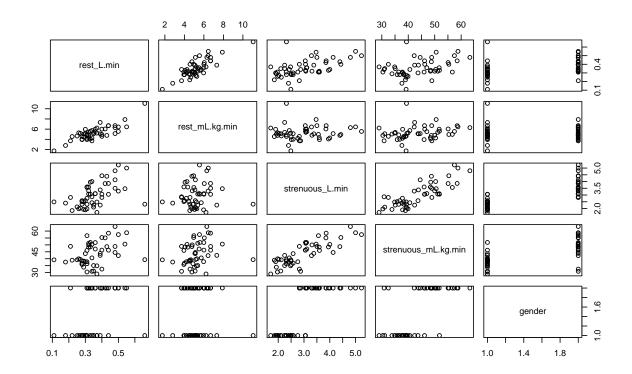
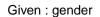
Homework 01 Joseph Blubaugh jblubau1@tamu.edu STAT 636-720

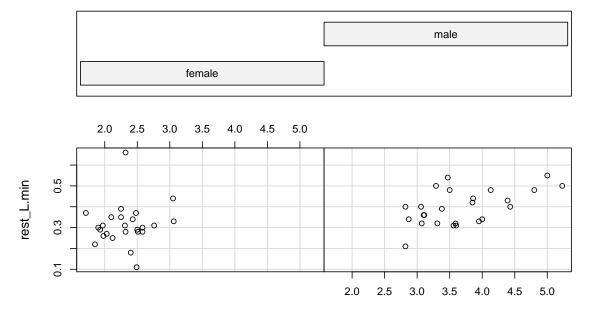
```
1)
```

```
oxygen =
  read.table("C:/Users/Joseph/Projects/learning/Statistics/STAT_636/Data/textbook/T6-12
             quote="\"",
             comment.char="")
colnames(oxygen) = c("rest_L.min", "rest_mL.kg.min", "strenuous_L.min",
                     "strenuous_mL.kg.min", "gender")
## a)
## Sample Averages
apply(X = oxygen[, 1:4], MARGIN = 2, FUN = mean)
         rest_L.min
                         rest_mL.kg.min
                                            strenuous_L.min
             0.3554
                                 5.2542
                                                      3.0014
strenuous_mL.kg.min
            43.7876
## Sample Standard Deviations
apply(X = oxygen[, 1:4], MARGIN = 2, FUN = sd)
         rest_L.min
                         rest_mL.kg.min
                                            strenuous_L.min
          0.1001878
                              1.3885863
                                                  0.8733796
strenuous_mL.kg.min
          8.4161326
## b)
pairs(oxygen)
```



c)
coplot(rest_L.min ~ strenuous_L.min | gender, data = oxygen)





strenuous_L.min

- a) The resting oxygen measurement means and standard deviation are lower than that measurements for the strenuous activity. Strenuous mL/kg/min is much higher and has more variation than the Resting mL/kg/min measurement.
- b) Resting L/min appears to be positively correlated with both Rest mL/kg/min and Strenuous L/min. Strenuous L/min also appears to be positively correlated with Strenuous mL/kg/min The observations of the female with a resting mL/kg/min of 11.05 appears to be an outlier and shows up as such on several plots.
- c) Visually there appears to be no obvious relationship between Rest L/min and Strenuous L/min for females, but the relationship appears to be positively correlated for men.

2)

a)

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \mu = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix} \Sigma = \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{12} & \sigma_{22} \end{bmatrix}$$

$$d(x, \mu) = \sqrt{(x - \mu)' \Sigma^{-1} (x - \mu)}$$

$$d(x, \mu) = \sqrt{\begin{bmatrix} x_1 - \mu_1 & x_2 - \mu_2 \end{bmatrix} \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{12} & \sigma_{22} \end{bmatrix}^{-1} \begin{bmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{bmatrix}}$$

$$= \sqrt{\sigma_{11} (x_1 - \mu_1)^2 + 2\sigma_{12} (x_1 - \mu_1) (x_2 - \mu_2) + \sigma_{22} (x_2 - \mu_2)^2}$$

$$a_{11} = \sigma_{11}, a_{12} = \sigma_{12}, a_{22} = \sigma_{22}$$

bi.

```
mu = c(1, -1)
Sigma = matrix(c(1, -1.6, -1.6, 4), nrow = 2)

mvn = function(x) {
   constant = 1/((2 * pi)^(2/2) * sqrt(det(Sigma)))
   scalar = exp(-t(x - mu) %*% solve(Sigma) %*% (x - mu)/2)

   out = constant * scalar

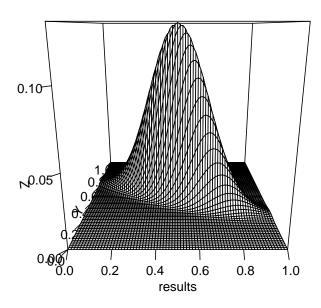
   return(as.numeric(out))
}
```

```
x1 = seq(-2, 4, length = 75)
x2 = seq(-13, 11, length = 75)

results = matrix(NA, nrow = 75, ncol = 75)

for (i in 1:75) {
   for (j in 1:75) {
     results[i, j] = mvn(c(x1[i], x2[j]))
   }
}

persp(results, ticktype = "detailed")
```



ii.

$$\mu = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \Sigma = \begin{bmatrix} 1.0 & -1.6 \\ -1.6 & 4 \end{bmatrix}$$
$$d(P,Q) = \sqrt{\sigma_{11}(x_1 - \mu_1)^2 + \sigma_{12}(x_1 - \mu_1)(x_2 - \mu_2) + \sigma_{22}(x_2 - \mu_2)^2}$$
$$O = (0,0), P = (0,-2)$$

Statistical Distance

$$d(O,Q) = \sqrt{1(0-1)^2 - 1.6(0-1)(0+1) + 4(0+1)^2}$$

$$= \sqrt{1+1.6+4}$$

$$= \sqrt{6.6} = 2.569$$

$$d(P,Q) = \sqrt{1(0-1)^2 - 1.6(0-1)(-2+1) + 4(-2+1)^2}$$

$$= \sqrt{1-1.6+4}$$

$$= \sqrt{3.4} = 1.843$$

Straightline Distance

$$d(O,Q) = \sqrt{(0-1)^2 + (0+1)^2}$$

$$= \sqrt{(2)}$$

$$= 1.41$$

$$d(P,Q) = \sqrt{0-1)^2 + (-2+1)^2}$$

$$= \sqrt{2}$$

$$= 1.41$$

Conclusion: Using the straightline distance both P and O are equally distant from Q, but when taking into accoun the variances, P is actually closer to Q than O is.

iii.

For R_P, x_2 appears to no longer by centered on P and as such will have less volume under the curve than R_O which is centered on O for both x_1 and x_2 , therefore $P(x \in R_O) > P(x \in R_P)$.