STA 135 HW 1

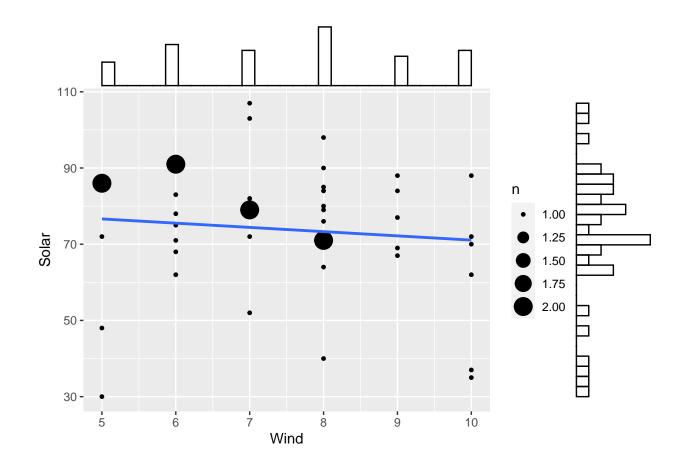
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3/30/2022

1.6)

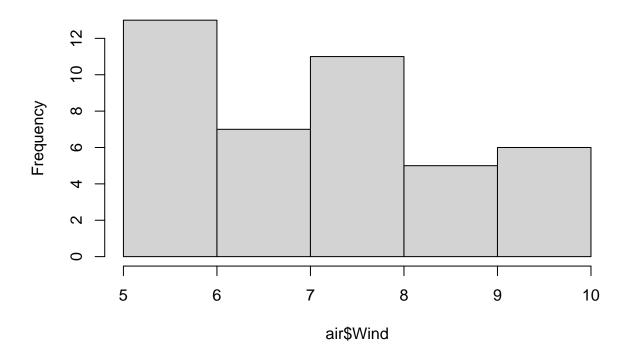
a)

```
#reading the data
air <- read_excel("air_pollution.xlsx", col_names = FALSE, col_types = "numeric")
## New names:
## * '' -> ...1
## * '' -> ...2
## * ' ' -> ...3
## * '' -> ...4
## * ' ' -> ...5
colnames(air) <- c("Wind", "Solar", "CO", "NO", "NO2", "O3", "HC")</pre>
head(air)
## # A tibble: 6 x 7
                                             HC
##
      Wind Solar
                    CO
                          NO
                               NO2
                                       03
     <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
## 1
         8
             98
                     7
                           2
                                12
## 2
         7
             107
                                 9
                                        5
                                              3
                     4
                           3
## 3
                                 5
        7
           103
                     4
                           3
                                       6
            88
                     5
                           2
                                 8
        10
                                       15
## 5
         6
              91
                     4
                           2
                                  8
                                       10
                                              3
              90
                                12
                     5
                                       12
# setting wind and solar values against each other as a graph
g <- ggplot(air, aes(Wind, Solar)) + geom_count() + geom_smooth(method="lm", se=F)
# plotting the marginal plot for wind and solar against each other
ggMarginal(g, type = "histogram", fill="transparent")
## 'geom_smooth()' using formula 'y ~ x'
## 'geom_smooth()' using formula 'y ~ x'
```



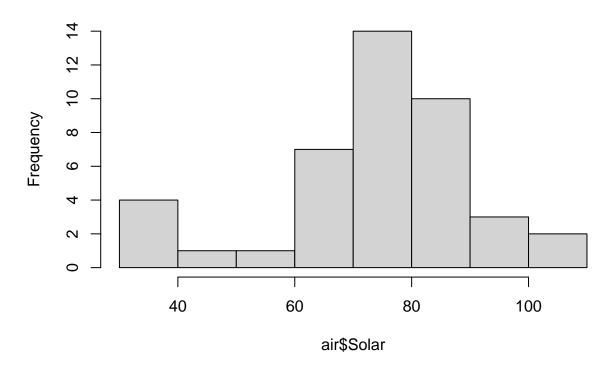
#histograms showing the dist of each given variable in the data set
 #You can speculate the graphs on your own :)
hist(air\$Wind)

Histogram of air\$Wind



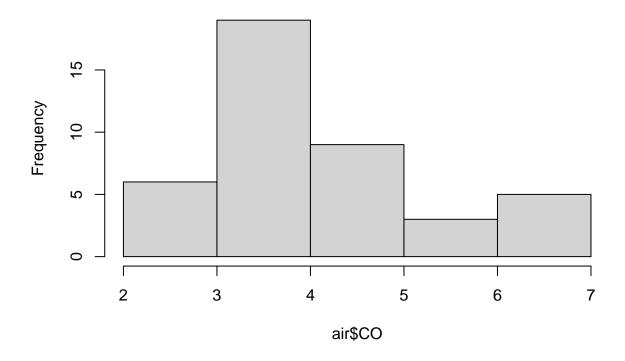
hist(air\$Solar)

Histogram of air\$Solar



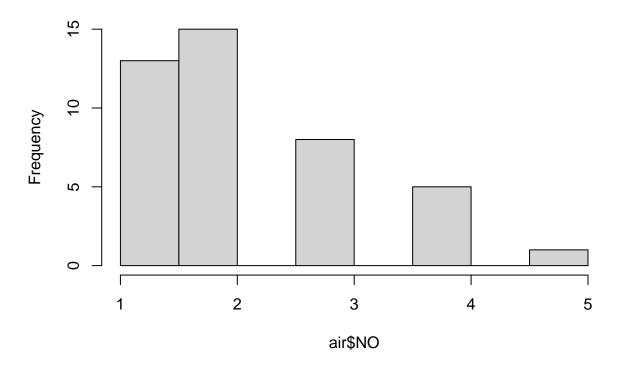
hist(air\$CO)

Histogram of air\$CO



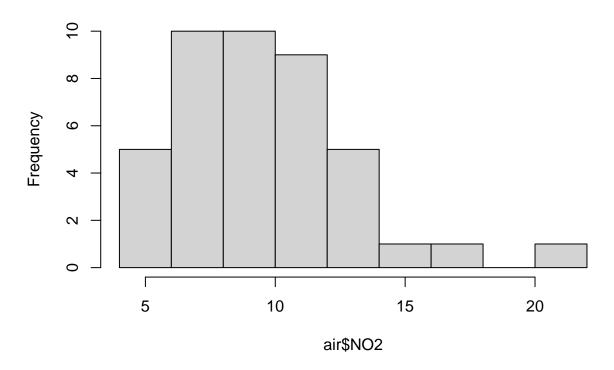
hist(air\$NO)

Histogram of air\$NO



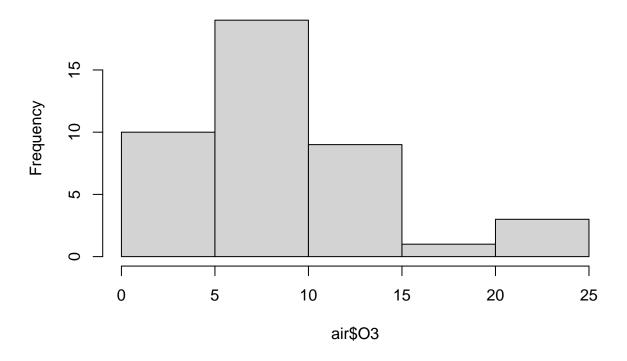
hist(air\$N02)

Histogram of air\$NO2



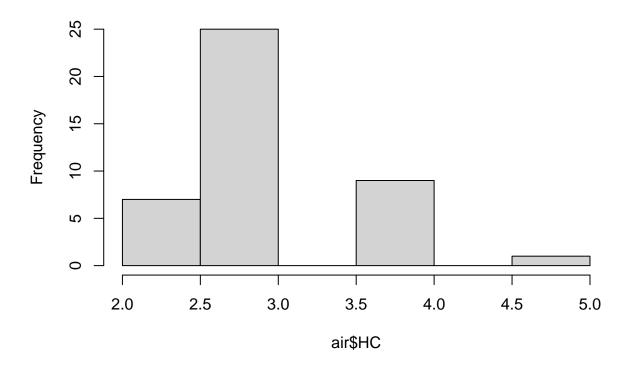
hist(air\$03)

Histogram of air\$O3

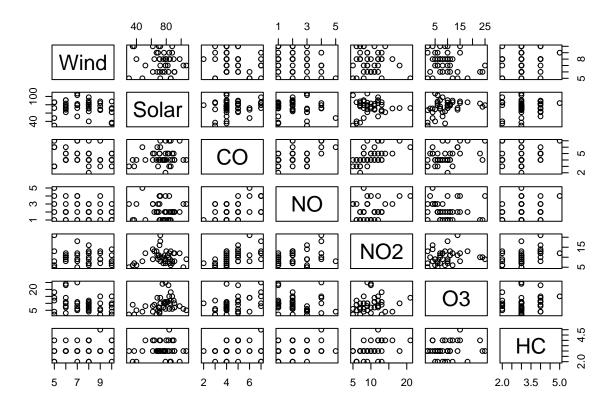


hist(air\$HC)

Histogram of air\$HC



pairs(air)



```
# number of observations
size = length(air$Wind)
```

b)

```
# means
air = as.matrix(air)
air_mean <- colMeans(air)
cat("Means:\n", air_mean)

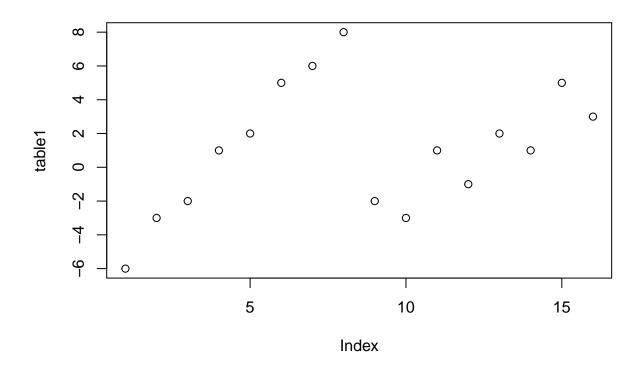
## Means:
## 7.5 73.85714 4.547619 2.190476 10.04762 9.404762 3.095238

# standard deviations for air data set
air_sd = apply(air, 2, sd)
cat( "\n\nStandard Deviations: \n", air_sd * (size/(size - 1)))

##
##
##
## $tandard Deviations:
## 1.619703 17.7582 1.263812 1.113878 3.453203 5.701587 0.7086185</pre>
```

```
# covariance matrix
cat("\n\nCovariance Matrix:\n", cov(air))
##
##
## Covariance Matrix:
## 2.5 -2.780488 -0.3780488 -0.4634146 -0.5853659 -2.231707 0.1707317 -2.780488 300.5157 3.909408 -1.3
# correlation matrix
cat("\n\nCorrelation Matrix:\n", cor(air))
##
##
## Correlation Matrix:
## 1 -0.1014419 -0.1938032 -0.2695426 -0.1098249 -0.2535928 0.1560979 -0.1014419 1 0.1827934 -0.073569
1.9)
a)
n = 8
# The given table
x1 = c(-6, -3, -2, 1, 2, 5, 6, 8)
x2 = c(-2, -3, 1, -1, 2, 1, 5, 3)
table1 = c(x1, x2)
# The scatterplot
```

plot(table1)



s 11: 5.529671 ## s 12: 4.326158 ## s 22: 3.039104

b)

```
#setting the rotating factor into code
theta = (26 * pi)/180

#now rotating both of the data vectors so that we get the question
xt1 = (x1 * cos(theta)) + (x2 * sin(theta))
xt2 = (-x1 * sin(theta)) + (x2 * cos(theta))

#printing final values into a data frame
data.frame(xt1, xt2)
```

```
xt1
## 1 -6.2695066 0.8326388
## 2 -4.0114956 -1.3812687
## 3 -1.3592169 1.7755363
## 4 0.4604229 -1.3371652
## 5 2.6743304 0.9208458
## 6 4.9323414 -1.2930617
## 7 7.5846200 1.8637434
## 8 8.5054658 -0.8105870
\mathbf{c}
# the standard deviations for the rotated data
sst11 = sd(xt1) * n/(n - 1)
sst22 = sd(xt2) * n/(n - 1)
cat("s-tilde 11:", sst11,
"\ns-tilde 22:", sst22)
## s-tilde 11: 6.096957
## s-tilde 22: 1.62497
d)
# new number of observations
m = 9
# new x1, x2 with added value
nx1 = c(x1, 4)
nx2 = c(x2, -2)
# recalculating the rotated data on the new observations
nxt1 = (nx1 * cos(theta)) + (nx2 * sin(theta))
nxt2 = (-nx1 * sin(theta)) + (nx2 * cos(theta))
# printing out the data with the new value and tilted
dat = data.frame(nx1, nx2, nxt1, nxt2)
colnames(dat) = c("X1", "X2", "Xt1", "Xt2")
dat
##
    X1 X2
                 Xt1
## 1 -6 -2 -6.2695066 0.8326388
## 2 -3 -3 -4.0114956 -1.3812687
## 3 -2 1 -1.3592169 1.7755363
## 4 1 -1 0.4604229 -1.3371652
## 5 2 2 2.6743304 0.9208458
## 6 5 1 4.9323414 -1.2930617
## 7 6 5 7.5846200 1.8637434
## 8 8 3 8.5054658 -0.8105870
## 9 4 -2 2.7184339 -3.5510727
```

```
# new ss values with tilt and extra val
nsst11 = sd(nxt1) * m/(m - 1)
nsst22 = sd(nxt2) * m/(m - 1)
# applying the formula in code
ntdist = sqrt(((nxt1^2)/nsst11) + ((nxt2^2)/nsst22))
# actual answer on the new value with the tilt
ntdist[9]
## [1] 2.748134
e)
# getting the ss values with the added values
ntable1 = c(nx1, nx2)
nss11 = sd(nx1) * m/(m - 1)
nss12 = sd(ntable1) * m/(m - 1)
nss22 = sd(nx2) * m/(m - 1)
# setup for getting the 1-19 formula into code...
ct2 = ((cos(theta))^2)
st2 = ((sin(theta))^2)
cs2 = (2 * sin(theta) * cos(theta))
# making the denominators for formula 1-19 understandable
d1 = (nss11 * ct2) + (nss12 * cs2) + (nss22 * st2)
d2 = (nss22 * ct2) + (nss12 * cs2) + (nss11 * st2)
```

[1] 1.334968

getting the dist formula all, al2, a22 values

 $ndist = sqrt((na11 * (nx1^2))+(na12 * nx1 * nx2)+(na22 * (nx2^2)))$

na12 = ((cs2 / 2)/(d1)) + ((cs2 / 2)/(d2))

na11 = (ct2/(d1)) + (st2/(d2))na22 = (st2/(d1)) + (ct2/(d2))

formula 1-19 into code

the actual answer.

1.18)

ndist[9]

```
# This is from the national track records for women data set, table 1.9
```