_II.1 Max_I
Min. :23.00 Min. :22.40 Min. :25.20
1st Qu.:28.90 1st Qu.:24.80 1st Qu.:30.20
Median :30.50 Median :25.50 Median :31.80
Mean :30.24 Mean :25.69 Mean :31.71
3rd Qu.:32.00 3rd Qu.:26.50 3rd Qu.:33.50
```

Max. :35.00 Max. :35.00 Max. :36.20

NA's :90 NA's :90 NA's :90

 Min_{II} Vapour.Pressure_I

Min. :16.30 : 90 1st Qu.:22.50 22.4 : 18

Median :23.30 22 : 17

Mean :23.17 21.9 : 15 3rd Qu.:24.10 21.7 : 14

Max. :29.20 22.1 : 14

(Other):287 NA's :90

Vapour.Pressure_II Humidity_I

Min. :15.10 Min. : 56.00

1st Qu.:20.60 1st Qu.: 87.00 Median :22.10 Median: 93.00

Mean :21.93 Mean : 91.43 3rd Qu.: 96.00 3rd Qu.:23.30

Max. :29.50 Max. :100.00

NA's :90 NA's :90

Humidity II Speed I Direction I Min.: 44.00 Min.: 0.010 C :353 : 92 1st Qu.: 60.00 1st Qu.: 0.740 Median: 67.00 Median: 1.090 c Mean: 70.54 Mean: 1.275 ESE

3rd Qu.: 76.00 3rd Qu.:1.800 SSW : 2 Max. :663.00 Max. :3.800 NNE : 1

NA's :90 NA's :93 (Other): 3

Direction II BSS Rain_Last.24hr

C :164 Min. : 0.000 Min. : 0.00 : 95 1st Qu.: 1.975 1st Qu.: 0.00

SW : 47 Median : 6.050 Median : 0.00 : 42 Mean : 5.311 Mean : 11.57 NW

WSW : 29 3rd Qu.: 8.600 3rd Qu.: 9.60 SSW : 27 Max. :11.300 Max. :173.00

(Other): 51 NA's :91 NA's :90

Eva_Last.24.hr Soil_temp_I Rainy.day

Min. :0.0000 Min. :0.200 1st Qu.:0.0000 1st Qu.:2.400 27

Median: 0.0000 Median: 3.100 26

Mean :0.3781 Mean :3.287 25

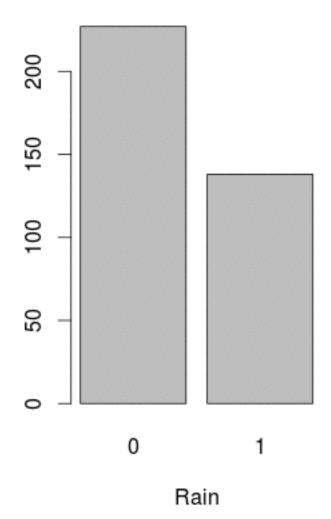
3rd Qu.:1.0000 3rd Qu.:4.200 26.5 : 23

Max. :1.0000 Max. :9.300 25.5 : 22

NA's :90 NA's :91 (Other):237

```
Soil temp II
    : 90
35
     : 13
47
     : 12
34
    : 11
36.5 : 11
36 : 9
(Other):309
#Dropping missing values
>data1=na.omit(data1)
#Summary after removing NA values.
>summary(data1)
    Date
             Dry I
                       Wet II
10/10/2010: 1 Min. :19.50 Min. :18.00
10/11/2010: 1 1st Qu.:23.50 1st Qu.:22.98
10/1/2010: 1 Median: 24.30 Median: 23.80
10/12/2010: 1 Mean :24.59 Mean :23.54
10/13/2010: 1 3rd Qu.:26.00 3rd Qu.:24.50
10/14/2010: 1 Max. :30.20 Max. :28.00
(Other) :354
             Wet_II.1
  Dry_I.1
                         Max I
Min. :23.00 Min. :22.40 Min. :25.2
Median: 30.50 Median: 25.50 Median: 31.8
Mean :30.24 Mean :25.69 Mean :31.7
3rd Qu.:32.00 3rd Qu.:26.50 3rd Qu.:33.5
Max. :35.00 Max. :35.00 Max. :36.2
           Vapour.Pressure I
  Min_{II}
Min. :16.30 22 : 17
1st Qu.:22.50 22.4 : 17
Median:23.30 21.9:15
Mean :23.16 21.7 : 14
3rd Qu.:24.12 22.1 : 14
Max. :29.20 20.8 : 13
         (Other):270
Vapour.Pressure_II Humidity_I
              Min. : 56.00
Min. :15.10
1st Qu.:20.60
               1st Qu.: 87.00
Median :22.10
                Median: 93.00
Mean :21.93
               Mean : 91.43
3rd Qu.:23.50
                3rd Qu.: 96.00
Max. :29.50
               Max. :100.00
 Humidity II
               Speed I
                          Direction_I
Min.: 44.00 Min.: 0.0100 C
1st Qu.: 60.00 1st Qu.: 0.7375 c
Median: 67.00 Median: 1.0750 ESE : 2
```

```
Mean: 70.57 Mean: 1.2716 SSW: 2
3rd Qu.: 76.00 3rd Qu.:1.8000 NNE : 1
Max. :663.00 Max. :3.8000 SE
                                  : 1
                    (Other): 2
               BSS
 Direction_II
                       Rain_Last.24hr
C
     :163 Min. : 0.000 Min. : 0.00
SW
           1st Qu.: 1.875 1st Qu.: 0.00
      : 46
NW
      : 41 Median : 6.000 Median : 0.00
WSW : 29 Mean : 5.294 Mean : 11.66
SSW : 27 3rd Qu.: 8.600 3rd Qu.: 9.95
WNW : 18 Max. :11.300 Max. :173.00
(Other): 36
             Eva_Last.24.hr Soil_temp_I
 Rainy.day
Min. :0.0000 Min. :0.200 27
                                : 31
1st Qu.:0.0000 1st Qu.:2.400 26
Median: 0.0000 Median: 3.100 25
                                    : 23
Mean :0.3778 Mean :3.283 26.5 : 23
3rd Qu.:1.0000 3rd Qu.:4.200 25.5 : 22
Max. :1.0000 Max. :9.300 28
                                 : 10
                    (Other):223
 Soil temp II
35
     : 13
47
     : 12
34
     : 11
36.5 : 11
     : 9
36
     : 9
37
(Other):295
#Checking if the feature is categorical or not
>length(unique(data1$Rainy.day))
[1] 3
> length(unique(data1$Direction_I))
[1] 8
> length(unique(data1$Direction_II))
[1] 17
#Plotting the categorical variable
>counts<- table(data1$Rainy.day)
>barplot(counts, xlab = 'Rain')
```



#Finding Correlation between feature columns

```
> cor(data1$Rain_Last.24hr,data1$BSS)
[1] -0.4707937
> cor(data1$Rain_Last.24hr,data1$Humidity_I)
[1] 0.3872418
> cor(data1$Rain_Last.24hr,data1$Humidity_II)
[1] 0.193109
```

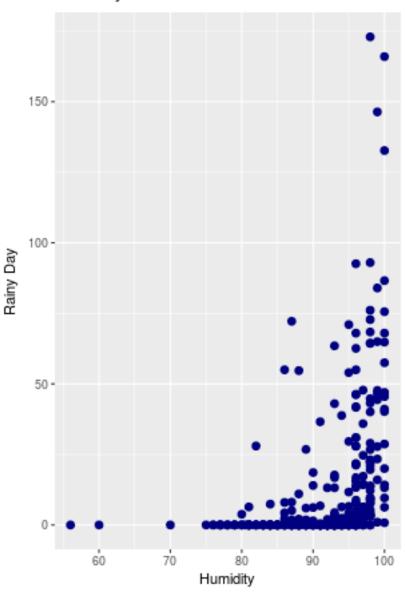
Some of the inferences drawn from variables in the data set:

- 1. There are total 90 days when the observations weren't taken for any factor, this could be due to failure of instruments/shutdown.
- 2. BSS Column has the value regarding best sunshine, the min value is 0 which is highly impossible(there goes no day without sunshine), hence the 0 value's are the missing data or the instrument is less accurate.
- 3. Rainy.day column is a categorical variable, where 0 means no rain and 1 means rain. Also it rained approximately 30% of the year.
- 4. The occurrence of rain and humidity are highly correlated.

#Visualising Data

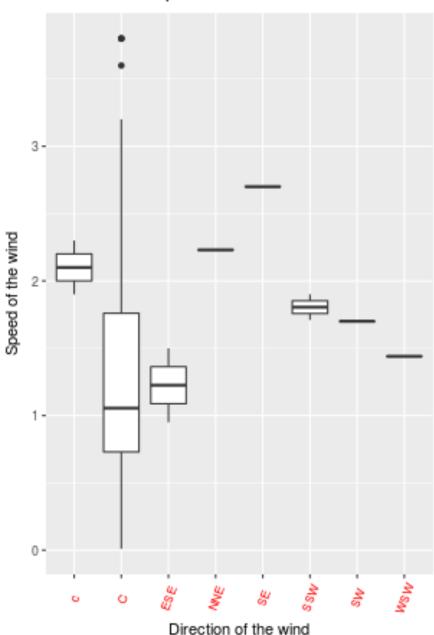
> ggplot(data1, aes(x= data1\$Humidity_I, y = data1\$Rain_Last.24hr)) + geom_point(size = 2.5, color="navy") + xlab("Humidity I") + ylab("Rainy Day") + ggtitle("Humidity vs Rain")

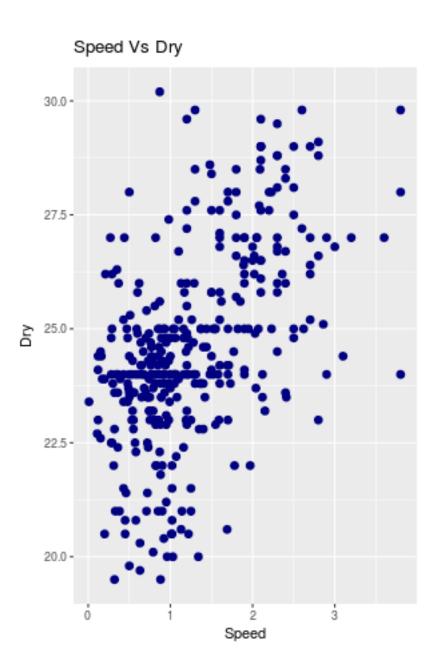
Humidity vs Rain

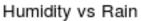


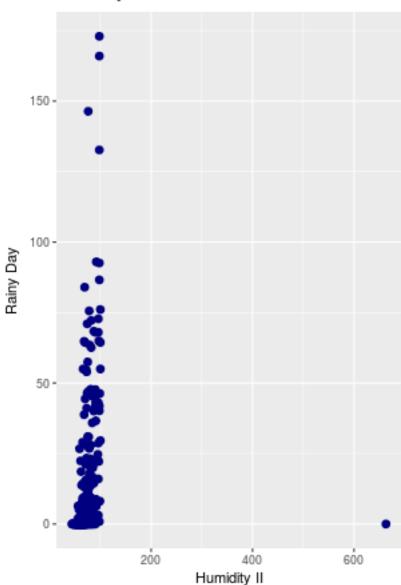
> ggplot(data1, aes(data1\$Direction_I, data1\$Speed_I))
+geom_boxplot() +ggtitle("Box Plot") + theme(axis.text.x =
element_text(angle = 70, vjust = 0.5, color = "red")) +
xlab("Direction of the wind") + ylab("Speed of the wind") +
ggtitle("Direction vs Speed of wind")

Direction vs Speed of wind



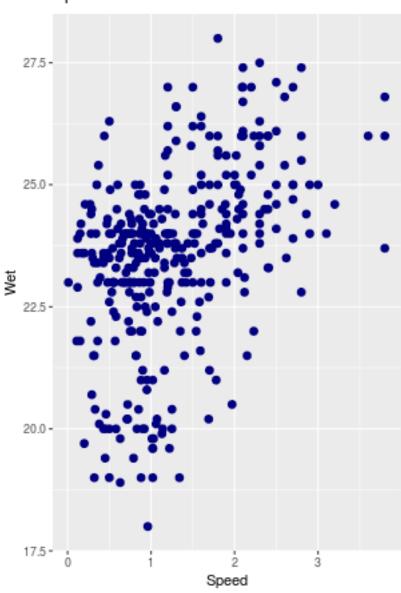






>ggplot(data1, aes(x= data1\$Speed_I, y = data1\$Wet_II)) + geom_point(size = 2.5, color="navy") + xlab("Speed") + ylab("Wet") + ggtitle("Speed Vs Wet")

Speed Vs Wet



Some of the inferences drawn from the visualisation:

- 1. From first 2 plots, it is inferred, If 75<Humidity<100 in last 24 hours then it rains.
- 2. From Box plot, it is inferred, Wind has variable speed in C direction, constant speed in NNE, SE, SW, WSW direction.
- 3. From last two plots, it is inferred that there is not much dependency between type of wind(dry or wet) and speed of the wind.

PREDICTIVE ANALYSIS IN R:

#Using Decision tree classifier to classify if It will rain or not on basis of all other weather factors.

```
library(rpart)
library(caret)
data1=read.csv(file.choose(),header = T)
data1=na.omit(data1)
#split data into test train
data<-data1
dt<-sort(sample(nrow(data),nrow(data)*0.8))
train<-data[dt,]
test<-data[-dt,]
trainx<-subset(train, select = -data$Rainy.day)</pre>
trainv<- train$Rainv.dav
testx<- subset(test, select = -data$Rainy.day)</pre>
testy<- test$Rainy.day
#fit Decision Tree
dtreeClass<-rpart(trainy~.,data=trainx,method="class")</pre>
#predict using fit model
pred<-predict(dtreeClass,testx,type = "class")</pre>
#show required parameters
xtab<-table(pred,testy)
#confusionMatrix(xtab)
confusionMatrix(xtab)
Confusion Matrix and Statistics
     testy
pred 0 1
    0 39 2
    1 0 31
```

Accuracy : 0.9722

95% CI: (0.9032, 0.9966)

No Information Rate : 0.5417 P-Value [Acc > NIR] : <2e-16

Kappa: 0.9438

Mcnemar's Test P-Value : 0.4795

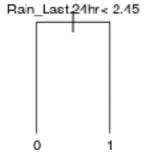
Sensitivity: 1.0000 Specificity: 0.9394 Pos Pred Value: 0.9512 Neg Pred Value: 1.0000 Prevalence: 0.5417

Detection Rate: 0.5417 Detection Prevalence: 0.5694 Balanced Accuracy: 0.9697

'Positive' Class: 0

>plot(dtreeClass,branch = 1,uniform = true(),margin = 1)

> text(dtreeClass,cex=.7)



> print(dtreeClass)

n= 288

node), split, n, loss, yval, (yprob)
 * denotes terminal node

- 1) root 288 103 0 (0.6423611 0.3576389)