

## Bayesian Statistics and Hierarchical Bayesian Modeling for Psychological Science

Lecture 02

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# **Logical Operators**

Operator	Summary		
<	Less than		
>	Greater than		
<=	Less than or equal to		
>=	Greater than or equal to		
==	Equal to		
! =	Not equal to		
! <i>x</i>	NOT x		
x y	x OR y		
x&y	x AND y		

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### **Control Flow**

```
if (cond) {
    ..statement..
} else {
    ..statement..
}
```

```
if (cond) {
    ..statement..
} else if (cond) {
    ..statement..
} else {
    ..statement..
}
```

for-loop

```
for ( j in 1:J) {
    ..statement..
}
```

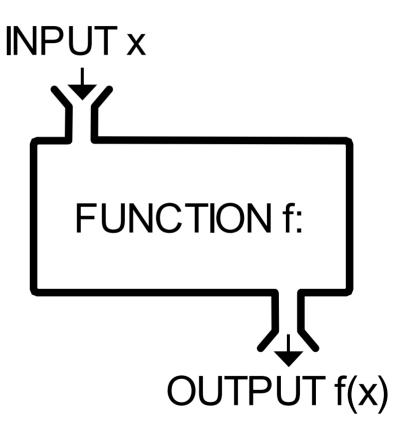
```
for ( j in 1:J ) {
    for ( k in 1:K ) {
        ..statement..
    }
}
```

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### **Functions**

The operation(s) to obtain some quantity, based on another quantity.

- built-in functions
- external functions (packages)
- user-defined functions



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### **User-defined Function**

```
funname <- function (input_arges) {
    .. function body ..
    .. function body ..
    return(output_arges)
}</pre>
```

$$sem = \sqrt{\frac{s^2}{n-1}}$$

```
sem <- function(x) {
   sqrt( var(x,na.rm=TRUE) / (length(na.omit(x))-1) )
}</pre>
```

## **Exercise II**

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```
.../01.R_basics/_scripts/R_basics.R
```

TASK: practise control flow and user-defined function

## **Exercise II**

- Generate a random number between 0 and I
- Compare it against I/3 and 2/3
- Print the random number and its position relative to 1/3 and 2/3.

```
# if-else
t <- runif(1) # random number between 0 and 1
if (t <= 1/3) {
    cat("t =", , ", t <= 1/3. \n")
} else if () {
    cat("t =", t, ", t > 2/3. \n")
} else {
    cat("t =", t, ", 1/3 < t <= 2/3. \n")
}</pre>
```

#### Example outcome:

```
t = 0.895, t > 2/3.
```

- Get the name of each month
- Print it one by one

```
# for-loop
month_name <- format(ISOdate(2018,1:12,1),"%B")
for (j in 1:length(month_name) ) {
   cat()
}</pre>
```

```
The month is January
The month is February
The month is March
The month is April
The month is May
The month is June
The month is July
The month is August
The month is September
The month is October
The month is November
The month is December
```

# Packages in R

R packages are collections of functions and data sets developed by the community, to make your life a lot easier!

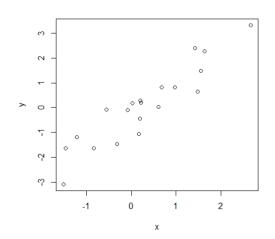
```
install.packages('ggplot2')
library(ggplot2)
detach('package:ggplot2')
```

## **Visualization**

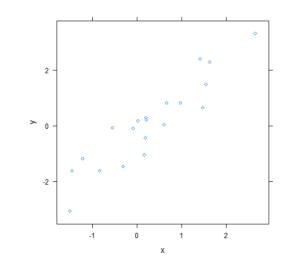
computing

- built-in plotting functions first attempt / quick look / exploratory
- {lattice} making nicer, similar to basic plotting functions (takes Im formulae)
- {ggplot2} making nicer, a layering philosophy

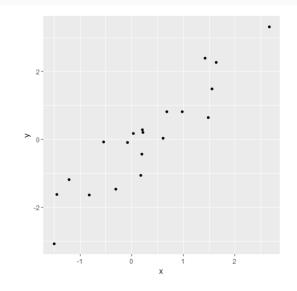
#### plot(x,y)



#### lattice::xyplot(y~x)



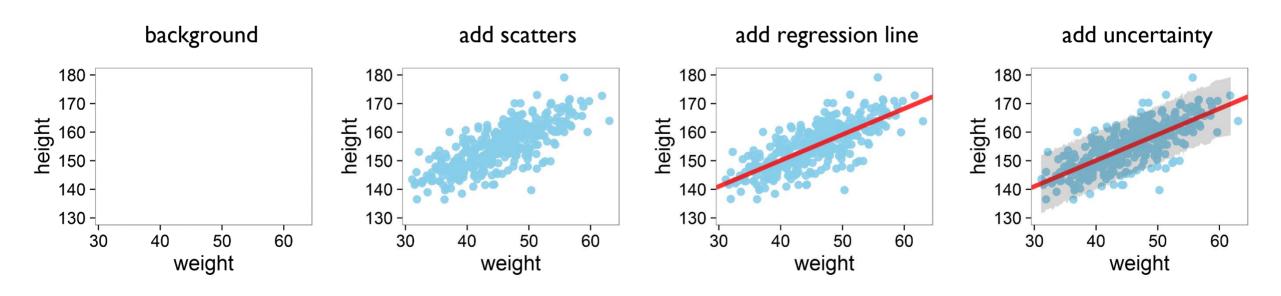
#### ggplot2::qplot(x,y)



# **Brief Intro to ggplot2**

computing

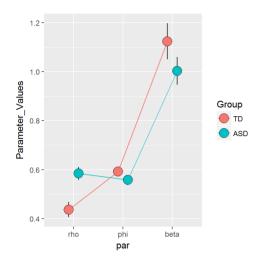
### game of adding layers!

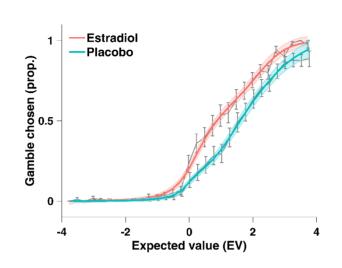


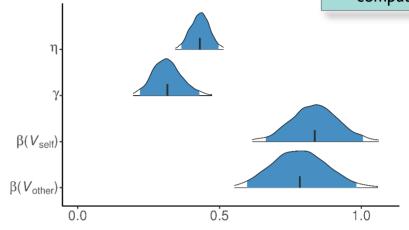
A taste of ggplot2

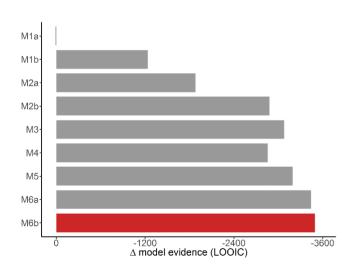
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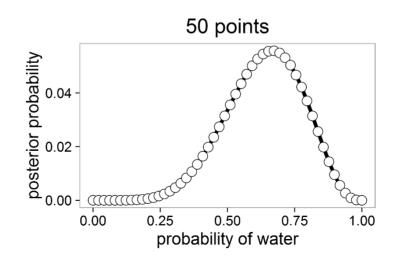
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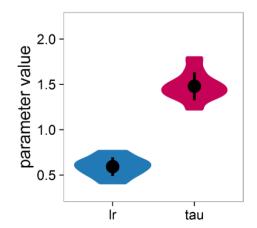








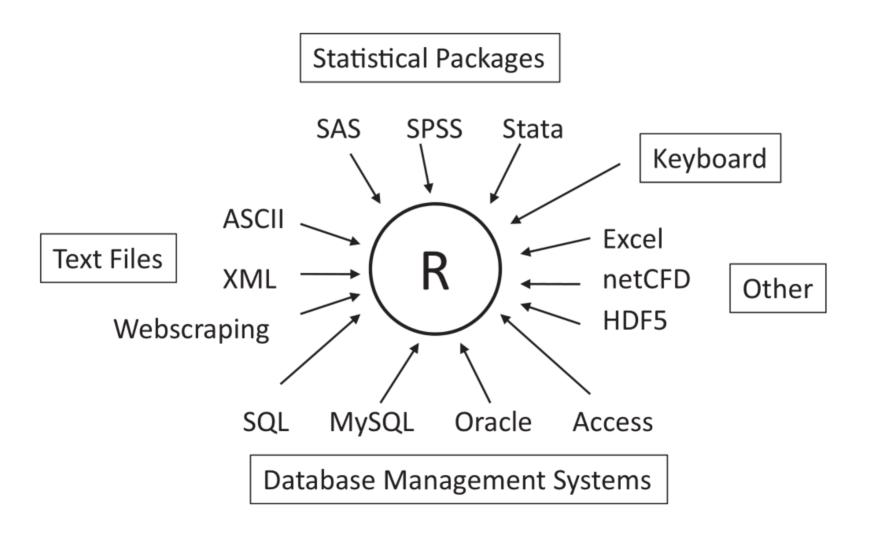






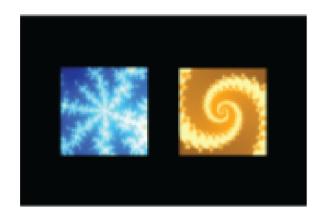
https://www.r-graph-gallery.com/

# **Data management**

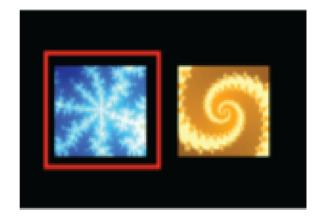


Kabacoff (2015)

# One simple experiment



choice presentation



action selection

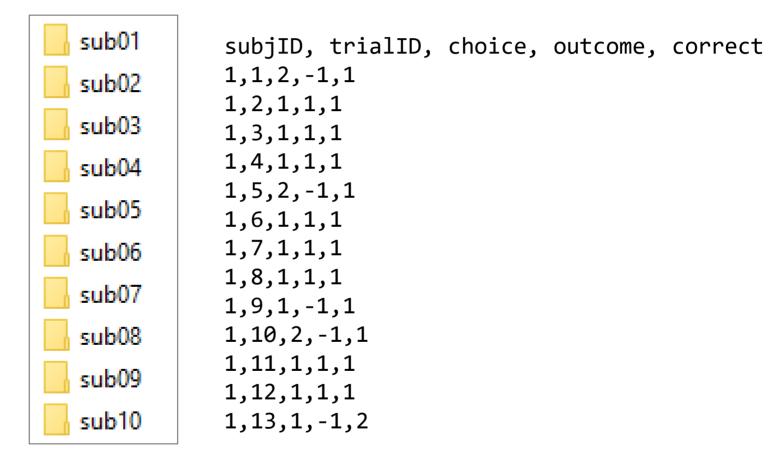


outcome

### The data

- nSub = 10
- nTrial = 80

./\_data/\_raw\_data/sub01/raw
\_data\_sub01.txt



## Import some data!

```
data dir = (' data/RL raw data/sub01/raw data sub01.txt')
data = read.table(data dir, header = T, sep = ",")
head(data)
  subjID trialID choice outcome correct
3
4
                     NA
5
               5
6
sum(complete.cases(data)) # number of valid trials
data = data[complete.cases(data),]
dim(data[complete.cases(data),])
```

# **Indexing**

```
data[1,1]
data[1,]
data[,1]
data[1:10,]
data[,1:2]
data[1:10, 1:2]
data[c(1,3,5,6), c(2,4)]

data$choice
```

>	data				
	subjID	trialID	choice	outcome	correct
1	1	1	1	1	1
2	1	2	1	1	1
3	1	3	1	1	1
5	1	5	1	-1	1
6	1	6	2	-1	1
7	1	7	1	1	1
8	1	8	1	1	1
9	1	9	1	1	1
10	1	10	1	1	1
11	l 1	11	1	1	1

### **Exercise III**

```
.../01.R_basics/_scripts/R_basics.R
```

```
TASK:

write a for loop

... which reads in each participant's raw data

... and reshape it in the "long format" by subj

TIP: complete line 173
```

```
for ( j in 1:n ) {
   read.table(file, header = T, sep = ",")
}
```

```
subID Choice
sub01
sub01 2
sub02 2
sub02 2
sub 10
sub10
```

### Read all the data!

```
ns = 10
data_dir = '_data/RL raw data'
rawdata = c();
for (s in 1:ns) {
    sub file = file.path(data dir, sprintf('sub%02i/raw data sub%02i.txt',s,s))
    sub data = read.table(sub file, header = T, sep = ",")
    rawdata = rbind(rawdata, sub data)
rawdata = rawdata[complete.cases(rawdata),]
rawdata$accuracy = (rawdata$choice == rawdata$correct) * 1.0
acc_mean = aggregate(rawdata$accuracy, by = list(rawdata$subjID), mean)[,2]
```

mean choice accuracy across trials, per participant.

### **Basic stats**

```
mean(acc mean)
sd(acc mean)
sem(acc mean)
t.test(acc mean, mu = 0.5) # one sample t-test
      One Sample t-test
data: acc mean
t = 13.788, df = 9, p-value = 2.34e-07
alternative hypothesis: true mean is not equal to 0.5
95 percent confidence interval:
 0.6962988 0.7733565
sample estimates:
mean of x
0.7348277
```

### **Basic correlation**

descriptive

10

subjID

```
1 123.98691 31.07218 0.8125
load(' data/RL descriptive.RData')
                                                             2 87.63187 30.13800 0.7125
descriptive$acc = acc mean
                                                             3 89.39930 23.44219 0.6875
df = descriptive
                                                             4 84.34607 27.44848 0.6500
                                                             5 134.72208 23.30624 0.7750
                                                             6 84.60797 25.67858 0.7250
cor.test(df$IQ, df$acc)
                                                             7 111.10238 24.36375 0.7750
                                                             8 117.89599 32.74026 0.8000
       Pearson's product-moment correlation
                                                             9 96.88233 22.80211 0.7500
                                                            10 76.01652 30.44258 0.6750
data: df$TO and df$acc
t = 4.8347, df = 8, p-value = 0.001297
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.5114810 0.9671586
sample estimates:
      cor
0.8631401
```

acc

### **Exercise IV**

```
.../01.R_basics/_scripts/R_basics.R
```

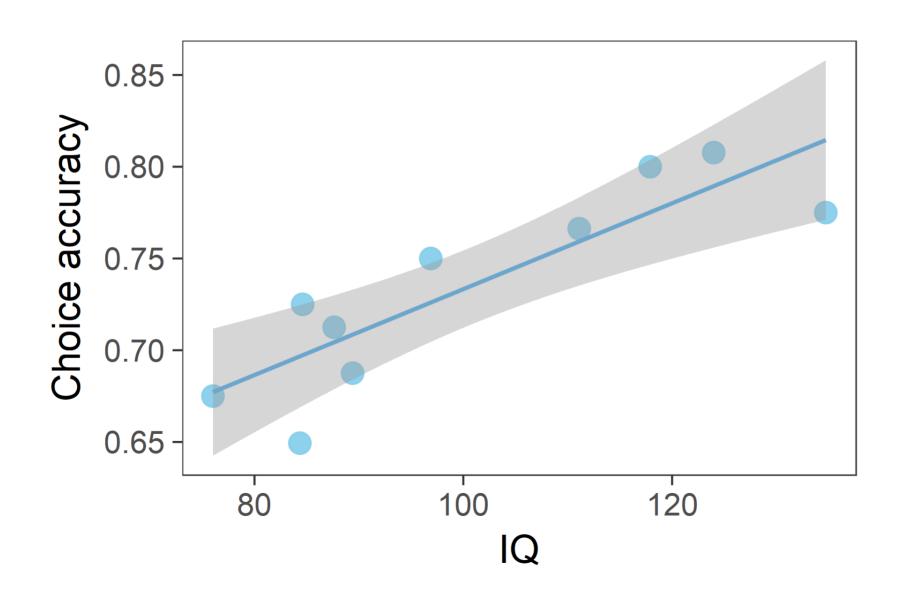
#### TASK:

Read in the descriptive data: \_\_data/descriptive.RData ...include 'acc\_mean' as a new column, and ...rename 'descriptive' as df.

Practice all the basic stats.

df\$new\_Col = new\_Col

# A simple linear regression



# What is exactly the regression line in R?

```
fit1 = 1m(acc \sim IQ, data = df)
                      summary(fit1)
                      Call:
                      lm(formula = acc \sim IQ, data = df)
                      Residuals:
egin{aligned} \mu_i &= lpha + eta x_i \ y_i &= \mu_i + oldsymbol{arepsilon} \end{aligned}
                            Min
                                        10 Median
                                                                        Max
                      -0.047305 -0.016277 0.007562 0.022577 0.027731
                      Coefficients:
                                   Estimate Std. Error t value Pr(>|t|)
                      (Intercept) 0.499292 0.049565 10.073 8.04e-06 ***
                                   0.002340 0.000484 4.835 0.0013 **
                      ΙQ
                      Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
                      Residual standard error: 0.02885 on 8 degrees of freedom
                      Multiple R-squared: 0.745, Adjusted R-squared: 0.7131
                      F-statistic: 23.37 on 1 and 8 DF, p-value: 0.001297
```

### **Exercise V**

.../01.R\_basics/\_scripts/R\_basics.R

#### TASK:

Read and make sense of the ggplot functions,

... experiment make some adjustments (color marker size etc.), and

... run the  $lm(acc \sim IQ)$