

USE OF WEATHER DATA TO FORECAST NEXT DAY'S WEATHER

INTRODUCTION

This project work uses weather data for the Australian Capital Territory, Canberra to forecast the next day's weather. R was used for the data analysis.

Data input

The data must be read into R first to determine the observations and variables. The dimension of the data has 242 observations with 22 variables. The head displays the first six rows of the data (weather-2022)

```
#Q1
dat <- read.csv("C:/Users/pelza/OneDrive/Desktop/Machine Learning/weather-2022.csv", encoding= 'UTF-8', check.names=FALSE, header = F)
dim(dat)

## [1] 242  22

head(dat)

##      V1      V2      V3      V4
## 1      Date Minimum temperature (\xb0C) Maximum temperature (\xb0C)
## 2 JAN 2022-01-1      10.6      30.5
## 3 JAN 2022-01-2      13.3      32.6
## 4 JAN 2022-01-3      13.6      28.9
## 5 JAN 2022-01-4      12.6      27.6
## 6 JAN 2022-01-5      13.6      26.1
##      V5      V6      V7
## 1 Rainfall (mm) Evaporation (mm) Sunshine (hours)
## 2      0      <NA>      <NA>
## 3      0      <NA>      <NA>
## 4      5      <NA>      <NA>
## 5      0.2      <NA>      <NA>
## 6      0      <NA>      <NA>
##      V8      V9
## 1 Direction of maximum wind gust Speed of maximum wind gust (km/h)
## 2      NNW      31
## 3      SE      59
## 4      E      46
## 5      ESE      35
## 6      ESE      43
##      V10      V11
## 1 Time of maximum wind gust 9am Temperature (\xb0C) 9am relative humidity (%)
```

```

## 2          10:35          19.1
68
## 3          17:39          22.8
64
## 4          15:49          21.9
57
## 5          13:19          18.4
88
## 6          12:53          20.5
66
##          V13          V14          V15
## 1 9am cloud amount (oktas) 9am wind direction 9am wind speed (km/h)
## 2          <NA>          N          4
## 3          <NA>          N          9
## 4          <NA>          SE          7
## 5          8          SE          13
## 6          8          SE          11
##          V16          V17          V18
## 1 9am MSL pressure (hPa) 3pm Temperature (\xb0C) 3pm relative humidity (%)
## 2          1013.8          29.8          35
## 3          1009.9          31.4          25
## 4          1010.6          27.4          48
## 5          1011.9          26.9          48
## 6          1012.7          25.1          61
##          V19          V20          V21
## 1 3pm cloud amount (oktas) 3pm wind direction 3pm wind speed (km/h)
## 2          <NA>          NW          13
## 3          6          W          11
## 4          2          ENE          24
## 5          <NA>          E          17
## 6          8          ESE          26
##          V22
## 1 3pm MSL pressure (hPa)
## 2          1008.2
## 3          1006.3
## 4          1007.8
## 5          1009.5
## 6          1010.5

```

```
str(dat)
```

```

## 'data.frame':    242 obs. of  22 variables:
## $ V1 : chr  "" "JAN" "JAN" "JAN" ...
## $ V2 : chr  "Date" "2022-01-1" "2022-01-2" "2022-01-3" ...
## $ V3 : chr  "Minimum temperature (\xb0C)" "10.6" "13.3" "13.6" ...
## $ V4 : chr  "Maximum temperature (\xb0C)" "30.5" "32.6" "28.9" ...
## $ V5 : chr  "Rainfall (mm)" "0" "0" "5" ...
## $ V6 : chr  "Evaporation (mm)" NA NA NA ...
## $ V7 : chr  "Sunshine (hours)" NA NA NA ...
## $ V8 : chr  "Direction of maximum wind gust " "NNW" "SE" "E" ...

```

```
## $ V9 : chr "Speed of maximum wind gust (km/h)" "31" "59" "46" ...
## $ V10: chr "Time of maximum wind gust" "10:35" "17:39" "15:49" ...
## $ V11: chr "9am Temperature (\xb0C)" "19.1" "22.8" "21.9" ...
## $ V12: chr "9am relative humidity (%)" "68" "64" "57" ...
## $ V13: chr "9am cloud amount (oktas)" NA NA NA ...
## $ V14: chr "9am wind direction" "N" "N" "SE" ...
## $ V15: chr "9am wind speed (km/h)" "4" "9" "7" ...
## $ V16: chr "9am MSL pressure (hPa)" "1013.8" "1009.9" "1010.6" ...
## $ V17: chr "3pm Temperature (\xb0C)" "29.8" "31.4" "27.4" ...
## $ V18: chr "3pm relative humidity (%)" "35" "25" "48" ...
## $ V19: chr "3pm cloud amount (oktas)" NA "6" "2" ...
## $ V20: chr "3pm wind direction" "NW" "W" "ENE" ...
## $ V21: chr "3pm wind speed (km/h)" "13" "11" "24" ...
## $ V22: chr "3pm MSL pressure (hPa)" "1008.2" "1006.3" "1007.8" ...
```

Data Cleaning and Preparation

Q1. Removing Time of Maximum Wind Gust

```
df<-dat[, -c(10)]
```

#Comment

"df<-dat [, -c(10)]" is used to remove the 10th column Time of Maximum Wind Gust

#Check the Dimension, and str again

```
dim(df)
```

```
## [1] 242 21
```

```
str(df)
```

```
## 'data.frame': 242 obs. of 21 variables:
## $ V1 : chr "" "JAN" "JAN" "JAN" ...
## $ V2 : chr "Date" "2022-01-1" "2022-01-2" "2022-01-3" ...
## $ V3 : chr "Minimum temperature (\xb0C)" "10.6" "13.3" "13.6" ...
## $ V4 : chr "Maximum temperature (\xb0C)" "30.5" "32.6" "28.9" ...
## $ V5 : chr "Rainfall (mm)" "0" "0" "5" ...
## $ V6 : chr "Evaporation (mm)" NA NA NA ...
## $ V7 : chr "Sunshine (hours)" NA NA NA ...
## $ V8 : chr "Direction of maximum wind gust " "NNW" "SE" "E" ...
## $ V9 : chr "Speed of maximum wind gust (km/h)" "31" "59" "46" ...
## $ V11: chr "9am Temperature (\xb0C)" "19.1" "22.8" "21.9" ...
## $ V12: chr "9am relative humidity (%)" "68" "64" "57" ...
## $ V13: chr "9am cloud amount (oktas)" NA NA NA ...
## $ V14: chr "9am wind direction" "N" "N" "SE" ...
## $ V15: chr "9am wind speed (km/h)" "4" "9" "7" ...
## $ V16: chr "9am MSL pressure (hPa)" "1013.8" "1009.9" "1010.6" ...
## $ V17: chr "3pm Temperature (\xb0C)" "29.8" "31.4" "27.4" ...
## $ V18: chr "3pm relative humidity (%)" "35" "25" "48" ...
```

```
## $ V19: chr "3pm cloud amount (oktas)" NA "6" "2" ...
## $ V20: chr "3pm wind direction" "NW" "W" "ENE" ...
## $ V21: chr "3pm wind speed (km/h)" "13" "11" "24" ...
## $ V22: chr "3pm MSL pressure (hPa)" "1008.2" "1006.3" "1007.8" ...
```

Q2. Rename Dataset Column

#Comment: The data set has been renamed as shown below because they were too long.

```
names(df) <- c("Month", "Date", "MinTemp", "MaxTemp", "Rainfall",
"Evaporation", "Sunshine", "WindGustDir", "WindGustSpeed",
"Temp9am", "Humidity9am", "Cloud9am", "WindDir9am",
"WindSpeed9am", "Pressure9am", "Temp3pm", "Humidity3pm",
"Cloud3pm", "WindDir3pm", "WindSpeed3pm", "Pressure3pm")
```

```
dim(df);
```

```
## [1] 242 21
```

```
names(df)
```

```
## [1] "Month"      "Date"      "MinTemp"   "MaxTemp"
## [5] "Rainfall"   "Evaporation" "Sunshine"   "WindGustDir"
## [9] "WindGustSpeed" "Temp9am"   "Humidity9am" "Cloud9am"
## [13] "WindDir9am" "WindSpeed9am" "Pressure9am" "Temp3pm"
## [17] "Humidity3pm" "Cloud3pm"   "WindDir3pm" "WindSpeed3pm"
## [21] "Pressure3pm"
```

#Q3. Printing out Unique values

#Comment: We printed out the unique values using the codes below:

```
vnames <- colnames(df)
n <- nrow(df)
out <- NULL
for (j in 1:ncol(df)){
  vname <- colnames(df)[j]
  x <- as.vector(df[,j])
  n1 <- sum(is.na(x), na.rm=TRUE) # NA
  n2 <- sum(x=="NA", na.rm=TRUE) # "NA"
  n3 <- sum(x==" ", na.rm=TRUE) # missing
  nmiss <- n1 + n2 + n3
  nmiss <- sum(is.na(x))
  ncomplete <- n-nmiss
  out <- rbind(out, c(col.num=j, v.name=vname, mode=mode(x), n.level=length(unique(x)),
                                                                ncom=ncomplete, nmiss= nmiss, miss.prop=nmiss/n))
}
out <- as.data.frame(out)
```

```
row.names(out) <- NULL
out
```

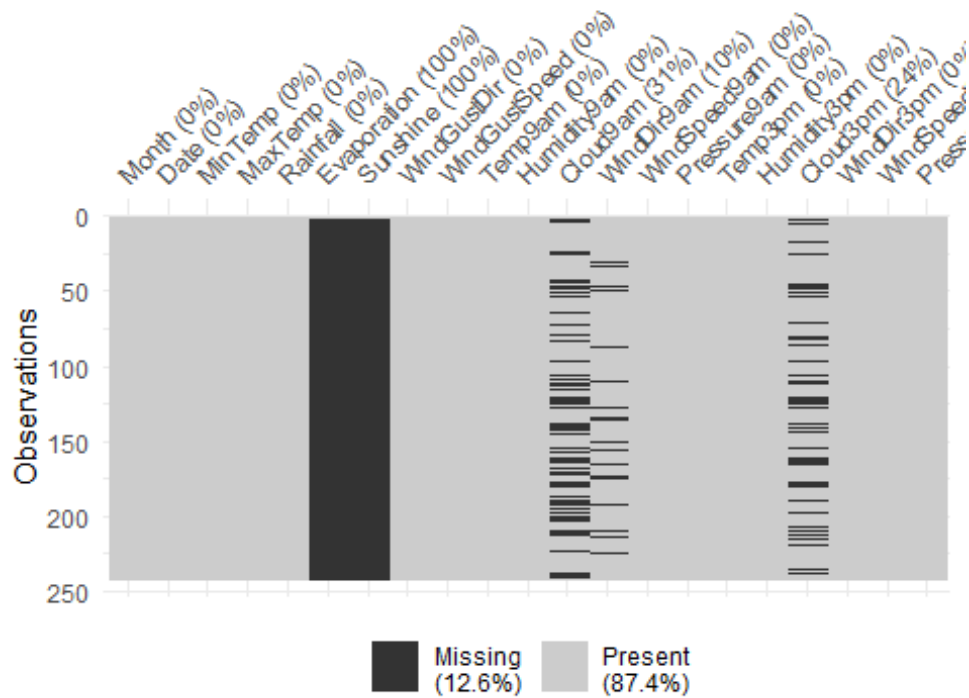
##	col.num	v.name	mode	n.level	ncom	nmiss	miss.prop
## 1	1	Month	character	9	242	0	0
## 2	2	Date	character	242	242	0	0
## 3	3	MinTemp	character	151	242	0	0
## 4	4	MaxTemp	character	153	241	1	0.00413223140495868
## 5	5	Rainfall	character	45	242	0	0
## 6	6	Evaporation	character	2	1	241	0.995867768595041
## 7	7	Sunshine	character	2	1	241	0.995867768595041
## 8	8	WindGustDir	character	18	241	1	0.00413223140495868
## 9	9	WindGustSpeed	character	31	241	1	0.00413223140495868
## 10	10	Temp9am	character	146	242	0	0
## 11	11	Humidity9am	character	44	242	0	0
## 12	12	Cloud9am	character	10	168	74	0.305785123966942
## 13	13	WindDir9am	character	17	217	25	0.103305785123967
## 14	14	WindSpeed9am	character	23	242	0	0
## 15	15	Pressure9am	character	156	242	0	0
## 16	16	Temp3pm	character	146	242	0	0
## 17	17	Humidity3pm	character	67	242	0	0
## 18	18	Cloud3pm	character	10	185	57	0.235537190082645
## 19	19	WindDir3pm	character	18	241	1	0.00413223140495868
## 20	20	WindSpeed3pm	character	23	242	0	0
## 21	21	Pressure3pm	character	159	242	0	0

```
colMeans(is.na(df))
```

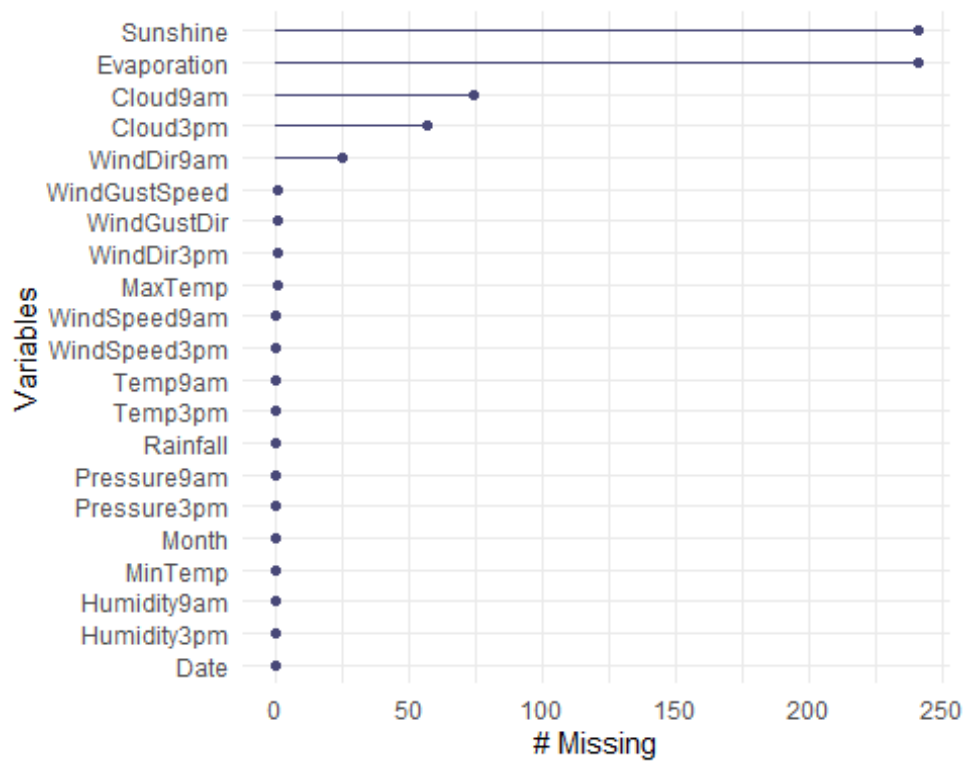
##	Month	Date	MinTemp	MaxTemp	Rainfall
##	0.000000000	0.000000000	0.000000000	0.004132231	0.000000000
##	Evaporation	Sunshine	WindGustDir	WindGustSpeed	Temp9am
##	0.995867769	0.995867769	0.004132231	0.004132231	0.000000000
##	Humidity9am	Cloud9am	WindDir9am	WindSpeed9am	Pressure9am
##	0.000000000	0.305785124	0.103305785	0.000000000	0.000000000
##	Temp3pm	Humidity3pm	Cloud3pm	WindDir3pm	WindSpeed3pm
##	0.000000000	0.000000000	0.235537190	0.004132231	0.000000000
##	Pressure3pm				
##	0.000000000				

Missing values by visualization

```
library(naniar)
vis_miss(df)
```



```
gg_miss_var(df)
```



Imputing the data

```
#install.packages("mice")
#library(mice, quietly=TRUE)
#fit.mice <- mice(df, m=1, maxit=50, method='pmm', seed=5474, printFlag=FALSE
)
#df <- complete(fit.mice, 1)

#df <- na.omit(df)
#dim(df)
```

#Q4. Showing Frequency Table of Each Value

```
#apply(df, 2, FUN = function(X){table(X, useNA="ifany")})
```

#Comment: A small section of the frequency table is shown below:

```
$Month
X
  APR AUG FEB JAN JUL JUN MAR MAY
1  30  29  28  31  31  30  31  31

$Date
X
2022-01-1 2022-01-10 2022-01-11 2022-01-12 2022-01-13 2022-01-14
      1      1      1      1      1      1
2022-01-15 2022-01-16 2022-01-17 2022-01-18 2022-01-19 2022-01-2
      1      1      1      1      1      1
2022-01-20 2022-01-21 2022-01-22 2022-01-23 2022-01-24 2022-01-25
      1      1      1      1      1      1
2022-01-26 2022-01-27 2022-01-28 2022-01-29 2022-01-3 2022-01-30
      1      1      1      1      1      1
2022-01-31 2022-01-4 2022-01-5 2022-01-6 2022-01-7 2022-01-8
      1      1      1      1      1      1
2022-01-9 2022-02-1 2022-02-10 2022-02-11 2022-02-12 2022-02-13
      1      1      1      1      1      1
2022-02-14 2022-02-15 2022-02-16 2022-02-17 2022-02-18 2022-02-19
      1      1      1      1      1      1
2022-02-2 2022-02-20 2022-02-21 2022-02-22 2022-02-23 2022-02-24
      1      1      1      1      1      1
2022-02-25 2022-02-26 2022-02-27 2022-02-28 2022-02-3 2022-02-4
      1      1      1      1      1      1
2022-02-5 2022-02-6 2022-02-7 2022-02-8 2022-02-9 2022-03-1
```

```
df <- df[-1, ]
```

#changing the value Calm to 0

#Comment: The character “Calm” was changed to integer “0” for data analysis purposes for the variable “dat\$WindSpeed9am”, which had that issue or problem.

```
df[df$WindSpeed9am == "Calm", ]$WindSpeed9am <- 0
df$WindSpeed9am
```

```
## [1] "4" "9" "7" "13" "11" "17" "9" "9" "13" "11" "13" "20" "15" "9"
"7"
## [16] "9" "7" "9" "15" "20" "11" "7" "11" "11" "6" "2" "13" "2" "0"
"11"
## [31] "4" "0" "13" "20" "20" "31" "20" "13" "17" "6" "6" "0" "13" "9"
"0"
## [46] "6" "4" "0" "7" "9" "7" "9" "13" "11" "9" "13" "6" "6" "11"
"13"
## [61] "15" "20" "11" "6" "17" "28" "19" "44" "17" "13" "7" "6" "9" "9"
"4"
## [76] "7" "9" "17" "9" "11" "7" "6" "2" "11" "9" "0" "6" "13" "11"
"17"
## [91] "35" "22" "22" "24" "2" "9" "9" "9" "15" "2" "7" "4" "6" "7"
"6"
## [106] "9" "2" "0" "0" "17" "7" "11" "9" "7" "7" "9" "9" "2" "20"
"26"
## [121] "4" "6" "2" "2" "13" "0" "26" "4" "7" "7" "9" "9" "0" "0"
"13"
## [136] "19" "19" "39" "7" "11" "7" "11" "9" "0" "9" "7" "2" "0" "0"
"9"
## [151] "35" "24" "7" "0" "26" "39" "35" "24" "20" "28" "24" "17" "19" "0"
"2"
## [166] "20" "17" "7" "6" "6" "0" "0" "0" "7" "30" "17" "11" "15" "9"
"7"
## [181] "2" "6" "20" "24" "33" "26" "13" "2" "20" "0" "9" "4" "4" "7"
"24"
## [196] "6" "2" "31" "24" "19" "7" "0" "13" "4" "11" "11" "24" "0" "19"
"0"
## [211] "6" "0" "26" "2" "28" "28" "15" "20" "9" "7" "9" "6" "0" "24"
"13"
## [226] "28" "31" "20" "2" "20" "26" "20" "6" "26" "30" "4" "17" "9" "4"
"2"
## [241] "2"
```

#Q.5 converting the column to numeric

#change WindSpeed data type from character to numeric

```
df$WindSpeed9am <- as.numeric(df$WindSpeed9am)
df$WindSpeed9am
```

```
## [1] 4 9 7 13 11 17 9 9 13 11 13 20 15 9 7 9 7 9 15 20 11 7 11
11 6
## [26] 2 13 2 0 11 4 0 13 20 20 31 20 13 17 6 6 0 13 9 0 6 4 0
7 9
## [51] 7 9 13 11 9 13 6 6 11 13 15 20 11 6 17 28 19 44 17 13 7 6 9
9 4
```



```
## [76] 7 9 17 9 11 7 6 2 11 9 0 6 13 11 17 35 22 22 24 2 9 9 9
15 2
## [101] 7 4 6 7 6 9 2 0 0 17 7 11 9 7 7 9 9 2 20 26 4 6 2
2 13
## [126] 0 26 4 7 7 9 9 0 0 13 19 19 39 7 11 7 11 9 0 9 7 2 0
0 9
## [151] 35 24 7 0 26 39 35 24 20 28 24 17 19 0 2 20 17 7 6 6 0 0 0
7 30
## [176] 17 11 15 9 7 2 6 20 24 33 26 13 2 20 0 9 4 4 7 24 6 2 31
24 19
## [201] 7 0 13 4 11 11 24 0 19 0 6 0 26 2 28 28 15 20 9 7 9 6 0
24 13
## [226] 28 31 20 2 20 26 20 6 26 30 4 17 9 4 2 2
```

#Q6

#Q6. Define a variable called “RainToday” and create an additional variable called “RainTomorrow”

#Comment: Here we created variable called RainToday with the ifelse statement condition that assigns 1 when Rainfall is ≥ 1 mm and 0 if Rainfall is < 1 . The additional variable termed “RainTomorrow” was created by shifting RainToday one day forward or upward.

```
RainToday: 1 if Rainfall > 1 mm, otherwise 0
df$RainToday <- ifelse(df$Rainfall > 1, 1, 0)
#RainTomorrow by shifting RainToday one day forward
df$RainTomorrow <- c(df$RainToday[2:nrow(df)], NA)
#Deleting NA data columns
df <- df[, !(names(df) %in% c("Evaporation", "Sunshine", "WindDir9am"))]

numeric_columns <- c(3, 4, 5, 7, 8, 9, 10, 11, 12, 13, 14, 19, 18, 16, 15)
df[, numeric_columns] <- lapply(df[, numeric_columns], function(x) as.numeric
(as.character(x)))

## Warning in FUN(X[[i]], ...): NAs introduced by coercion

str(df)

## 'data.frame':    241 obs. of  20 variables:
## $ Month          : chr  "JAN" "JAN" "JAN" "JAN" ...
## $ Date           : chr  "2022-01-1" "2022-01-2" "2022-01-3" "2022-01-4" ...
## $ MinTemp        : num  10.6 13.3 13.6 12.6 13.6 17.9 16.8 14.2 17.2 16.6 .
..
## $ MaxTemp        : num  30.5 32.6 28.9 27.6 26.1 27.9 23.5 28.1 27 31.8 ...
## $ Rainfall       : num   0 0 5 0.2 0 1.6 35.4 13.2 0 0 ...
## $ WindGustDir    : chr   "NNW" "SE" "E" "ESE" ...
## $ WindGustSpeed  : num   31 59 46 35 43 50 35 43 37 28 ...
## $ Temp9am        : num  19.1 22.8 21.9 18.4 20.5 21.4 19.7 17.7 19.8 22.8 .
```

```

..
## $ Humidity9am : num 68 64 57 88 66 78 95 99 74 76 ...
## $ Cloud9am : num NA NA NA 8 8 7 4 8 8 7 ...
## $ WindSpeed9am : num 4 9 7 13 11 17 9 9 13 11 ...
## $ Pressure9am : num 1014 1010 1011 1012 1013 ...
## $ Temp3pm : num 29.8 31.4 27.4 26.9 25.1 24.9 21.4 27.6 25.5 30.4 .
..
## $ Humidity3pm : num 35 25 48 48 61 65 76 45 60 33 ...
## $ Cloud3pm : num NA 6 2 NA 8 5 8 1 2 7 ...
## $ WindDir3pm : num NA NA NA NA NA NA NA NA NA NA ...
## $ WindSpeed3pm : chr "13" "11" "24" "17" ...
## $ Pressure3pm : num 1008 1006 1008 1010 1010 ...
## $ RainToday : num 0 0 1 0 0 1 1 1 0 0 ...
## $ RainTomorrow : num 0 1 0 0 1 1 1 0 0 0 ...

```

```
dim(df)
```

```
## [1] 241 20
```

#Our target variable is categorical, hence, we converted it to a factor.

```
df$RainToday <- as.factor(df$RainToday)
```

```
df$RainTomorrow <- as.factor(df$RainTomorrow)
```

#Q7. Save cleaned data set

```
write.csv(df, file="Cleaned_Weather-2022.csv", row.names =FALSE)
```

```
#Weather_csv<- read.csv("Weather.csv", header = TRUE, sep = ",")
```

#Exploratory Data Analysis

```
set.seed(1000)
```

```
tab <- table(df$Month, df$Cloud9am, useNA="no");
```

```
tab1 <- table(df$Month, df$Cloud3pm, useNA = "no")
```

```
tab2 <- table(df$Month, df$WindGustSpeed, useNA = "no")
```

```
tab3 <- table(df$Month, df$Humidity3pm, useNA = "no")
```

```
tab4 <- table(df$Month, df$Humidity9am, useNA = "no")
```

```
##
```

```
## Fisher's Exact Test for Count Data with simulated p-value (based on
## 2000 replicates)
```

```
##
```

```
## data: tab
```

```
## p-value = 0.7546
```

```
## alternative hypothesis: two.sided
```

```
##
```

```
## Fisher's Exact Test for Count Data with simulated p-value (based on
## 2000 replicates)
```

```
##
```

```
## data: tab1
```

```

## p-value = 0.92
## alternative hypothesis: two.sided

##
## Fisher's Exact Test for Count Data with simulated p-value (based on
## 2000 replicates)
##
## data:  tab2
## p-value = 0.003998
## alternative hypothesis: two.sided

##
## Fisher's Exact Test for Count Data with simulated p-value (based on
## 2000 replicates)
##
## data:  tab3
## p-value = 0.2824
## alternative hypothesis: two.sided

##
## Fisher's Exact Test for Count Data with simulated p-value (based on
## 2000 replicates)
##
## data:  tab4
## p-value = 0.06797
## alternative hypothesis: two.sided

## Warning in chisq.test(tab): Chi-squared approximation may be incorrect

##
## Pearson's Chi-squared test
##
## data:  tab
## X-squared = 40.247, df = 49, p-value = 0.8091

## Warning in chisq.test(tab1): Chi-squared approximation may be incorrect

##
## Pearson's Chi-squared test
##
## data:  tab1
## X-squared = 36.02, df = 49, p-value = 0.9162

## Warning in chisq.test(tab2): Chi-squared approximation may be incorrect

##
## Pearson's Chi-squared test
##
## data:  tab2
## X-squared = 245.06, df = 196, p-value = 0.009902

## Warning in chisq.test(tab3): Chi-squared approximation may be incorrect

```

```
##
## Pearson's Chi-squared test
##
## data:  tab3
## X-squared = 473.63, df = 455, p-value = 0.2639

## Warning in chisq.test(tab4): Chi-squared approximation may be incorrect
##
## Pearson's Chi-squared test
##
## data:  tab4
## X-squared = 300.94, df = 294, p-value = 0.3778
```

#Comments on the explanatory Data Analysis

#We see from the frequency tables above (tab ~ tab4), that there were some relationships between the selected variables and months of the year. To further justify the extent of relationship, we test for the hypothesis using an alpha level of 0.1 for fisher's test and chi-square test to determine the association between categorical variable and binary outcome.

#For the Fisher's test, we see that;

#tab, tab1 and tab3 all have p-values greater than the alpha level of 0.1. Hence, we accept the null hypothesis and reject the alternate hypothesis. We conclude that:

#For tab: there is no association between month and cloud

#For tab 1: there is no association between month and cloud

#For tab 3: there is no association between month and humidity3pm

#We also observed for tab 2 and tab 4 that their p-values were less than the alpha level of 0.1. Hence, we reject the null hypothesis and accept the alternate hypothesis. We can conclude for tab 2 and tab 4 that:

#For tab 2: there is an association between month and windgustspeed

#For tab 4: there is an association between month and humidity9am.

#Chi square test

#When it came to the chi square test, tab, tab1 tab3 and tab4 all have p-values greater than the alpha level of 0.1. Hence, we accept the null hypothesis and reject the alternate hypothesis. We conclude that: tab, tab1, tab3 and tab4 all do not have an association between month and the variables (cloud, cloud, humidity3pm and humidity9am respectively).

#For tab2: We see that the p-values were less than the alpha level of 0.1. Hence, we reject the null hypothesis and accept the alternate hypothesis. We can conclude tab2 has an association between month and windgustspeed.

cor(tab) *#correlation coefficient gives us matrix,*

```
##           1           2           3           4           5           6
## 1  1.0000000  0.0000000  0.25819889  0.25819889 -0.2236068 -0.48038446
## 2  0.0000000  1.0000000 -0.18257419 -0.18257419  0.7905694  0.56613852
## 3  0.2581989 -0.1825742  1.00000000  0.06666667 -0.4618802 -0.53748385
## 4  0.2581989 -0.1825742  0.06666667  1.00000000  0.0000000 -0.12403473
## 5 -0.2236068  0.7905694 -0.46188022  0.00000000  1.0000000  0.42966892
## 6 -0.4803845  0.5661385 -0.53748385 -0.12403473  0.4296689  1.00000000
## 7 -0.1581139 -0.3354102 -0.57154761  0.24494897  0.0000000  0.05063697
## 8 -0.7337994  0.0000000  0.18946619  0.24359938  0.2344036  0.18464591
##           7           8
## 1 -0.15811388 -0.73379939
## 2 -0.33541020  0.00000000
## 3 -0.57154761  0.18946619
## 4  0.24494897  0.24359938
## 5  0.00000000  0.23440362
## 6  0.05063697  0.18464591
## 7  1.00000000  0.09944903
## 8  0.09944903  1.00000000
```

#Comment

For a continuous predictor variable and Binary outcome: #Here, we used the Wilcoxon rank sum test (two-sample t test) to check the association between continuous variable and binary outcome

```
#df$RainToday <- as.numeric(df$RainToday)
#df$MaxTemp <- as.numeric(df$MaxTemp)
# Convert RainTomorrow to numeric if it's not already
#df$RainTomorrow <- as.numeric(df$RainTomorrow)

# Remove rows with missing values in MaxTemp or RainTomorrow
#df <- df[!is.na(df$MaxTemp) & !is.na(df$RainTomorrow), ]

str(df)

## 'data.frame':    241 obs. of  20 variables:
## $ Month          : chr  "JAN" "JAN" "JAN" "JAN" ...
## $ Date           : chr  "2022-01-1" "2022-01-2" "2022-01-3" "2022-01-4" ...
## $ MinTemp        : num  10.6 13.3 13.6 12.6 13.6 17.9 16.8 14.2 17.2 16.6 .
## ..
## $ MaxTemp        : num  30.5 32.6 28.9 27.6 26.1 27.9 23.5 28.1 27 31.8 ...
## $ Rainfall       : num  0 0 5 0.2 0 1.6 35.4 13.2 0 0 ...
## $ WindGustDir    : chr   "NNW" "SE" "E" "ESE" ...
## $ WindGustSpeed  : num   31 59 46 35 43 50 35 43 37 28 ...
## $ Temp9am        : num   19.1 22.8 21.9 18.4 20.5 21.4 19.7 17.7 19.8 22.8 .
## ..
## $ Humidity9am    : num   68 64 57 88 66 78 95 99 74 76 ...
## $ Cloud9am       : num   NA NA NA 8 8 7 4 8 8 7 ...
## $ WindSpeed9am   : num    4 9 7 13 11 17 9 9 13 11 ...
## $ Pressure9am    : num  1014 1010 1011 1012 1013 ...
## $ Temp3pm        : num   29.8 31.4 27.4 26.9 25.1 24.9 21.4 27.6 25.5 30.4 .
```

```

..
## $ Humidity3pm : num 35 25 48 48 61 65 76 45 60 33 ...
## $ Cloud3pm : num NA 6 2 NA 8 5 8 1 2 7 ...
## $ WindDir3pm : num NA NA NA NA NA NA NA NA NA NA ...
## $ WindSpeed3pm : chr "13" "11" "24" "17" ...
## $ Pressure3pm : num 1008 1006 1008 1010 1010 ...
## $ RainToday : Factor w/ 2 levels "0","1": 1 1 2 1 1 2 2 2 1 1 ...
## $ RainTomorrow : Factor w/ 2 levels "0","1": 1 2 1 1 2 2 2 1 1 1 ...

str(df)

## 'data.frame': 241 obs. of 20 variables:
## $ Month : chr "JAN" "JAN" "JAN" "JAN" ...
## $ Date : chr "2022-01-1" "2022-01-2" "2022-01-3" "2022-01-4" ...
## $ MinTemp : num 10.6 13.3 13.6 12.6 13.6 17.9 16.8 14.2 17.2 16.6 .
..
## $ MaxTemp : num 30.5 32.6 28.9 27.6 26.1 27.9 23.5 28.1 27 31.8 ...
## $ Rainfall : num 0 0 5 0.2 0 1.6 35.4 13.2 0 0 ...
## $ WindGustDir : chr "NNW" "SE" "E" "ESE" ...
## $ WindGustSpeed: num 31 59 46 35 43 50 35 43 37 28 ...
## $ Temp9am : num 19.1 22.8 21.9 18.4 20.5 21.4 19.7 17.7 19.8 22.8 .
..
## $ Humidity9am : num 68 64 57 88 66 78 95 99 74 76 ...
## $ Cloud9am : num NA NA NA 8 8 7 4 8 8 7 ...
## $ WindSpeed9am : num 4 9 7 13 11 17 9 9 13 11 ...
## $ Pressure9am : num 1014 1010 1011 1012 1013 ...
## $ Temp3pm : num 29.8 31.4 27.4 26.9 25.1 24.9 21.4 27.6 25.5 30.4 .
..
## $ Humidity3pm : num 35 25 48 48 61 65 76 45 60 33 ...
## $ Cloud3pm : num NA 6 2 NA 8 5 8 1 2 7 ...
## $ WindDir3pm : num NA NA NA NA NA NA NA NA NA NA ...
## $ WindSpeed3pm : chr "13" "11" "24" "17" ...
## $ Pressure3pm : num 1008 1006 1008 1010 1010 ...
## $ RainToday : Factor w/ 2 levels "0","1": 1 1 2 1 1 2 2 2 1 1 ...
## $ RainTomorrow : Factor w/ 2 levels "0","1": 1 2 1 1 2 2 2 1 1 1 ...

wilcox.test(df$MaxTemp ~ df$RainTomorrow, alternative = "two.sided")

##
## Wilcoxon rank sum test with continuity correction
##
## data: df$MaxTemp by df$RainTomorrow
## W = 4901, p-value = 0.6102

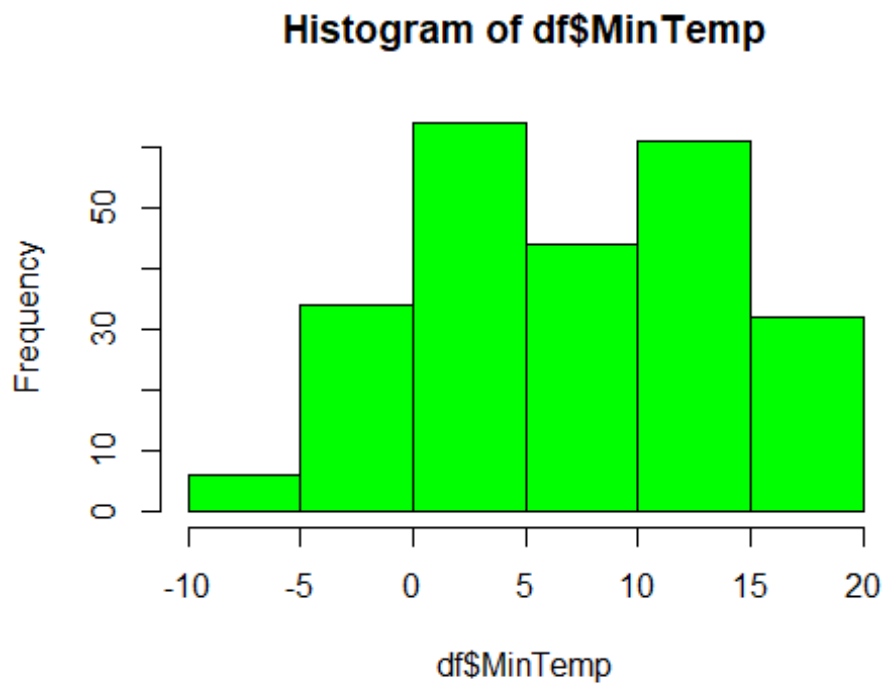
```

#Comment: The p -value above greater than the alpha level of 0.1, hence accept the null hypothesis and reject the alternate hypothesis. There is no association between MinTemp and RainTomorrow

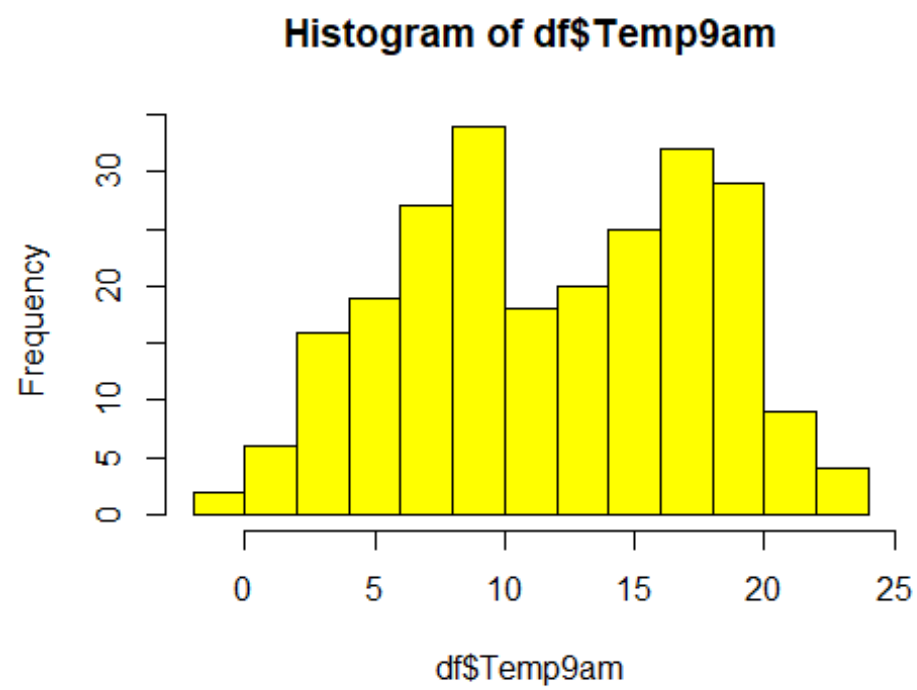
#Distribution Graphs Of Any Three Continuous Variables

#Below are the graphs for the histograms of the MinTemp, Temp9am and Raintoday.

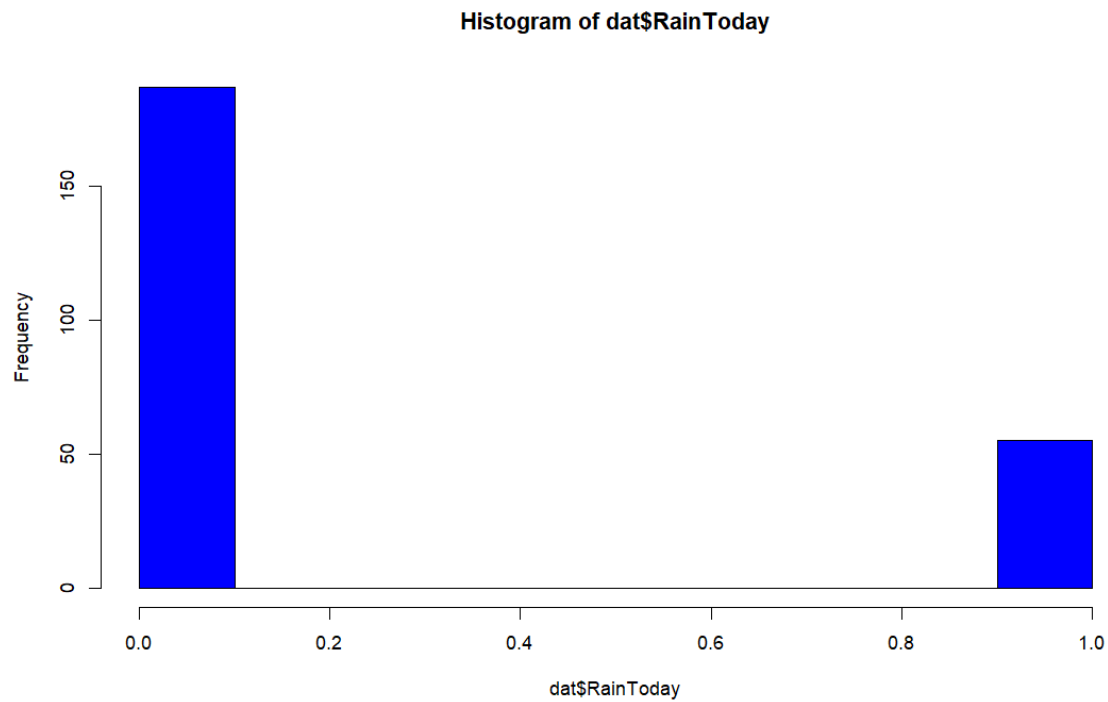
```
hist(df$MinTemp, col="green")
```



```
hist(df$Temp9am, col="yellow")
```



```
hist(df$Raintoday, col="blue")
```

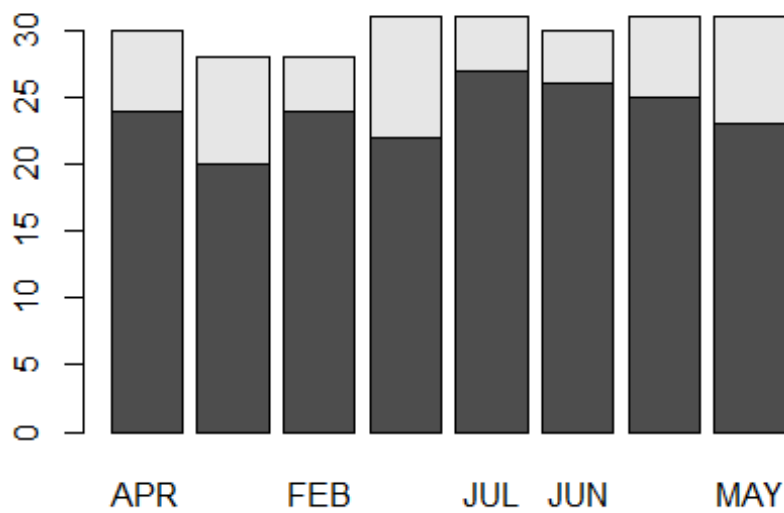
Distribution Graphs Of Any Three Categorical Variables

#Below are the bar graphs for Rainfall distribution for Windir9am, WindGustDir, and Winddir3pm.

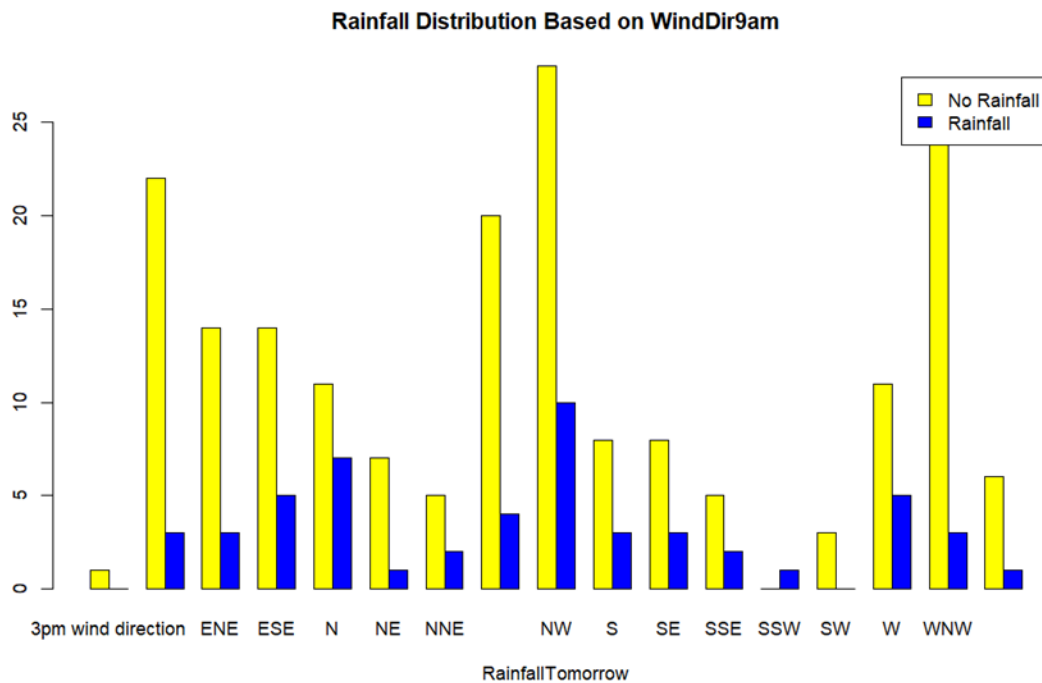
```
mytable<-table(df$RainTomorrow, df$Month) # to create table with proportions
mytable
```

```
##
##      APR  AUG  FEB  JAN  JUL  JUN  MAR  MAY
##    0   24   20   24   22   27   26   25   23
##    1    6    8    4    9    4    4    6    8
```

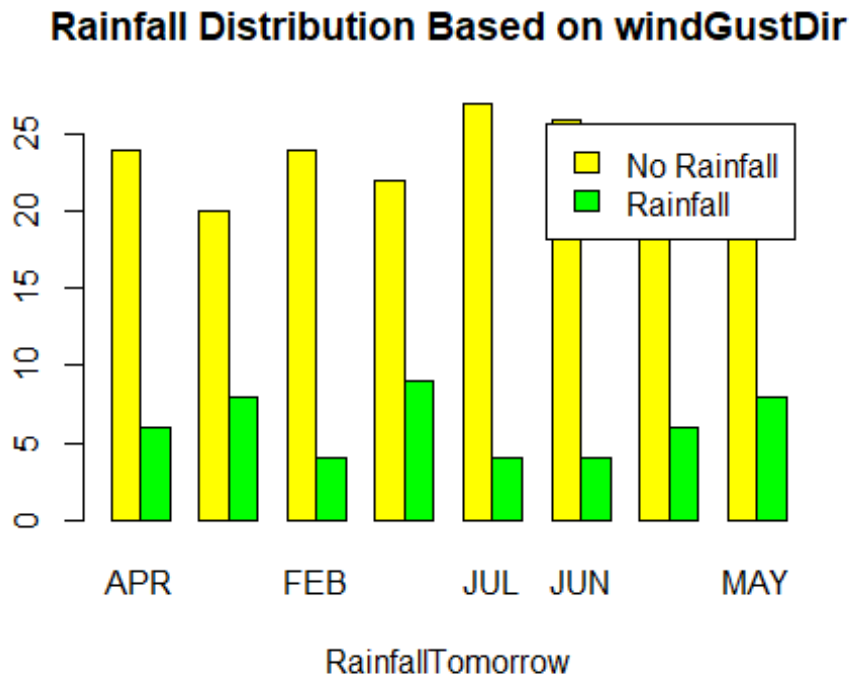
```
barplot((mytable))
```



```
#barplot(mytable*100)
barplot(mytable, main=" Rainfall Distribution Based on windDir9am",
        xlab="RainfallTomorrow", col=c("yellow","blue"),
        legend = rownames(mytable), beside = TRUE)
```



```
#barplot(mytable*100)
barplot(mytable, main="Rainfall Distribution Based on windGustDir",xlab="RainfallTomorrow",col=c("yellow","green"),legend = c('No Rainfall','Rainfall'), beside = TRUE)
```



```
#barplot(mytable*100)
barplot(mytable, main="Rainfall Distribution Based on WindDir3pm",xlab="RainfallTomorrow", col=c("blue","yellow"), legend = c('No Rainfall','Rainfall'), beside = TRUE)
```

Rainfall Distribution Based on WindDir3pm

