

GSP Workshop: Intermediate R: Statistics for Graduate Students

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Outline

Part I

- ▶ Conditionals and Control Work flows
 - ▶ Equality (or not)
 - ▶ & and |
 - ▶ The if statement

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 - ▶ & and |
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- ▶ Loops
 - ▶ Write a while loop
 - ▶ Write a for loop
 - ▶ Mix up loops with control flow

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Part II

- ▶ Functions and Arguments: Review

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- ▶ Functions and Arguments: Review
- ▶ Subset functions in R
 - ▶ Demonstration with data set
 - ▶ Resources for developing packages for tidying data

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 - ▶ Write a while loop
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Part II

- ▶ Functions and Arguments: Review
- ▶ Subset functions in R
 - ▶ Demonstration with data set
 - ▶ Resources for developing packages for tidying data
- ▶ Advanced: Tidyverse

Link to slides: [Intermediate-R](#)

Prelude: R Markdown

R Markdown provides an platform for data science. In an R Markdown file you can both

- ▶ Both save and execute code
- ▶ create reports and/or articles that can be shared to various audiences.

First:

```
```{r}  
install.packages("rmarkdown")
```
```

Here is a link to the code used : [R Code](#)

Here is a link to an R Markdown cheet sheet: [R Markdown](#)

Conditionals and Control Flow

A **logical statement** is a declarative sentence which conveys information about the truth of a statement.



TRUE



FALSE

Boolean

Figure: Figure credit: Boolean

Conditionals and Control Flow

A **logical statement** is a declarative sentence which conveys information about the truth of a statement.

- ▶ Examples 1 and 2: Equality (or not): If $x = 3, y = 4$ then we can assess the truth about the following statements:

```
> #Example 1: Equality  
>  
> x=3  
> y=4  
> x==y  
[1] FALSE
```

Conditionals and Control Flow

A **logical statement** is a declarative sentence which conveys information about the truth of a statement.

- Examples 1 and 2: Equality (or not): If $x = 3, y = 4$ then we can assess the truth about the following statements:

```
> #Example 1: Equality
>
> x=3
> y=4
> x==y
[1] FALSE
```

```
> #Example 2: Not equality
>
> x!=y
[1] TRUE
>
```

Conditionals and Control Flow

- ▶ Example 3: Equality (or not): Comparison of strings.
 - ▶ Set the two vectors below equal to each other
 - ▶ This creates a logical vector (with TRUE/FALSE)

```
user1<-c("userR", "user", "UserR")
```

```
user2<-c("userR", "userR", "UserR")
```

```
user1==user2 #this is a logical vector
```

```
[1] TRUE FALSE TRUE
```

Conditionals and Control Flow

- ▶ Example 3: Equality (or not): Comparison of strings.
 - ▶ Set the two vectors below equal to each other
 - ▶ This creates a logical vector (with TRUE/FALSE)

```
user1<-c("userR", "user", "UserR")  
user2<-c("userR", "userR", "UserR")
```

```
user1==user2 #this is a logical vector  
[1]  TRUE FALSE  TRUE
```

Demonstration: Let's use logical operators to subset our data.

Conditionals and Control Flow

- ▶ Example 5: The greater than ($>$), and less than ($<$) operators. Dealing with unusual points:

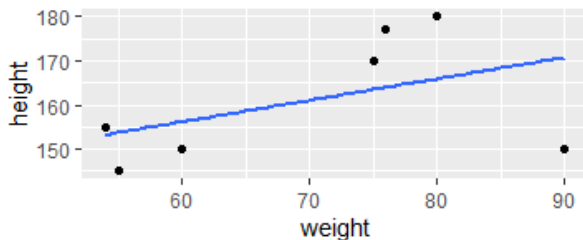


Figure: There appears to be an unusual point at (90, 150)

Conditionals and Control Flow

- ▶ Example 5: The greater than ($>$), and less than ($<$) operators. Dealing with unusual points:

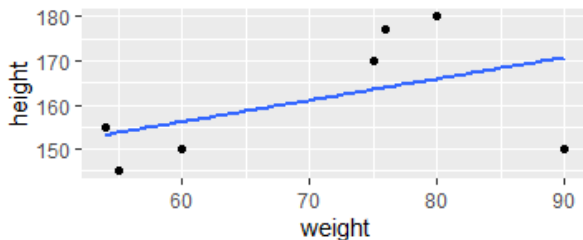


Figure: There appears to be an unusual point at (90, 150)

- . **Demonstration:** Let's use logical operators to subset our data.

Conditionals and Control Flow

- ▶ Example 6: The & and | operators. Dealing with unusual points:

```
{r}  
#recall  
features |
```

| height
<dbl> | weight
<dbl> |
|-----------------|-----------------|
| 150 | 60 |
| 145 | 55 |
| 150 | 90 |
| 155 | 54 |
| 170 | 75 |
| 180 | 80 |
| 177 | 76 |

7 rows

Figure: There appears to be an unusual point at (90, 150)

Conditionals and Control Flow

- ▶ Example 6: The & and | operators:

```
outlier <-features[ which(weight==90 & height==150), ]  
outlier
```

```
features3<-features[ which(weight==90 | height==150), ]  
features3
```


Conditionals and Control Flow

- ▶ Example 6: The & and | operators:

```
outlier <-features[ which(weight==90 & height==150), ]  
outlier
```

```
features3<-features[ which(weight==90 | height==150), ]  
features3
```

```
> outlier  
# A tibble: 1 x 2  
  height weight  
  <dbl> <dbl>  
1    150     90  
>
```

(a) The subset with & operator

```
>  
> features3  
# A tibble: 2 x 2  
  height weight  
  <dbl> <dbl>  
1    150     60  
2    150     90  
>
```

(b) The subset with | operator.

Figure: Data frames created from the AND and OR logical operators

Loops

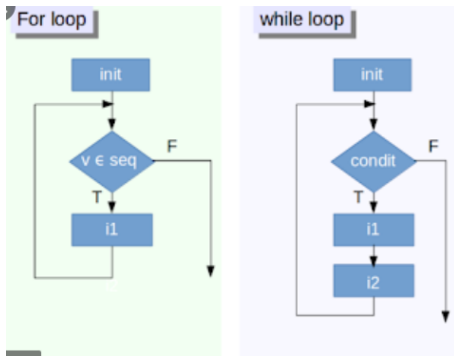


Figure: Credit: A figure of a for and while loop loops

A **loop** is a replication of instructions. It's a mini-step process that organizes a sequence of actions into parts that need to be repeated. Diamonds, on the other hand, are called **decision symbols**, and translate into questions which only have two possible logical answers; TRUE or FALSE.

Loops

Example 7: For loop

- ▶ Start with an initialization (e.g. vector or matrix) of empties.
- ▶ Tests to whether a current value is within a specified defined range (i.e. within 1:100).
- ▶ If the condition is not met and the resulting outcome is False, the loop is never executed.

Loops

Example 7: For loop

- ▶ Start with an initialization (e.g. vector or matrix) of empties.
- ▶ Tests to whether a current value is within a specified defined range (i.e. within 1:100).
- ▶ If the condition is not met and the resulting outcome is False, the loop is never executed.

```
set.seed(1)
rv <- rnorm(1000, 0, 1)
usq<-matrix("NA", 100, 1)

for(i in 1:100) {
  usq[i] <- rv[i]*rv[i]
  print(usq[i])
}
usq<-data.frame(usq=as.numeric(unlist(usq)))
attach(usq)
```

Loops

Example 7: For loop

- ▶ Once the condition is past 100, the evaluation is False, and the loop ends:

```
> dim(usq)
[1] 100    1
>
```

- ▶ The data frame is entries for the _____ distribution.

Loops

Example 7: For loop

- ▶ Once the condition is past 100, the evaluation is False, and the loop ends:

```
> dim(usq)
[1] 100    1
>
```
- ▶ The data frame is entries for the _____ distribution.
- ▶ The data frame is entries for the chi distribution (denoted χ_1^2).

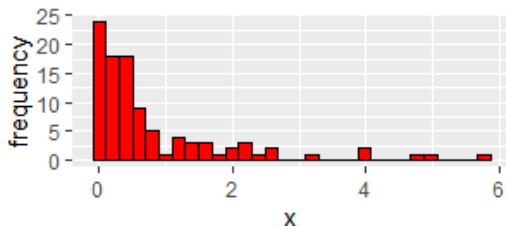


Figure: The square of a standard normal is chi

Loops

Example 8: While loop

- ▶ Start with an initialization (e.g. vector or matrix) of empties.
- ▶ Now followed by a logical comparison between a control variable and a value (using earlier defined conditionals).
- ▶ If the condition is not met and the resulting outcome is False, the loop is never executed.

Loops

Example 8: While loop

- ▶ Start with an initialization (e.g. vector or matrix) of empties.
- ▶ Now followed by a logical comparison between a control variable and a value (using earlier defined conditionals).
- ▶ If the condition is not met and the resulting outcome is False, the loop is never executed.

```
set.seed(1)
usq<-matrix("NA", 100, 1)
n = length(usq)
x=0
while (x<=n) {
  x<-sum(x, 1)
  i<-sum(i, 1)
  usq[i] <- rv[i]*rv[i]
  print(usq[i])
}
usq<-data.frame(usq=as.numeric(unlist(usq)))
attach(usq)
```


Loops

Example 8 : While loop

- ▶ Once the condition is past 100, the evaluation is False, and the loop ends:

```
> dim(usq)
[1] 100    1
>
.
```

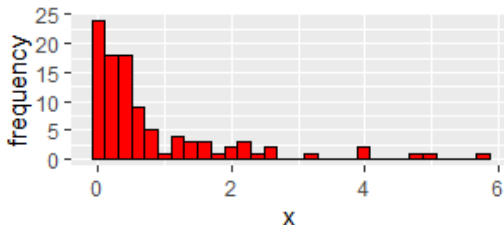


Figure: The square of a standard normal is chi

Note: The while loop was constructed to yield the same output as the for loop. This was less efficient.

Loops

Example 8: Mix up loops with control flow

- ▶ Example 4 (revisted): Outlier point in *weight* and height.

weight

height

```
# Code the for loop with conditionals
for (i in weight) {
  if (i > 80) {
    print("You're an outlier!")
  }
  else {
    print("Nothing to see here!")
  }
  print(i)
}
```

Loops

Example 8: Mix up loops with control flow

- ▶ Example 4 (revisted): Outlier point in *weight* and height.

```
weight
```

```
height
```

```
# Code the for loop with conditionals
for (i in weight) {
  if (i > 80) {
    print("You're an outlier!")
  }
  else {
    print("Nothing to see here!")
  }
  print(i)
}
```

Demonstration: This is a bit silly but let's see what we get.

Part II

Question: What have we examined and why?

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1. Conditionals and logical statements
2. Loops and the role of logical statements/conditions
3. The role of if statements in executing logicals

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1. Conditionals and logical statements
2. Loops and the role of logical statements/conditions
3. The role of if statements in executing logicals

Answer: All of these components are important features of tidying data in R.

Functions: Review

An **R function** is created by using the keyword `*function*`. The basic syntax of an R function definition is as follows:

```
function_name <- function(arg_1,...) {  
Function_body  
}
```

The different parts of a function are:

- ▶ **Function Name** = This is the actual name of the function. It is stored in the R environment as its name.
- ▶ **Arguments** - This is an placeholder to pass an input value into your function (i.e. $f(x)$).
- ▶ **Function Body** - The function contains the statements that determines what the function does.

Example 9: A perfect square in R

```
>
> perfect.squares<-function(x){for (x in 1:10){
+   if(x%%x == 0)
+     b<-x^2
+     print (b)
+   }}
> perfect.squares(10)
```

Figure: A function in R to compute perfect squares of all the numbers in the range of 1 to 10

Example 9: A perfect square in R

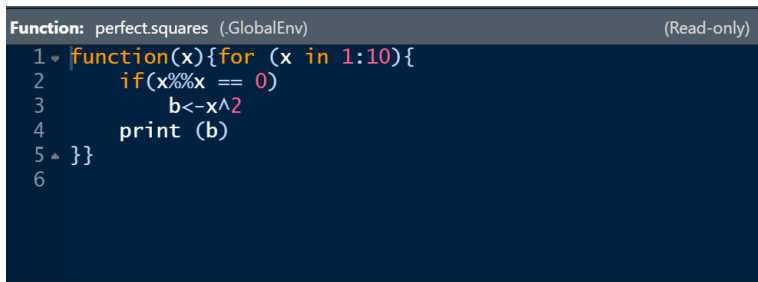
```
>
> perfect.squares<-function(x){for (x in 1:10){
+   if(x%%x == 0)
+     b<-x^2
+     print (b)
+   }}
> perfect.squares(10)
```

Figure: A function in R to compute perfect squares of all the numbers in the range of 1 to 10

Demonstration: Let's see what we have printed.

Example 9: A perfect square in R

Here is how the function is stored as an object in the R environment:



The screenshot shows an R console window with a dark blue background. At the top, a grey header bar displays "Function: perfect.squares (.GlobalEnv)" on the left and "(Read-only)" on the right. Below the header, the function code is displayed in a light-colored font. The code is as follows:

```
1 ▾ function(x){for (x in 1:10){  
2     if(x%%x == 0)  
3         b<-x^2  
4     print (b)  
5 ▲ }}  
6
```

Figure: A function in R stored in the global environment.

Example 10: Processing Grades in R

Here is a function that evaluates at two inputs:

```
translate<-function(x,y){  
  if (x=="NA"){x=0}  
  else if (x!="NA"){x=x}  
  print(x)  
  if (y== -1){y="NA"}  
  else if (y!=-1){y=y}  
  print(y)  
}
```

- ▶ If x is "NA" then it will assign a value of 0.
- ▶ else if y is equal to -1 then it will assign a value of "NA"
- ▶ else let y=y.

Example 10: Processing Grades in R

Here is the output at different pairs of (x, y) .

```
> #Testing the function with input values
> x<-"NA"
> y<:"-1"
>
> x2<-1
> y2<-0
>
> translate(x, y)
[1] 0
[1] "NA"
> translate(x2, y2)
[1] 1
[1] 0
> |
```

Figure: A function in R stored in the global environment.

Question: Does anyone see a problem with the way in which the function is evaluating the output?

Example 10: Processing Grades in R

Here is the output at different pairs of (x, y) .

```
> #Testing the function with input values
> x<-"NA"
> y<:"-1"
>
> x2<-1
> y2<-0
>
> translate(x, y)
[1] 0
[1] "NA"
> translate(x2, y2)
[1] 1
[1] 0
> |
```

Figure: A function in R stored in the global environment.

Question: Does anyone see a problem with the way in which the function is evaluating the output?

- Ans: It does so term by term, which is inefficient if we have a vector of entries that we want to process (let's see).

Example 11: ifelse function in R

In order to understand the brilliance of this function let's go back to Example 10.

```
> x3<-c(0, "NA", 1)
> y3<-c(-1, "NA", 80)
>
> translate(x3,y3)
[1] "0"  "NA" "1"
[1] "NA"
Warning messages:
1: In if (x == "NA") { :
  the condition has length > 1 and only the first element will be used
2: In if (x != "NA") { :
  the condition has length > 1 and only the first element will be used
3: In if (y == -1) { :
  the condition has length > 1 and only the first element will be used
> |
```

Figure: The translate function in R with vector inputs.

Example 11: ifelse function in R

The **'ifelse'** is a built in base R function that returns a value with the same length as the test, rather than the evaluation at the first element.

```
x3<-c(0, "NA", 1)
> z<-x3
> ifelse(z==0, "NA", z)
[1] "NA" "NA" "1"
>
```

Aside:

- ▶ In statistics education it is common to use placeholders of "-1" when a student has a valid documentation for absence.
- ▶ If no valid documentation is received, the student receives a value of 0.
- ▶ The ifelse function evaluates these condition term by term on the entire vector.

Subset function in R

```
library(myPackage)
attach(Grades)
view(Grades)
```

```
head(Grades)
```

```
TT1 = `Term Test 1 (/505)`
```

```
...
```

Term Test 1 (/505)

<dbl>

| | |
|---|----|
| 1 | 34 |
| 2 | 36 |
| 3 | 26 |
| 4 | 45 |
| 5 | 18 |
| 6 | 34 |

6 rows | 1-2 of 4 columns

Figure: Processed marks for an undergraduate statistics course.

Subset function in R

The **subset** function is the easiest way to select variables and observations (at least within base R).

Example 10-11 revisited: Grade Processing: Suppose that I want to select only those students who have no noted absences in Term Test 1 (TT1) and Term Test 2 (TT2)

```
dim(Grades)
```

```
newdata <-newdata <- subset(Grades, TT1 !=-1 | TT2!= -1,)  
dim(newdata)
```

```
newgrades <- subset(Grades, TT1 !=-1 & TT2!= -1,)  
dim(newgrades)
```

Subset function in R

The **subset** function is the easiest way to select variables and observations (at least within base R).

Example 10-11 revisited: Grade Processing: Suppose that I want to select only those students who have no noted absences in Term Test 1 (TT1) and Term Test 2 (TT2)

```
dim(Grades)
```

```
newdata <-newdata <- subset(Grades, TT1 !=-1 | TT2!= -1,)  
dim(newdata)
```

```
newgrades <- subset(Grades, TT1 !=-1 & TT2!= -1,)  
dim(newgrades)
```

Demonstration: What is the difference?

new.function() function in myPackage

The data set *newgrades* is closer to what would use to process grades for undergraduate students.

- ▶ This is still not the ideal realization of my grades vectors (I need to account for 0s).

```
> view(new.grades)
> new.grades
function(x)
{
  x=ifelse(x=="NA", 0,
           ifelse(x==-1, "NA", x))
}
<bytecode: 0x0000023167a2b188>
<environment: namespace:myPackage>
> |
```

Figure: The new.grades function translates "-1" to "NA" and "NA" to "0".

Demonstration: Revisit the TT1 and TT2 grades in the Statistics data set.

Tidyverse

Next information session: The **tidyverse** package is a collection of R packages designed for data science.

- ▶ Efficiency of data clean up
- ▶ Tidy pipeline structures
- ▶ Sophisticated plots (in fact I used ggplot here)

Resource: [Tidyverse](#)