



MULTIVARIATE PROBABILITY DISTRIBUTIONS IN R

# Reading multivariate data

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## Course topics

- Read and analyze multivariate data
- Explore plotting techniques
- Use common statistical distributions
  - Gaussian and T distribution
- Techniques for high-dimensional data
  - Principal component analysis (PCA)



#### Structure of multivariate data

- Rectangular in shape organized by rows and columns
  - Rows represent observations
  - Columns represent variables
- May or may not include:
  - Row names or numbers
  - Column headers
- Possible missing data



#### Multivariate data examples

Iris Data from Cambridge University website

```
5.1 3.5 1.4 0.2 1
4.9 3.0 1.4 0.2 1
4.7 3.2 1.3 0.2 1
```

#### Birth Weight data (CSV with column header)

```
"", "case", "bwt", "gestation", "parity", "age", "height", "weight", "smoke"
"1", 1, 120, 284, 0, 27, 62, 100, 0
"2", 2, 113, 282, 0, 33, 64, 135, 0
```



## Reading data

#### From a URL

```
iris_url <- "http://mlg.eng.cam.ac.uk/teaching/3f3/1011/iris.data"
iris_raw <- read.table(iris_url, sep ="", header = FALSE)</pre>
```

#### Locally

```
iris_raw <- read.table("iris.txt", sep = "", header = FALSE)</pre>
```



## Viewing the dataset

```
head(iris_raw, n = 4)

V1 V2 V3 V4 V5

1 5.1 3.5 1.4 0.2 1

2 4.9 3.0 1.4 0.2 1

3 4.7 3.2 1.3 0.2 1

4 4.6 3.1 1.5 0.2 1
```



## Assigning column names

```
Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1 5.1 3.5 1.4 0.2 1
2 4.9 3.0 1.4 0.2 1
3 4.7 3.2 1.3 0.2 1
4 4.6 3.1 1.5 0.2 1
5 5.0 3.6 1.4 0.2 1
6 5.4 3.9 1.7 0.4 1
```



## Accessing specific columns

Check current names of columns

```
names(iris_raw)
"Sepal.Length" "Sepal.Width" "Petal.Length" "Petal.Width" "Species"
```

Accessing Sepal length and Sepal width columns

```
iris_raw[, 1:2]
iris[, c('Sepal.Length', 'Sepal.Width')]
```



## Changing data types

Change the last variable Species to a factor

```
iris_raw$species <- as.factor(iris_raw$species)

str(iris_raw)

'data.frame': 150 obs. of 5 variables:
   $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
   $ Sepal.Width: num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
   $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
   $ Petal.Width: num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
   $ Species : Factor w/ 3 levels "1", "2", "3": 1 1 1 1 1 1 1 1 1 1 ...</pre>
```



## Assigning factor labels

Recode the species labels from 1, 2 and 3 to setosa, versicolor and virginica

- Assign factor labels
- Change first variable to a factor

```
str(iris_raw)

'data.frame': 150 obs. of 5 variables:
   $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
   $ Sepal.Width: num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
   $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
   $ Petal.Width: num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
   $ Species : Factor w/ 3 levels "setosa", "versicolor", ..: 1 1 1 1 1 1 1 1
```



#### Reading csv data with named columns

#### Birth Weight data (CSV with column header)

```
"","case","bwt","gestation","parity","age","height","weight","smoke"
"1",1,120,284,0,27,62,100,0
"2",2,113,282,0,33,64,135,0
"3",3,128,279,0,28,64,115,1
"4",4,123,NA,0,36,69,190,0
```

#### Reading Birth Weight data





MULTIVARIATE PROBABILITY DISTRIBUTIONS IN R

# Let's read some multivariate data!



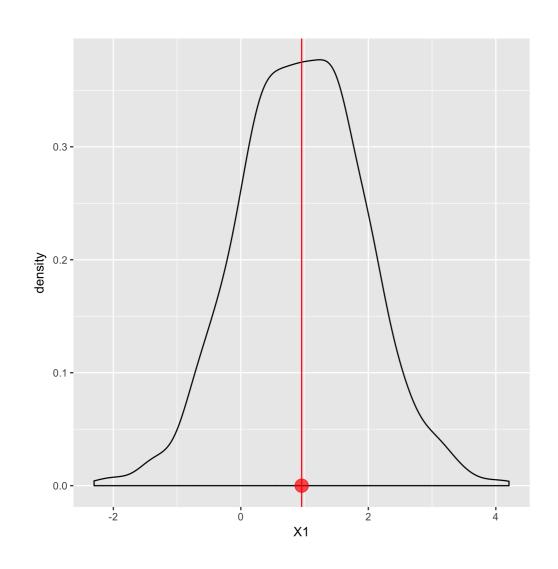


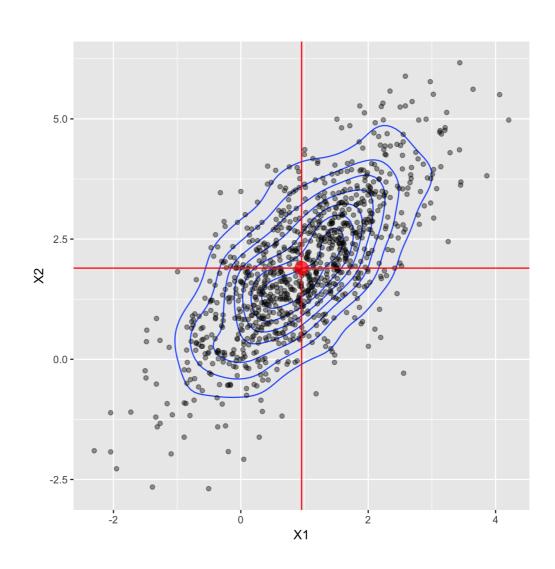
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# Mean vector and variancecovariance matrix

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## Mean represents the location of the distribution



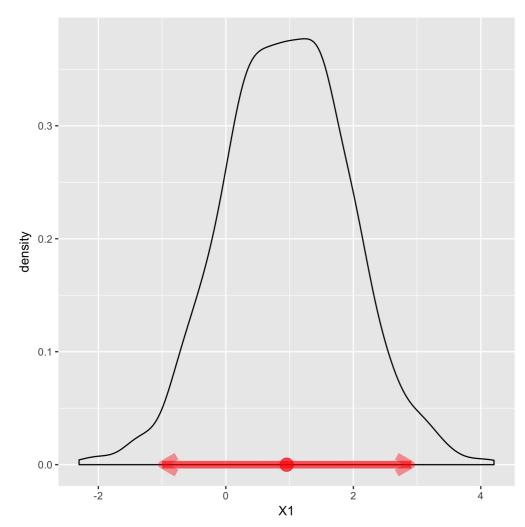


Mean 0.95

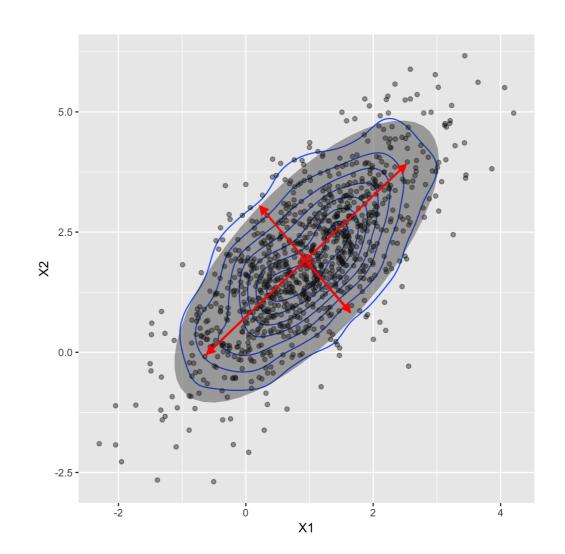
Mean vector (0.95 1.89)



#### Variance-covariance matrix represents the spread



Variance 1.02



Variance-covariance 
$$\begin{pmatrix} 1.02 & 0.97 \\ 0.97 & 2 \end{pmatrix}$$



## Calculating the mean

```
colMeans(iris_raw[, 1:4])

Sepal.Length Sepal.Width Petal.Length Petal.Width

5.84 3.05 3.76 1.20
```

#### Functions that calculate means by subgroups

- by()
- aggregate()



## Calculating the group mean using by



#### Calculating the group mean using aggregate

```
aggregate(. ~ Species, iris_raw, mean)

Species Sepal.Length Sepal.Width Petal.Length Petal.Width

1 setosa 5.01 3.42 1.46 0.244

2 versicolor 5.94 2.77 4.26 1.326

3 virginica 6.59 2.97 5.55 2.026
```



## Calculating the variance-covariance and correlation matrices

#### Variance

```
var(iris raw[, 1:4])
          Sepal.Length Sepal.Width Petal.Length Petal.Width
              0.6857
                      -0.0393
                                 1.274
                                           0.517
Sepal.Length
Sepal.Width -0.0393 0.1880
                                 -0.322
                                           -0.118
Petal.Length 1.2737 -0.3217
                                 3.113 1.296
                                  1.296
Petal.Width 0.5169
                       -0.1180
                                           0.582
```

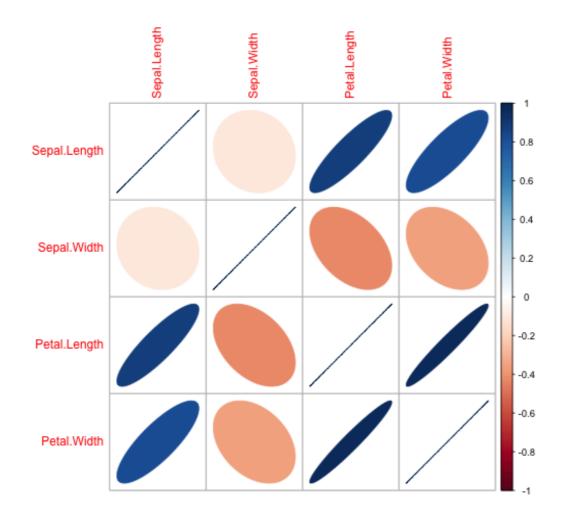
#### Correlation

```
cor(iris raw[, 1:4])
         Sepal.Length Sepal.Width Petal.Length Petal.Width
                            0.872
Sepal.Length
              1.000
                      -0.109
                                          0.818
                                -0.421
Sepal.Width
             -0.109 1.000
                                         -0.357
Petal.Length 0.872 -0.421 1.000 0.963
Petal.Width
         0.818
                   -0.357 0.963
                                         1.000
```

#### Visualization of correlation matrix

corrplot function to visualize correlation plot

```
corrplot(cor(iris_raw[, 1:4]), method = "ellipse")
```

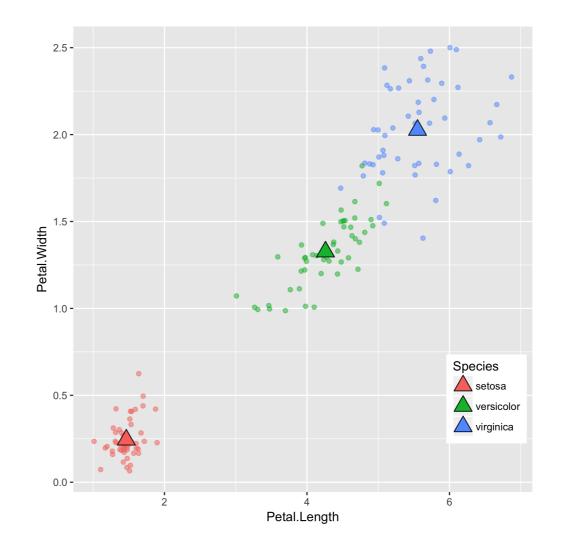




# Interpretation of means

#### Means

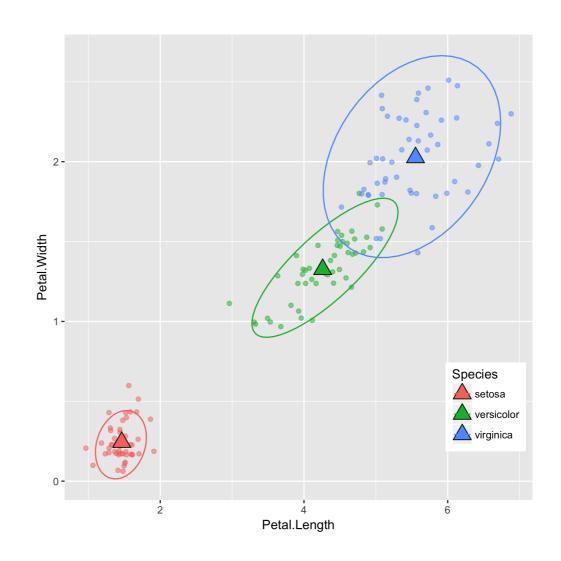
Species	Petal.Length	Petal.Width
setosa	1.46	0.244
versicolor	4.26	1.326
virginica	5.55	2.026





# Interpretation of variances

setosa	Petal.Length	Petal.Width
Petal.Length	0.030	0.006
Petal.Width	0.006	0.011
versicolor	Petal.Length	Petal.Width
Petal.Length	0.221	0.073
Petal.Width	0.073	0.039
virginica	Petal.Length	Petal.Width
Petal.Length	0.305	0.049
Petal.Width	0.049	0.075







#### MULTIVARIATE PROBABILITY DISTRIBUTIONS IN R

# Let's practice!





MULTIVARIATE PROBABILITY DISTRIBUTIONS IN R

# Plotting multivariate data

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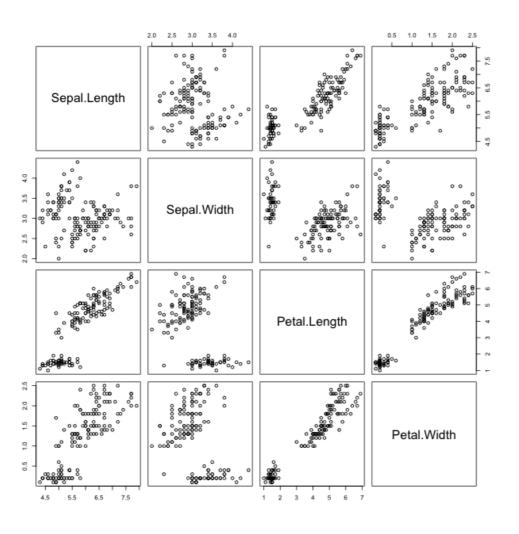
## Various plotting options

- Basic R plot
- lattice library
- ggplot
- 3D plotting options



## Basic R plot for multivariate data

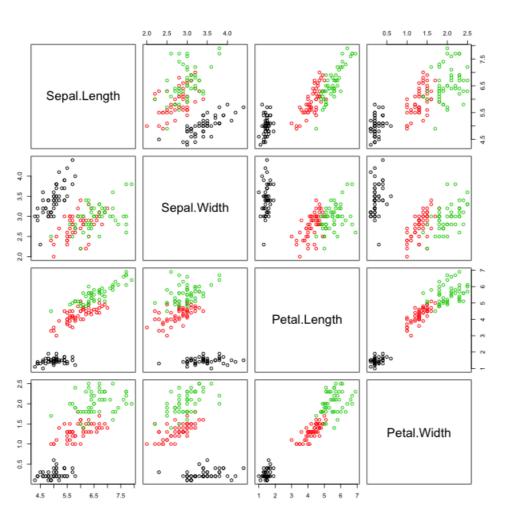
```
pairs(iris_raw[, 1:4])
```



Plot not as useful with many variables

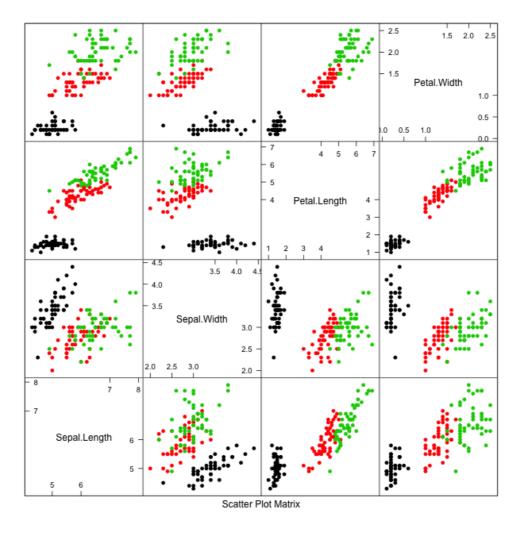
# Pairs plot by color

```
pairs(iris_raw[, 1:4], col = iris_raw$Species)
```



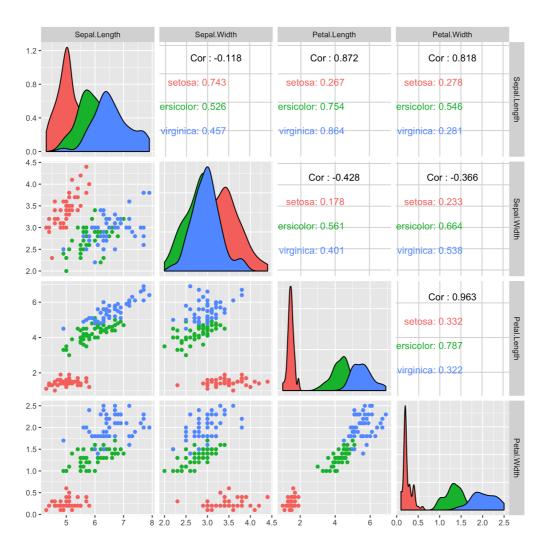
## Lattice

```
library(lattice)
splom(~iris_raw[, 1:4], col = iris_raw$Species, pch = 16)
```



## Using ggplot

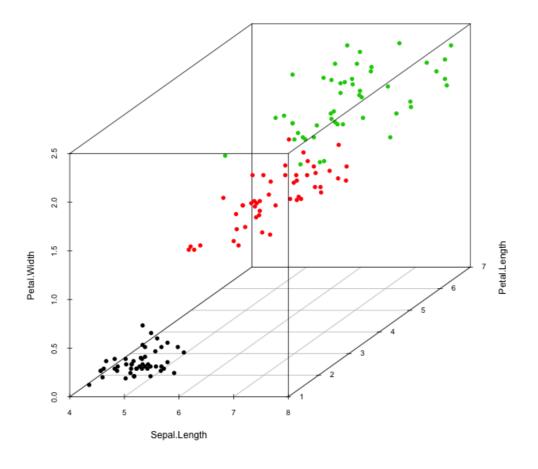
```
library(ggplot2)
library(GGally)
ggpairs(data = iris_raw, columns = 1:4, mapping = aes(color = Species))
```



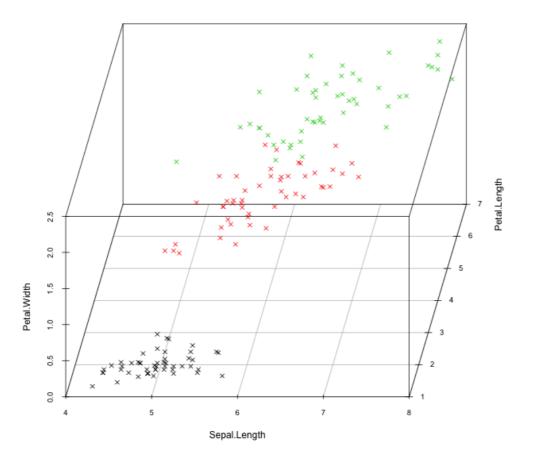


# 3D plots

```
library(scatterplot3d)
scatterplot3d(iris_raw[, c(1, 3, 4)], color = as.numeric(iris_raw$Species))
```



# 3D plots







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# Let's practice some plotting with the wine data!