# DSA 8070 R Session 1: Characterizing and Displaying Multivariate Data

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# August 24, 2021

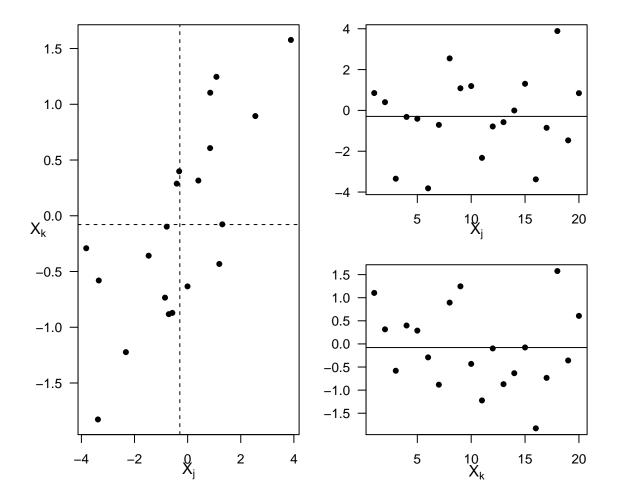
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## Descriptive Statistics

#### Sample covariance visualization

```
set.seed(123)
library(MASS)
dat \leftarrow mvrnorm(n = 20, mu = c(0, 0), Sigma = matrix(c(4, 1.4, 1.4, 1), 2))
n \leftarrow dim(dat)[1]
par(mar = c(3.6, 3.6, 0.8, 0.6), las = 1)
layout(matrix(c(1, 1, 2, 3), nrow = 2, ncol = 2))
plot(dat, pch = 16, las = 1, xlab = "", ylab = "")
mtext(expression(X[j]), 1, line = 2); mtext(expression(X[k]), 2, line = 2)
text(-4, 2, expression(paste(S[jk], " = ")))
text(-3.3, 2, round(cov(dat[, 1], dat[, 2]), 2))
abline(h = mean(dat[, 2]), lty = 2); abline(v = mean(dat[, 1]), lty = 2)
plot(1:n, dat[, 1], pch = 16, xlab = "", ylab = "")
abline(h = mean(dat[, 1]))
mtext(expression(X[j]), 1, line = 2)
plot(1:n, dat[, 2], pch = 16, xlab = "", ylab = "")
abline(h = mean(dat[, 2]))
mtext(expression(X[k]), 1, line = 2)
```

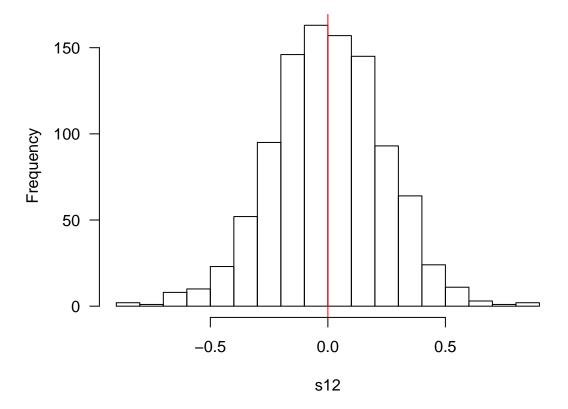


#### Sample and population covariance

Here we simulate data with size sample n=20 from a bivariate normal distribution with population covariance  $\rho_{12}=0$ . We calculate the sample covariance  $s_{12}$  for each simulated data set, and we repeat this process 1,000 times.

The main purpose of this exercise is to demonstrate that one can conduct *Monte Carlo* experiment to approximate the *sampling distribution* of  $s_{12}$ .

```
dat <- replicate(1000, mvrnorm(n = 20, mu = c(0, 0), Sigma = matrix(c(1, 0, 0, 1), 2)))
s12 <- apply(dat, 3, function(x) cov(x[, 1], x[, 2]))
hist(s12, 20, las = 1, main = "")
abline(v = 0, col = "red")</pre>
```



## Bivariate Data Example

```
data <- cbind(x1 = c(42, 52, 88, 58, 60), x2 = c(4, 5, 7, 4, 5))
(means <- apply(data, 2, mean))</pre>
## x1 x2
## 60 5
cov(data)
            x2
##
       x1
## x1 294 19.0
## x2 19 1.5
cor(data)
##
                       x2
             x1
## x1 1.0000000 0.9047619
## x2 0.9047619 1.0000000
```

#### Generliazed Variance

```
data(mtcars)
vars <- which(names(mtcars) %in% c("mpg", "disp", "hp", "drat", "wt"))
car <- mtcars[, vars]; S <- cov(car)
(genVar <- det(S))</pre>
```

## [1] 3951786

# Graphs and Visualization

#### pairs

```
head(iris)
     Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1
              5.1
                           3.5
                                        1.4
                                                     0.2 setosa
## 2
              4.9
                           3.0
                                        1.4
                                                     0.2 setosa
              4.7
                           3.2
                                        1.3
                                                     0.2 setosa
## 3
## 4
              4.6
                           3.1
                                        1.5
                                                     0.2 setosa
              5.0
                           3.6
## 5
                                        1.4
                                                     0.2 setosa
## 6
              5.4
                           3.9
                                        1.7
                                                     0.4 setosa
pairs(iris[, -5], las = 1, col = rep(c("green", "red", "blue"), each = 50), cex = 0.8)
                     2.0
                           3.0
                                 4.0
                                                           0.5
                                                                 1.5
                                                                       2.5
     Sepal.Length
4.0
3.5
                       Sepal.Width
3.0
2.5
2.0
                                         Petal.Length
                                                                           3
                                                                           2
2.5
2.0
1.5
                                                           Petal.Width
1.0
0.5
    4.5 5.5 6.5 7.5
                                        1 2 3 4 5 6 7
```

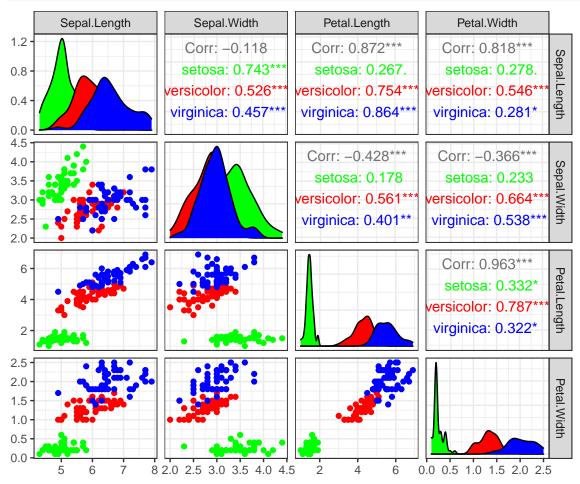
## ggpairs

```
library(GGally)

## Loading required package: ggplot2

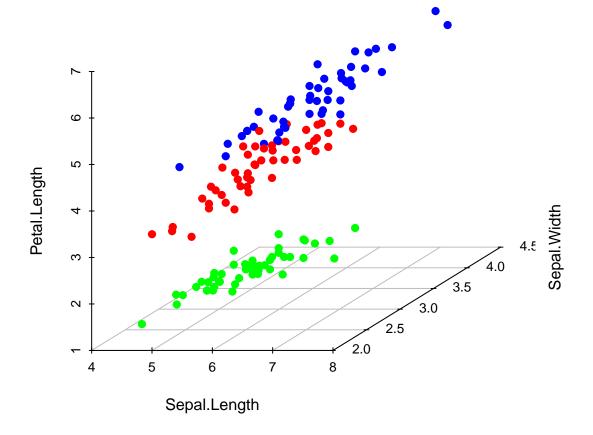
library(ggplot2)
p <- ggpairs(iris[, -5], aes(color = iris$Species)) + theme_bw()
# Change color manually.
# Loop through each plot changing relevant scales
for(i in 1:p$nrow) {</pre>
```

```
for(j in 1:p$ncol){
   p[i, j] <- p[i, j] +
        scale_fill_manual(values = c("green", "red", "blue")) +
        scale_color_manual(values = c("green", "red", "blue"))
}
</pre>
```

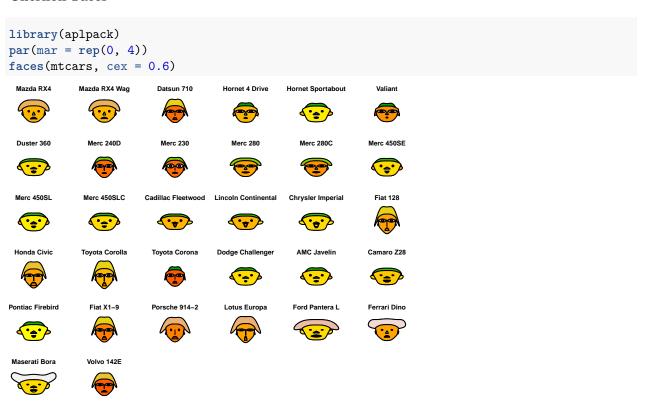


#### 3D Scatter Plot

```
library(scatterplot3d)
scatterplot3d(iris[, 1:3], pch = 19, color = rep(c("green", "red", "blue"), each = 50), grid = TRUE, box
```



# **Chernoff Faces**



```
## effect of variables:
## modified item Var
## "height of face " "mpg"
## "width of face " "cyl"
## "structure of face" "disp"
## "height of mouth " "hp"
## "width of mouth " "drat"
                   " "wt"
## "smiling
## "height of eyes " "qsec"
## "width of eyes " "vs"
## "height of hair " "am"
## "width of hair " "gear"
## "style of hair " "carb"
## "height of nose " "mpg"
## "width of nose " "cyl"
## "width of ear " "disp"
                   " "hp"
## "height of ear
```

## Visualizing Summary Statistics

