

# Data Structures: Data Frames

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PSICOSTAT

# A type of data structure you are already familiar with

	A	B	C	D	E	F	G	H	I
1	Year	Type of course	Title	Teacher	Hours	ECTS	Mandatory	Delivery	Language
2	1	METHODOLOGY	CURRENT ISSUES IN STATISTICAL INFERENCE FOR MASSIMILIANO PASTORE		10	2	YES	IN PERSON	ENGLISH
3	1	METHODOLOGY	LINEAR AND MIXED EFFECT MODELS WITH SPSS	JEFF KIESNER	15	3	NO	IN PERSON	ENGLISH
4	1	METHODOLOGY	BASES OF STATISTICAL INFERENCE WITH R	UMBERTO GRANZIOL	20	4	YES	IN PERSON	ENGLISH
5	1	PROGRAMMING	BASES OF R FOR DATA SCIENCE	ENRICO TOFFALINI	10	2	YES	IN PERSON	ENGLISH
6	1	METHODOLOGY	PSYCHOLOGICAL MEASUREMENT	LUCA STEFANUTTI	15	3	YES	IN PERSON	ENGLISH
7	1	METHODOLOGY	POWER AND DESIGN ANALYSIS	GIANMARCO ALTOE	5	1	YES	IN PERSON	ENGLISH
8	1	METHODOLOGY	EVALUATION OF OUTLIERS AND INFLUENTIAL	GIANMARCO ALTOE	5	1	NO	IN PERSON	ENGLISH
9	1	METHODOLOGY	QUESTIONABLE MEASUREMENT PRACTICES AND	TATIANA MARCI	5	1	YES	IN PERSON	ENGLISH
10	1	METHODOLOGY	DATA VISUALISATION WITH GGPLOT2	MICHELE VICOVARO	5	1	NO	IN PERSON	ENGLISH
11	1	SOFT SKILLS	CRAFTING EFFECTIVE SCIENTIFIC	FILIPPO GAMBAROTA	5	1	NO	IN PERSON	ENGLISH
12	1	PROGRAMMING	BASES OF MATLAB FOR DATA SCIENCE	LUCA STEFANUTTI	10	2	NO	IN PERSON	ENGLISH
13	1	PROGRAMMING	BASES OF PYTHON FOR DATA SCIENCE	ENRICO TOFFALINI	10	2	NO	IN PERSON	ENGLISH
14	2-3	SOFT SKILLS	ADVANCING RESEARCH PARADIGMS: OPEN	GIULIA CALIGNANO	5	1	NO	IN PERSON	ENGLISH
15	2-3	THEMATIC COURSE	NEUROPSYCHOLOGY OF VISION	LUCA BATTAGLINI	5	1	NO	IN PERSON	ENGLISH
16	2-3	METHODOLOGY	METHODOLOGY IN NEUROSCIENCES	SIMONE CUTINI	10	2	NO	IN PERSON	ENGLISH
17	2-3	METHODOLOGY	BAYESIAN DATA ANALYSIS IN PSYCHOLOGICAL	MASSIMILIANO PASTORE	10	2	NO	IN PERSON	ENGLISH
18	2-3	METHODOLOGY	GENERALISED LINEAR MODELS	FILIPPO GAMBAROTA	15	3	NO	IN PERSON	ENGLISH
19	2-3	METHODOLOGY	STRUCTURAL EQUATION MODELING	TOMMASO FERACO	20	4	NO	IN PERSON	ENGLISH
20	2-3	METHODOLOGY	CONDUCTING SYSTEMATIC REVIEWS	ENRICO SELLA	5	1	NO	IN PERSON	ENGLISH
21	2-3	METHODOLOGY	INTRODUCTION TO ITEM RESPONSE THEORY	MARINA OTTAVIA EPIFANIA	15	3	NO	IN PERSON	ENGLISH
22	2-3	SOFT SKILLS	HOW TO WIN RESEARCH GRANTS	CHRISTIAN AGRILLO	5	1	NO	IN PERSON	ENGLISH
23	2-3	SOFT SKILLS	CAREER COUNSELING	NICOLA CELLINI	10	2	NO	IN PERSON	ENGLISH
24	2-3	SOFT SKILLS	OUTSIDE ACADEMIA	ALESSIA BASTIANELLI	5	1	NO	IN PERSON	ENGLISH
25	2-3	METHODOLOGY	INTRODUCTION TO META-ANALYSIS WITH	GIANMARCO ALTOE	5	1	NO	IN PERSON	ENGLISH
26	2-3	METHODOLOGY	DATA SIMULATION IN PSYCHOLOGICAL STUDIES	MASSIMILIANO PASTORE	10	2	NO	IN PERSON	ENGLISH
27	2-3	SOFT SKILLS	PUBLISHING IN HIGH-IMPACT JOURNALS	MARA CADINU	15	3	NO	IN PERSON	ENGLISH
28	2-3	THEMATIC COURSE	PSYCHONEUROENDOCRINOLOGY	JEFF KIESNER	5	1	NO	IN PERSON	ENGLISH
29	2-3	PROGRAMMING	BASES OF LINUX FOR DATA SCIENCE	FRANCESCO VESPIGNANI	5	1	NO	IN PERSON	ENGLISH

# A type of data structure you are already familiar with

here is how I would import it in R ([download here](#)), and display the first few rows:

```
library(readxl)
df = data.frame(read_excel("data/Courses40Cycle.xlsx"))
head(df)
```

	Year	TypeOfCourse	Title					
1	1	METHODOLOGY	CURRENT ISSUES IN STATISTICAL INFERENCE FOR PSYCHOLOGY					
2	1	METHODOLOGY		LINEAR AND MIXED EFFECT MODELS WITH SPSS				
3	1	METHODOLOGY			BASICS OF STATISTICAL INFERENCE WITH R			
4	1	PROGRAMMING				BASICS OF R FOR DATA SCIENCE		
5	1	METHODOLOGY				PSYCHOLOGICAL MEASUREMENT		
6	1	METHODOLOGY				POWER AND DESIGN ANALYSIS		
		Teacher	Hours	ECTS	MandatoryAttendance	DeliveryMethod	Language	
1	MASSIMILIANO PASTORE		10	2		YES	IN PERSON	ENGLISH
2	JEFF KIESNER		15	3		NO	IN PERSON	ENGLISH
3	UMBERTO GRANZIOL		20	4		YES	IN PERSON	ENGLISH
4	ENRICO TOFFALINI		10	2		YES	IN PERSON	ENGLISH
5	LUCA STEFANUTTI		15	3		YES	IN PERSON	ENGLISH
6	GIANMARCO ALTOE		5	1		YES	IN PERSON	ENGLISH

# Dataframes as collections of vectors

In fact, **dataframes** are just collections (lists) of **vectors of different types**, all with the same length. Each column in a dataframe is a vector (a variable):

```
df$Teacher
```

```
[1] "MASSIMILIANO PASTORE"      "JEFF KIESNER"  
[3] "UMBERTO GRANZIOL"        "ENRICO TOFFALINI"  
[5] "LUCA STEFANUTTI"         "GIANMARCO ALTOE"  
[7] "GIANMARCO ALTOE"         "TATIANA MARCI"  
[9] "MICHELE VICOVARO"        "FILIPPO GAMBAROTA"  
[11] "LUCA STEFANUTTI"        "ENRICO TOFFALINI"  
[13] "GIULIA CALIGNANO"       "LUCA BATTAGLINI"  
[15] "SIMONE CUTINI"          "MASSIMILIANO PASTORE"  
[17] "FILIPPO GAMBAROTA"      "TOMMASO FERACO"  
[19] "ENRICO SELLA"           "MARINA OTTAVIA EPIFANIA"  
[21] "CHRISTIAN AGRILLO"      "NICOLA CELLINI"  
[23] "ALESSIA BASTIANELLI"    "GIANMARCO ALTOE"  
[25] "MASSIMILIANO PASTORE"    "MARA CADINU"  
[27] "JEFF KIESNER"           "FRANCESCO VESPIGNANI"
```

```
df$Hours
```

```
[1] 10 15 20 10 15 5 5 5 5 5 10 10 5 5 10 10 15 20 5 15 5 10 5 5 10  
[26] 15 5 5
```

# Basic Functions on Dataframes

To know the names of all variables in a dataframe, use the `names()` function:

```
names(df)
```

```
[1] "Year"           "TypeOfCourse"    "Title"  
[4] "Teacher"        "Hours"          "ECTS"  
[7] "MandatoryAttendance" "DeliveryMethod" "Language"
```

Use the `dim()` function to view the dimensions of a dataframe:

```
# first value is number of rows, second is number of columns (variables)  
dim(df)
```

```
[1] 28 9
```

Alternatively, you can use `nrow()` and `ncol()`:

```
nrow(df) # number of rows
```

```
[1] 28
```

```
ncol(df) # number of columns
```

```
[1] 9
```

# Basic Functions on Dataframes

The `str()` function provides a quick overview of the structure of a dataframe, including its dimensions, variables, their data types, and first few observations:

```
str(df)
```

```
'data.frame': 28 obs. of 9 variables:  
 $ Year : chr "1" "1" "1" "1" ...  
 $ TypeOfCourse : chr "METHODOLOGY" "METHODOLOGY" "METHODOLOGY" "PROGRAMMING"  
 ...  
 $ Title : chr "CURRENT ISSUES IN STATISTICAL INFERENCE FOR PSYCHOLOGY"  
 "LINEAR AND MIXED EFFECT MODELS WITH SPSS" "BASICS OF STATISTICAL INFERENCE WITH R"  
 "BASICS OF R FOR DATA SCIENCE" ...  
 $ Teacher : chr "MASSIMILIANO PASTORE" "JEFF KIESNER" "UMBERTO GRANZIOL"  
 "ENRICO TOFFALINI" ...  
 $ Hours : num 10 15 20 10 15 5 5 5 5 5 ...  
 $ ECTS : num 2 3 4 2 3 1 1 1 1 1 ...  
 $ MandatoryAttendance: chr "YES" "NO" "YES" "YES" ...  
 $ DeliveryMethod : chr "IN PERSON" "IN PERSON" "IN PERSON" "IN PERSON" ...  
 $ Language : chr "ENGLISH" "ENGLISH" "ENGLISH" "ENGLISH" ...
```

# Accessing Elements in Dataframes

The “\$” (dollar) operator is essential to access variables in a dataframe:

The screenshot shows an R console window with three tabs: Console, Terminal, and Background Jobs. The current tab is 'Console'. The command entered is 'df\$', which is highlighted in orange. The output is a list of variables from a dataframe:

<code>Year</code>	[df]
<code>TypeOfCourse</code>	[df]
<code>Title</code>	[df]
<code>Teacher</code>	[df]
<code>Hours</code>	[df]
<code>ECTS</code>	[df]
<code>MandatoryAttendance</code>	[df]
<code>DeliveryMethod</code>	[df]
<code>Language</code>	[df]

The 'Teacher' variable is highlighted with a mouse cursor. The output for 'Teacher' is shown in a brown box:

<character> [28]  
chr [1:28] "MASSIMILIANO PASTORE" "JEFF KIESNER" "UMBERTO  
GRANZIOL" "ENRICO TOFFALINI" ...

Annotations with arrows explain the following:

- Name of the dataframe object followed by "\$"
- Type of the highlighted variable
- Length of the highlighted variable (vector)
- A preview of the content
- Pointing a variable highlights it
- A list of all variables (vectors) composing this dataframe

# Accessing and Working with Elements in Dataframes

As an exercise, let's check whether `ECTS` is actually always `Hours*5`

We may use many different, increasingly sophisticated, strategies:

```
df$Hours / df$ECTS
```

```
[1] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
```

```
(df$Hours / df$ECTS) == 5
```

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

```
[16] TRUE TRUE
```

```
sum((df$Hours / df$ECTS) == 5)
```

```
[1] 28
```

```
sum((df$Hours / df$ECTS) == 5) == nrow(df)
```

```
[1] TRUE
```

```
sum((df$Hours / df$ECTS) != 5)
```

```
[1] 0
```

# Accessing and Working with Elements in Dataframes

Variables in a dataframe can be manipulated like any other vector:

```
log(df$Hours)
```

```
[1] 2.302585 2.708050 2.995732 2.302585 2.708050 1.609438 1.609438 1.609438  
[9] 1.609438 1.609438 2.302585 2.302585 1.609438 1.609438 2.302585 2.302585  
[17] 2.708050 2.995732 1.609438 2.708050 1.609438 2.302585 1.609438 1.609438  
[25] 2.302585 2.708050 1.609438 1.609438
```

Also, new variables can easily be created and added at any time:

```
df$newVar = log(df$Hours)  
names(df)
```

```
[1] "Year"                  "TypeOfCourse"        "Title"  
[4] "Teacher"               "Hours"                 "ECTS"  
[7] "MandatoryAttendance" "DeliveryMethod"    "Language"  
[10] "newVar"
```

```
df$newVar
```

```
[1] 2.302585 2.708050 2.995732 2.302585 2.708050 1.609438 1.609438 1.609438  
[9] 1.609438 1.609438 2.302585 2.302585 1.609438 1.609438 2.302585 2.302585  
[17] 2.708050 2.995732 1.609438 2.708050 1.609438 2.302585 1.609438 1.609438  
[25] 2.302585 2.708050 1.609438 1.609438
```

# Indexing Elements in a Dataframe

In addition to using the “\$” (dollar) operator, you can directly access a variable of a dataframe using **indexing with square brackets [ ]**:

```
df[ , "Hours"]
```

```
[1] 10 15 20 10 15 5 5 5 5 10 10 5 5 10 10 15 20 5 15 5 10 5 5 10  
[26] 15 5 5
```

Notice the comma “,” above.

Unlike vectors, dataframes must be indexed by **both row and column**. In the example above, we’re specifying only the desired column (“**Hours**”), leaving the row index blank before the comma **,** This selects all rows for the column named “**Hours**”. Remember that blank index means “all”. Importantly, the “**,**” must always be there when indexing dataframes!

# Indexing Elements in a Dataframe - Examples

```
df[ 1 , "Hours"]
```

```
[1] 10
```

```
df[ 1:5 , "Hours"]
```

```
[1] 10 15 20 10 15
```

```
df[ 1 , c("Teacher","Hours","TypeOfCourse")]
```

	Teacher	Hours	TypeOfCourse
1	MASSIMILIANO PASTORE	10	METHODOLOGY

```
df[ 1:5 , c("Teacher","Hours","TypeOfCourse")]
```

	Teacher	Hours	TypeOfCourse
1	MASSIMILIANO PASTORE	10	METHODOLOGY
2	JEFF KIESNER	15	METHODOLOGY
3	UMBERTO GRANZIOL	20	METHODOLOGY
4	ENRICO TOFFALINI	10	PROGRAMMING
5	LUCA STEFANUTTI	15	METHODOLOGY

```
df[ 1 , c(4, 5, 2)]
```

	Teacher	Hours	TypeOfCourse
1	MASSIMILIANO PASTORE	10	METHODOLOGY

# Indexing and Modifying Elements in a Dataframe

Editing/modifying elements in a dataframe is similar to what you do in vectors

```
dx = data.frame(name = letters[1:10],  
                 score = rnorm(10))  
dx
```

	name	score
1	a	1.262954285
2	b	-0.326233361
3	c	1.329799263
4	d	1.272429321
5	e	0.414641434
6	f	-1.539950042
7	g	-0.928567035
8	h	-0.294720447
9	i	-0.005767173
10	j	2.404653389

```
# Replace some elements with 'NA'  
dx[3:5, "score"] = NA  
dx
```

	name	score
1	a	1.262954285
2	b	-0.326233361
3	c	NA
4	d	NA
5	e	NA
6	f	-1.539950042
7	g	-0.928567035
8	h	-0.294720447
9	i	-0.005767173
10	j	2.404653389

# Indexing and Modifying Elements in a Dataframe

Editing/modifying elements in a dataframe is similar to what you do in vectors

```
dx = data.frame(name = letters[1:10],  
                 score = rnorm(10))  
dx
```

	name	score
1	a	1.262954285
2	b	-0.326233361
3	c	1.329799263
4	d	1.272429321
5	e	0.414641434
6	f	-1.539950042
7	g	-0.928567035
8	h	-0.294720447
9	i	-0.005767173
10	j	2.404653389

```
# Multiply all elements times 10  
dx$score = dx$score * 10  
dx
```

	name	score
1	a	12.62954285
2	b	-3.26233361
3	c	13.29799263
4	d	12.72429321
5	e	4.14641434
6	f	-15.39950042
7	g	-9.28567035
8	h	-2.94720447
9	i	-0.05767173
10	j	24.04653389

# Indexing and Modifying Elements in a Dataframe

Editing/modifying elements in a dataframe is similar to what you do in vectors

```
dx = data.frame(name = letters[1:10],  
                 score = rnorm(10))  
dx
```

	name	score
1	a	1.262954285
2	b	-0.326233361
3	c	1.329799263
4	d	1.272429321
5	e	0.414641434
6	f	-1.539950042
7	g	-0.928567035
8	h	-0.294720447
9	i	-0.005767173
10	j	2.404653389

```
# Linearly transform all elements  
dx$score = round( 100 + dx$score * 15 )  
dx
```

	name	score
1	a	119
2	b	95
3	c	120
4	d	119
5	e	106
6	f	77
7	g	86
8	h	96
9	i	100
10	j	136

# Indexing and Modifying Elements in a Dataframe

Assigning a single value to a column replaces the value for all rows (this is unlike vectors)

```
dx = data.frame(name = letters[1:10],  
                 score = rnorm(10))  
dx
```

	name	score
1	a	1.262954285
2	b	-0.326233361
3	c	1.329799263
4	d	1.272429321
5	e	0.414641434
6	f	-1.539950042
7	g	-0.928567035
8	h	-0.294720447
9	i	-0.005767173
10	j	2.404653389

```
# Replace all elements with a constant  
dx$score = 3  
dx
```

	name	score
1	a	3
2	b	3
3	c	3
4	d	3
5	e	3
6	f	3
7	g	3
8	h	3
9	i	3
10	j	3

# Indexing Dataframes by Logical Conditions

Just like for vectors, you can use **logical conditions** for indexing a dataframe.

How do you extract **only the rows** related to "MASSIMILIANO PASTORE"?

```
head(df)
```

	Year	TypeOfCourse	Title				
1	1	METHODOLOGY	CURRENT ISSUES IN STATISTICAL INFERENCE FOR PSYCHOLOGY				
2	1	METHODOLOGY		LINEAR AND MIXED EFFECT MODELS WITH SPSS			
3	1	METHODOLOGY			BASICS OF STATISTICAL INFERENCE WITH R		
4	1	PROGRAMMING				BASICS OF R FOR DATA SCIENCE	
5	1	METHODOLOGY				PSYCHOLOGICAL MEASUREMENT	
6	1	METHODOLOGY				POWER AND DESIGN ANALYSIS	
		Teacher	Hours	ECTS	MandatoryAttendance	DeliveryMethod	Language
1	MASSIMILIANO PASTORE		10	2	YES	IN PERSON	ENGLISH
2	JEFF KIESNER		15	3	NO	IN PERSON	ENGLISH
3	UMBERTO GRANZIOL		20	4	YES	IN PERSON	ENGLISH
4	ENRICO TOFFALINI		10	2	YES	IN PERSON	ENGLISH
5	LUCA STEFANUTTI		15	3	YES	IN PERSON	ENGLISH
6	GIANMARCO ALTOE		5	1	YES	IN PERSON	ENGLISH

# Indexing Dataframes by Logical Conditions

Just like for vectors, you can use **logical conditions** for indexing a dataframe.

Let's consider this logical condition:

```
df$Teacher == "MASSIMILIANO PASTORE"
```

```
[1] TRUE FALSE  
[13] FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE  
[25] TRUE FALSE FALSE FALSE
```

Let's use it to extract some dataframe rows:

```
df[df$Teacher == "MASSIMILIANO PASTORE", ] # some rows by condition , all columns
```

	Year	TypeOfCourse	Title			
1	1	METHODOLOGY	CURRENT ISSUES IN STATISTICAL INFERENCE FOR PSYCHOLOGY			
16	2-3	METHODOLOGY	BAYESIAN DATA ANALYSIS IN PSYCHOLOGICAL RESEARCH			
25	2-3	METHODOLOGY	DATA SIMULATION IN PSYCHOLOGICAL STUDIES			
	Teacher	Hours	ECTS	MandatoryAttendance	DeliveryMethod	Language
1	MASSIMILIANO PASTORE	10	2	YES	IN PERSON	ENGLISH
16	MASSIMILIANO PASTORE	10	2	NO	IN PERSON	ENGLISH
25	MASSIMILIANO PASTORE	10	2	NO	IN PERSON	ENGLISH
	newVar					
1	2.302585					

16 2.302585

25 2.302585

# Indexing Elements in a Dataframe - A Summary

	[ , "Type" ]	[ , "Teacher" ]	[ , "Hours" ]
[1 , ]	"METHODOLOGY"	"MASSIMILIANO PASTORE"	10
[2 , ]	"METHODOLOGY"	"JEFF KIESNER"	15
[3 , ]	"METHODOLOGY"	"UMBERTO GRANZIOL"	20
[4 , ]	"PROGRAMMING"	"ENRICO TOFFALINI"	10
• • •		• • •	• • •
[28 , ]	"PROGRAMMING"	"FRANCESCO VESPIGNANI"	5

# Subset

Base function `subset()` can also be used as an alternative to indexing

```
subset(df, Teacher == "MASSIMILIANO PASTORE", select=c("Teacher", "Hours", "TypeOfCourse"))
```

	Teacher	Hours	TypeOfCourse
1	MASSIMILIANO PASTORE	10	METHODOLOGY
16	MASSIMILIANO PASTORE	10	METHODOLOGY
25	MASSIMILIANO PASTORE	10	METHODOLOGY

```
df[df$Teacher == "MASSIMILIANO PASTORE", c("Teacher", "Hours", "TypeOfCourse")]
```

	Teacher	Hours	TypeOfCourse
1	MASSIMILIANO PASTORE	10	METHODOLOGY
16	MASSIMILIANO PASTORE	10	METHODOLOGY
25	MASSIMILIANO PASTORE	10	METHODOLOGY

However, indexing with `[]` is more “computationally focused”, computationally faster (especially if working with large datasets), and more similar to programming in other languages (e.g., Python), so should probably be favoured by data scientists!

# Combine Two Dataframes Using `rbind()`

Imagine you have two datasets collected by two students, each including different participants:

`df1`

```
subjName age accuracy
1 Julie   12    0.92
2 Tommy   10    0.78
3 Phil    10    0.85
```

`df2`

```
subjName age accuracy
1 Amber   9     0.87
2 Max     13    0.90
```

Our goal is to get one single dataset including all participant's data for the final analysis. Of course, you could combined these files manually outside R (e.g., in Excel). However, it's simpler and more efficient to do this directly in R using `rbind()`

# Combine Two Dataframes Using `rbind()`

```
dfTotal = rbind(df1, df2)
```

```
dfTotal
```

	subjName	age	accuracy
1	Julie	12	0.92
2	Tommy	10	0.78
3	Phil	10	0.85
4	Amber	9	0.87
5	Max	13	0.90

**Important:** for `rbind()` to work, the two to-be-combined dataframes must:

- have the **exact same number of columns**;
- the **column names must be identical** (remember that R is case-sensitive).

# Merge Two Dataframes Using `merge()`

Another frequent case is having data collected from the **same participants** across **different dataframes**, and having to analyze all information together:

`df1`

```
subjName age
1    Julie  12
2   Amber   9
3  Tommy  10
4    Phil  10
```

`df2`

```
subjName accuracy time
1    Julie      0.92 1203
2   Tommy      0.78 3302
3    Phil      0.85  994
4   Amber      0.87 1163
```

# Merge Two Dataframes Using `merge()`

You can merge the two dataframes into a single, comprehensive dataframe:

```
dfTotal = merge(df1, df2, by="subjName")
```

```
dfTotal
```

	subjName	age	accuracy	time
1	Amber	9	0.87	1163
2	Julie	12	0.92	1203
3	Phil	10	0.85	994
4	Tommy	10	0.78	3302

**Important:** `merge()` will work even if some or even all values that should be used for merging do not match... but in that case part of or all data will be lost

# Contingency Tables

The `table()` function, which counts frequencies, can also be used on dataframes. Importantly, it can also create **contingency tables** when applied to multiple variables at once

```
table(df$Hours) # just counts frequencies
```

```
5 10 15 20  
13 8 5 2
```

```
table(df>TypeOfCourse, df$Hours) # creates contingency table
```

	5	10	15	20
METHODOLOGY	6	4	4	2
PROGRAMMING	1	3	0	0
SOFT SKILLS	4	1	1	0
THEMATIC COURSE	2	0	0	0