

**Hacettepe University  
Department of Industrial Engineering  
Undergraduate Program  
2023-2024 Fall**

**EMU 430 – Data Analytics  
Week3  
October 20, 2023**

**Instructor:** Erdi Dasdemir

[edasdemir@hacettepe.edu.tr](mailto:edasdemir@hacettepe.edu.tr)  
[www.erdidasdemir.com](http://www.erdidasdemir.com)

20 Ekim 2023  
15:00



# Endüstri Müh. ve Veri Bilimi

Hacettepe Üniversitesi Endüstri Mühendisliği Bölümü  
Veri Bilimi ve Yöneylem Araştırması Üzerine  
Davetli Konuşmalar Serisi

EMÜ 430 ve 679 dersleri



Konuşmacı

**Cem Vardar**

Data Scientist & Co-Founder  
of Decision Science Lab &  
Former Data Scientist at  
Intel, Revionics and Carvana



Moderatör


**Erdi Dasedemir**

Data Scientist & Assistant  
Professor of Industrial  
Engineering

- We learn R because it greatly facilitates data analysis and implementation of analytical approaches.
- However, R is not just a data analysis environment, but a programming language.
- Advanced R programmers can develop user applications and perform other complex programming tasks.


- Three fundamental programming concepts:
  - conditional execution,
  - iteration,
  - and creating functions.
  
- These are not only foundational elements of computer programming but are frequently useful in the context of data analysis.

# Conditional Execution



6/32

# Functions



22/32

# Iterations



29/32

# Conditional Execution

## Logical (boolean) expressions

- A logical (boolean) expression is an expression that is either **true** or **false**.

```
430 == 430
```

```
[1] TRUE
```

```
430 == 679
```

```
[1] FALSE
```

## Comparison Operators

`x == y` # x is equal to y

`x != y` # x is not equal to y

`x > y` # x is greater than y

`x < y` # x is less than y

`x >= y` # x is greater than or equal to y

`x <= y` # x is less than or equal to y



## Logical Operators

& (and), |(or), !(not)

```
x <- 500
```

```
x > 430 & x < 679
```

```
[1] TRUE
```

```
x <- 300
```

```
x > 430 & x < 679
```

```
[1] FALSE
```

## General form:

```
if (boolean condition) {  
    expressions  
} else {  
    alternative expressions  
}
```

### Conditional statements

x <- 430

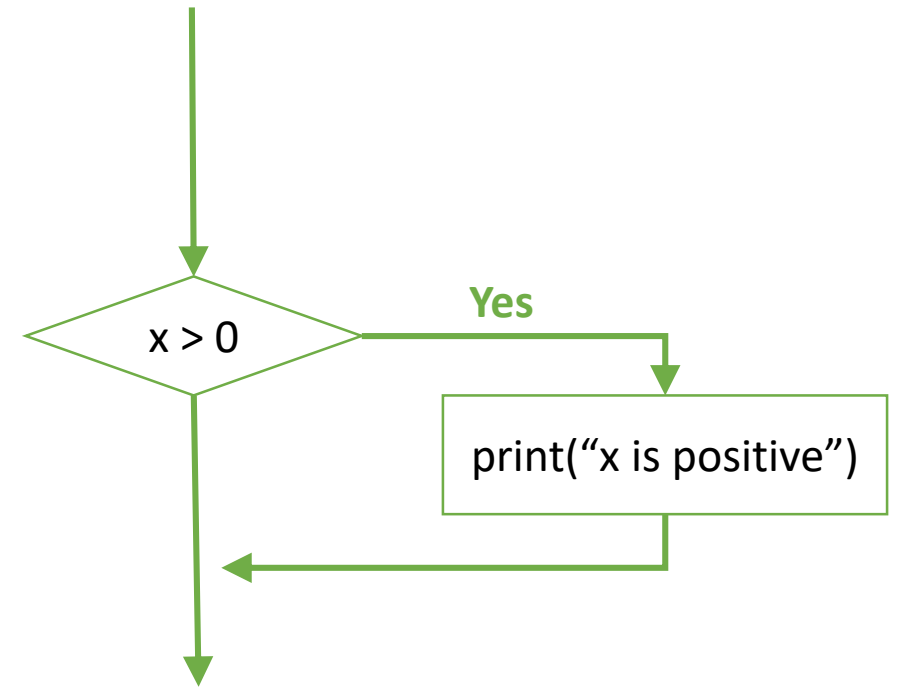
```
if (x > 0) {
  print("x is positive")
}
```

my recommended format

```
if (x > 0) {print("x is positive")}
```

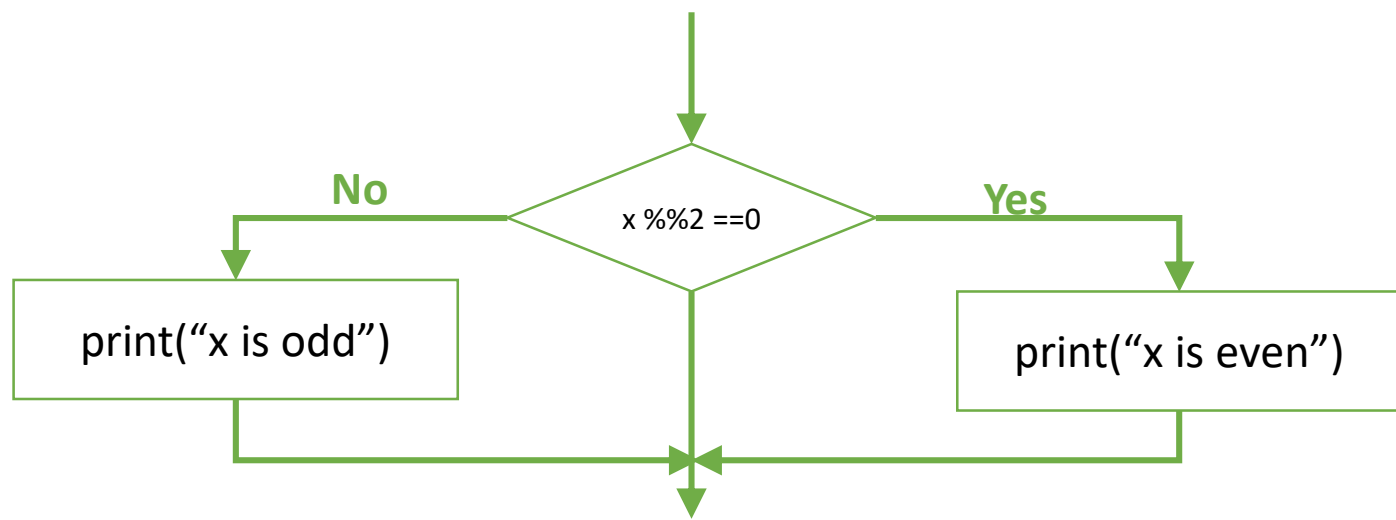
```
if (x > 0)
  print("x is positive")
```

```
if (x > 0)
print("x is positive")
```



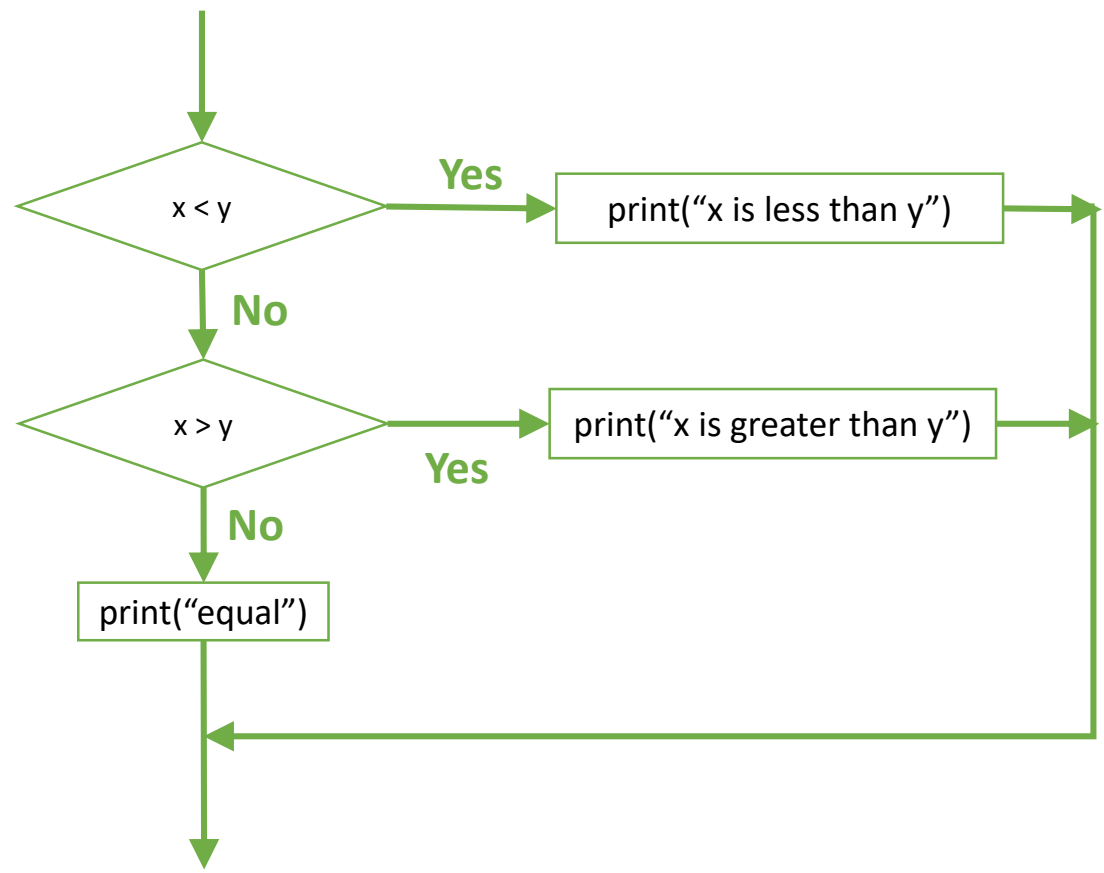
### Alternative Execution

```
x <- 430  
  
if (x %% 2 == 0) {  
  print("x is even")  
} else {  
  print("x is odd")  
}
```



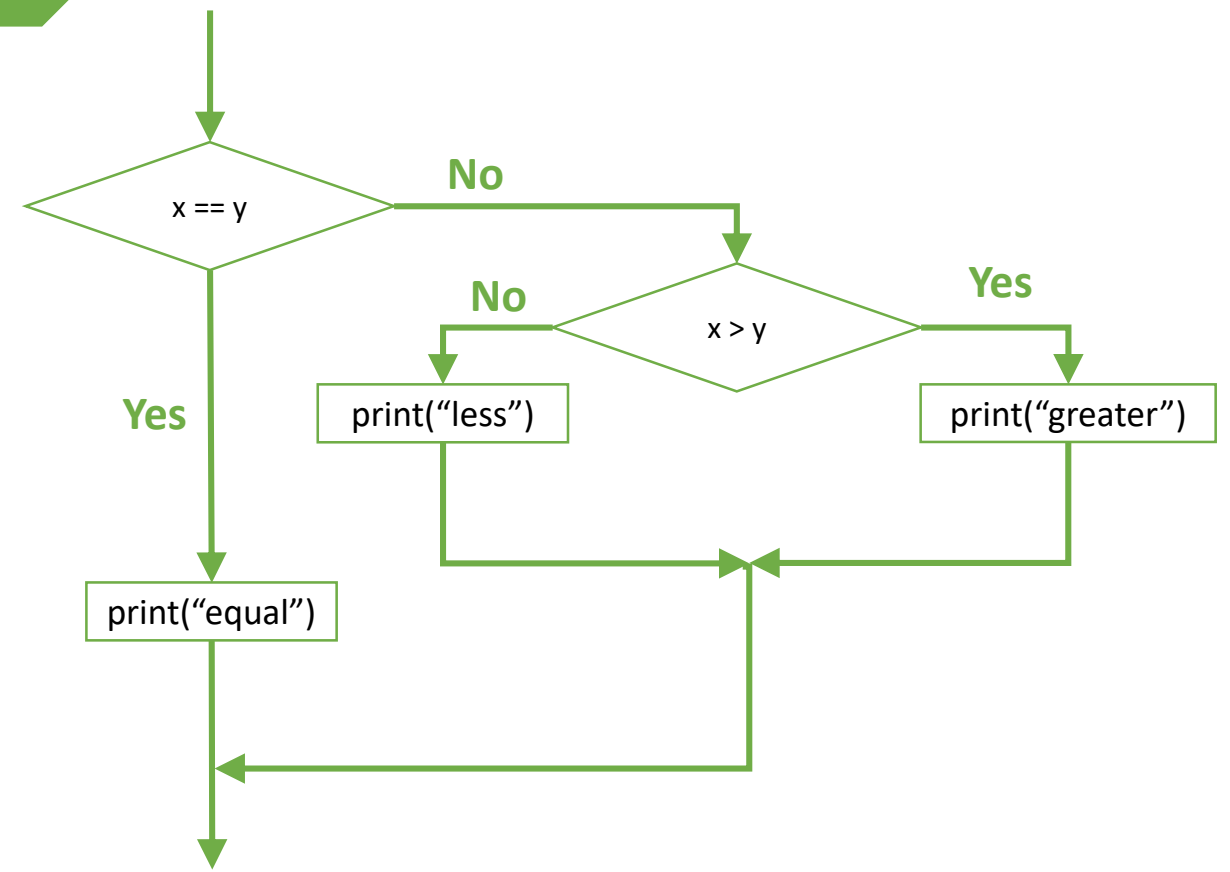
### Chained Conditionals

```
if (x < y) {  
  print("x is less than y")  
} else if (x > y) {  
  print("x is greater than y")  
} else {  
  print("x and y are equal")  
}
```



### Nested Conditionals

```
if (x == y) {  
  print("x and y are equal")  
} else {  
  if (x < y) {  
    print("x is less than y")  
  } else {  
    print("x is greater than y")  
  }  
}
```



## Nested Conditionals

```
if (0 < x){  
    if (x < 10) {  
        print("x is a positive single-digit number")  
    }  
}
```

```
if (0 < x & x < 10) {  
    print("x is a positive single-digit number")  
}
```

## Nested Conditionals

```
if (0 < x){  
    if (x < 10) {  
        print("x is a positive single-digit number")  
    }  
}
```

```
if (0 < x & x < 10) {  
    print("x is a positive single-digit number")  
}
```



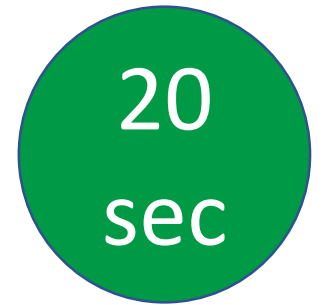
## Murders Data Set

```
library(ds1abs)
data(murders)
murder_rate <- (murders$total/murders$population)*100000
```

If the murder rate of the state with the lowest murder rate is lower than 0.5, print the name of that state.

```
ind <- which.min(murder_rate)
if (murder_rate[ind] < 0.5) {
  print (murders$state[ind])
} else {
  print("No state has murder rate that low")
}
```

Output: [1] Vermont



## Murders Data Set

If we change the threshold level to 0.25

```
ind <- which.min(murder_rate)
if (murder_rate[ind] < 0.25) {
  print (murders$state[ind])
} else {
  print("No state has murder rate that low")
}
```

Output: [1] "No state has murder rate that low"

## ifelse statement

```
a <- 0
ifelse(a > 0, 1/a, NA)
Output: [1] NA
```

```
a <- 5
ifelse(a > 0, 1/a, NA)
Output: [1] 0.2
```

ifelse {base}

## Conditional Element Selection

### Description

`ifelse` returns a value with the same shape as `test` which is filled with elements selected from either `yes` or `no` depending on whether the element of `test` is `TRUE` or `FALSE`.

### Usage

```
ifelse(test, yes, no)
```

### Arguments

- `test` an object which can be coerced to logical mode.
- `yes` return values for true elements of `test`.
- `no` return values for false elements of `test`.

**ifelse** is particularly useful because it works with vectors.

## ifelse statement with vectors

```
a <- c(0, 1, 2, -4, 5)
```

```
ifelse(a > 0, 1/a, NA)
```

```
Output: [1] NA 1.0 0.5 NA 0.2
```

<b>a</b>	<b>is_a_positive</b>	<b>answer1</b>	<b>answer2</b>	<b>result</b>
<b>0</b>	<b>FALSE</b>	<b>Inf</b>	<b>NA</b>	<b>NA</b>
<b>1</b>	<b>TRUE</b>	<b>1.00</b>	<b>NA</b>	<b>1.0</b>
<b>2</b>	<b>TRUE</b>	<b>0.50</b>	<b>NA</b>	<b>0.5</b>
<b>-4</b>	<b>FALSE</b>	<b>0.25</b>	<b>NA</b>	<b>NA</b>
<b>5</b>	<b>TRUE</b>	<b>0.20</b>	<b>NA</b>	<b>0.2</b>

A common usage of ifelse is replacing NAs with some other value.

```
data(na_example)
```

```
sum(is.na(na_example)) # there are 145 NAs
```

```
Output: [1] 145
```

```
# convert na_example to a vector that does not have any NAs
```

```
no_nas <- ifelse(is.na(na_example), 0, na_example) # note that the last argument is also a vector
```

```
sum(is.na(no_nas))
```

```
Output: [1] 0
```

## *any* and *all* functions

# any function takes a vector of logical and returns if any element is true

```
z <- c(TRUE, TRUE, TRUE)
```

```
any(z) → Output: [1] TRUE
```

```
all(z) → Output: [1] TRUE
```

```
z <- c(TRUE, TRUE, FALSE)
```

```
any(z) → Output: [1] TRUE
```

```
all(z) → Output: [1] FALSE
```

# Functions

- Perform the same operations over and over.
- Example:

You compute the average all the time when you are doing data science

`sum(x) / length(x)`

Write a function that does this calculation:

`mean()` already exists.

- In many situations, the function that you need is not defined. → You have to write your own.

```
avg <- function(x) {  
  s <- sum(x)  
  n <- length(x)  
  s/n  
}
```

```
x <- 1:100  
avg(x)  
[1] 50.5
```

```
identical(mean(x), avg(x))  
[1] TRUE
```

- **General form:**

```
my_function <- function(x, y, z) {
```

operations that operate on x, y, z, which are defined by the user when they call this function.

```
}
```

- Functions are objects, so we assign them to variable names.

- Define a function that does arithmetic or geometric mean calculation

```
avg <- function(x, arithmetic = TRUE){
```

```
  n <- length(x)
```

```
  ifelse(arithmetic, sum(x)/n, prod(x)^(1/n))
```

```
}
```



- **Lexical Scope:**
- Variables defined inside a function are not saved in the workspace.

```
avg <- function(x){  
  s <- sum(x)  
  n <- length(x)  
  s/n  
}
```

```
s <- 3  
avg(1:10)  
S  
Output: [1] 3
```

Inside the function, an s is created that is not 3, it is something else.  
But that only happens inside the function.

- Return Value

```
# Version 1: A function to calculate the area of a rectangle
calculate_rect_area <- function(width, height){
  return(width * height) # return a specific result
}
```

```
# Version 2: A function to calculate the area of a rectangle
calculate_rect_area <- function(width, height){
  width * height # return a specific result
}
```

```
# Version 3: A function to calculate the area of a rectangle
calculate_rect_area <- function(width, height){
  area <- width * height # calculate area
  area
}
```

- **Debugging Functions**

```
# version 3: A function to calculate the area of a rectangle
calculate_rect_area <- function(width, height){
  area <- width * height # calculate area
  area
}
```

**assign sample values to your arguments, and then run through the function line by line.**

# Iterations

## ○ for-loops

```
for (i in range of values) {  
    operations that use i, which is changing across the  
    range of values  
}
```

- Example

$$1 + 2 + \dots + n = \frac{n(n+1)}{2}$$

We want to do this calculation for  $n = 1, 2, \dots, 25$

1. Define a function:

```
compute_s_n <- function(n){
  x <- 1:n
  sum(x)
}
```

```
compute_s_n(3)
```

```
[1] 6
```

```
compute_s_n(100)
```

```
[1] 5050
```

2. calculation for  $n = 1, 2, \dots, 25$

```
m <- 25
```

```
# create an empty vector
```

```
s_n <- vector(length = m)
```

```
for (i in 1:m){
```

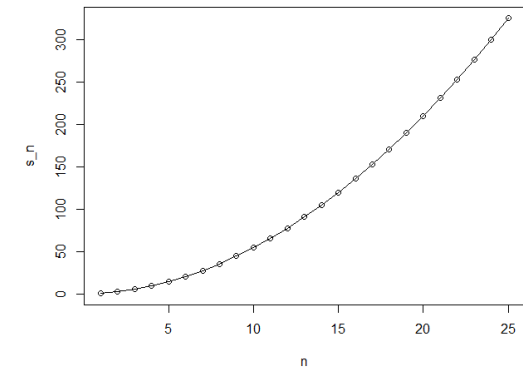
```
  s_n[i] <- compute_s_n(i)
```

```
}
```

```
n <- 1:m
```

```
plot(n, s_n)
```

```
lines(n, n*(n+1)/2)
```



We normally rarely use for loops in R

R functions:

apply

sapply

lapply

tapply

mapply

```
m <- 25
# create an empty vector
s_n <- vector(length = m)
for (i in 1:m){
  s_n[i] <- compute_s_n(i)
}
```



```
sapply(1:m, compute_s_n)
```

## ○ while

```
n <- 430
while (n > 0){
  print(n)
  n <- n - 50
}
```

```
[1] 430
[1] 380
[1] 330
[1] 280
[1] 230
[1] 180
[1] 130
[1] 80
[1] 30
```



