

# Intro 2 R & Stats

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Intro 2 R

Intro 2 Stats.

## Intro 2 R

# Content

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  - 2.R classes
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  - 2.**Rcpp**
  - 3.**R shiny**
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# R Basic

- ▶ Install R  
<http://www.r-project.org/>
- ▶ Rstudio is a powerful IDE for R.  
<http://www.rstudio.com/>
- ▶ Working directory

```
getwd()  
setwd("D:/")
```

# R classes

character, numeric, integer, logical, **vector**, **matrix**, **factor**, **data frame**, **list**.

## ► vector

```
x <- c(10,3,2,1,NA); x[4]; mode(x); length(x)
```

## ► matrix

```
X <- matrix(1:12,nrow=3,byrow=TRUE,dimnames=NULL);X[1,]  
## [1] 1 2 3 4  
rbind(X,c(3,2,1,3))  
##      [,1] [,2] [,3] [,4]  
## [1,]    1    2    3    4  
## [2,]    5    6    7    8  
## [3,]    9   10   11   12  
## [4,]    3    2    1    3  
apply(X,2,mean)  
## [1] 5 6 7 8
```

## R classes

- ▶ factor: a compact way to handle categorical data.

```
sex <- c("M", "F", "M", "F", "F")
sex.factor <- as.factor(sex); table(sex.factor)
## sex.factor
## F M
## 3 2
height <- c(174, 165, 180, 171, 160)
tapply(height, sex.factor, mean)
##      F      M
## 165.3 177.0
gl(2, 3)
## [1] 1 1 1 2 2 2
## Levels: 1 2
```

## R classes

- ▶ list: a useful way to combine a collection of different objects.

```
family <- list(name="Fred",wife="Jane",
               children=c("XY","XX"))
family$name;family[[3]][2];family[["wife"]]
## [1] "Fred"
## [1] "XX"
## [1] "Jane"
family[1:2]
## $name
## [1] "Fred"
##
## $wife
## [1] "Jane"
unlist(family)
##      name      wife children1 children2
## "Fred"  "Jane"      "XY"      "XX"
```



## R classes

- data frame: a specific list of vectors and/or factors of the same length.

```
df<-data.frame(  
  Name=c("Alice", "Becka", "James", "Jeffrey", "John"),  
  Sex=c("F", "F", "M", "M", "M"),  
  Age=c(13, 13, 12, 13, 12),  
  Height=c(56.5, 65.3, 57.3, 62.5, 59.0),  
  Weight=c(84.0, 98.0, 83.0, 84.0, 99.5)  
); df
```

##		Name	Sex	Age	Height	Weight
## 1		Alice	F	13	56.5	84.0
## 2		Becka	F	13	65.3	98.0
## 3		James	M	12	57.3	83.0
## 4		Jeffrey	M	13	62.5	84.0
## 5		John	M	12	59.0	99.5

## R statements

```
if(cond){expr}  
else if(cond){expr}  
else{expr}  
  
for(var in seq){expr}  
  
while(cond){expr}  
  
break;next;  
repeat {expr}
```

```
switch(3,1,2,3,4)  
## [1] 3  
switch("mean",mean=mean(c(1,3,2)),median=3)  
## [1] 2
```

# R functions and libraries

```
install.packages(packname)
library(packname)

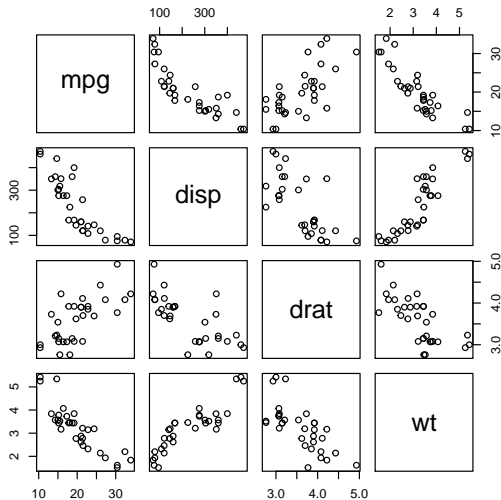
funname <- function(arg1,arg2,...){
  statements
  return(objects)
}
```

```
centre <- function(x, type){
  switch(type,
    mean = mean(x),
    median = median(x))
}
x <- rcauchy(10)
centre(x, "mean")
## [1] -0.09672
```

# Basic graph

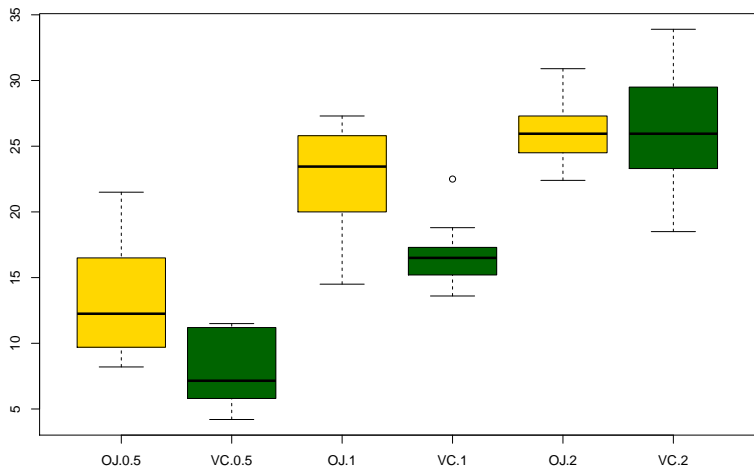
```
# Basic Scatterplot Matrix
```

```
pairs(~mpg+dis+drat+wt,data=mtcars)
```



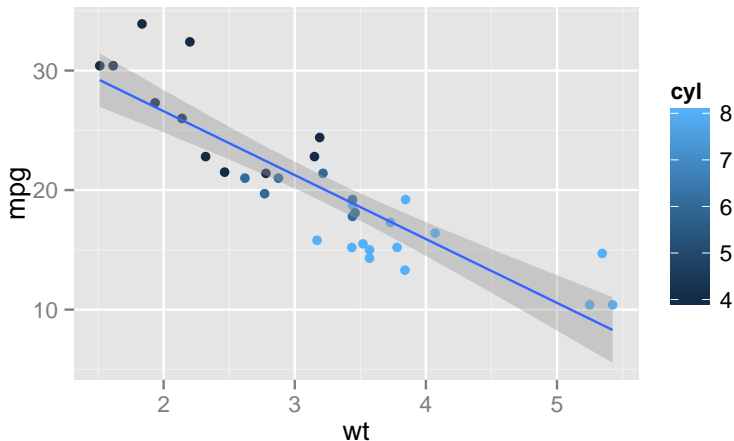
## Basic graph

```
boxplot(len~supp*dose, data=ToothGrowth, notch=FALSE,  
col=(c("gold","darkgreen")))
```



# ggplot

```
library(ggplot2)
qplot(wt, mpg, data=mtcars, geom=c("point", "smooth"),
      method="lm", formula=y~x, color=cyl)
```



## Intro 2 Stats.

# Content

- ▶ Bayesian Stats
  - 1.the Bayes Rule
  - 2.prior
  - 3.Hierachical Bayes
- ▶ Frequentist Stats
  - 1.Large sample theory for the MLE
  - 2.p-value, FDR



# Bayesian Stats

- ▶ Bayes Rule

$$Pr(\theta|Data) \propto Pr(\theta)Pr(Data|\theta)$$

- ▶ prior

1. Conjugate prior distribution
2. Jeffreys prior

- ▶ Hierarchical Bayes and an example.

# Frequentist Stats

- ▶ Large sample theory for the MLE
- ▶ p-value
- ▶ FDR

# Reference

- ▶ Peter Dalgaard, *Introductory Statistics with R*
- ▶ Winston Wang, *R Graphics Cookbook*
- ▶ <http://www.statmethods.net/>
- ▶ Kevin P. Murphy, *Machine Learning: A Probabilistic Perspective*