# Criação básica de DataFrames

	Α	В	Fixed
1	1	5	1
2	2	6	1
3	3	7	1

```
begin
DataFrame(A=1:3,B=5:7,Fixed = 1)
end
```

```
A B

1 1 "M"

2 2 "F"

3 3 "F"
```

```
begin
data_frame = DataFrame(A = Int[], B = String[])
push!(data_frame, (1, "M"))
push!(data_frame, [2, "F"])
push!(data_frame, Dict(:B => "F", :A => 3))
end
```

ra	no	IOM	ta	h	=

	a	b	С
1	0.41124	0.434895	0.675803
2	0.551547	0.110778	0.558085
3	0.0387148	0.175463	0.865529
4	0.00766722	0.55863	0.437181
5	0.799352	0.868233	0.933774
6	0.870858	0.363279	0.819898
7	0.884659	0.880002	0.72473
8	0.476225	0.879316	0.531207
9	0.942984	0.215322	0.0769115
10	0.265046	0.696807	0.154383

```
random_tab = DataFrame(rand(10, 3), [:a, :b, :c])
```

	customer age	first name
1	15	"Rohit"
2	20	"Rahul"
3	25	"Akshat"

```
• DataFrame("customer age" => [15, 20, 25],
• "first name" =>["Rohit", "Rahul", "Akshat"])
```

	customer age	first name
1	15	"Rohit"
2	20	"Rahul"
3	25	"Akshat"

	a	b
1	1	3
2	2	4

- #DataFrames a partir de vetor de tupleDataFrame((a=[1, 2], b=[3, 4]))

	a	b
1	1	0
2	2	0

```
• #DataFrame a partir de vetores
```

0	<pre>DataFrame( </pre>	[(a=1,	<b>b</b> =0), (	(a=2, b=0)	)]	)
---	------------------------	--------	-----------------	------------	----	---

	Х1	Х2
1	1	0
2	2	0

```
• #DataFrame a partir de MATRIZES
```

<b>DataFrame</b>	<b>(Γ1</b>	0:	2	01.	:auto)

	a	b	С	d
1	1	2	4	5
2	15	58	69	41
3	23	21	26	69

```
• begin
        data = [1 2 4 5; 15 58 69 41; 23 21 26 69]
nomes = ["a", "b", "c", "d"]
        DataFrame(data, nomes)
end
```

## Leitura de CSV

	id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Du
1	0	67	"male"	2	"own"	"NA"	"little"	1169	6
2	1	22	"female"	2	"own"	"little"	"moderate"	5951	48
3	2	49	"male"	1	"own"	"little"	"NA"	2096	12
4	3	45	"male"	2	"free"	"little"	"little"	7882	42
5	4	53	"male"	2	"free"	"little"	"little"	4870	24
6	5	35	"male"	1	"free"	"NA"	"NA"	9055	36
7	6	53	"male"	2	"own"	"quite rich"	"NA"	2835	24
8	7	35	"male"	3	"rent"	"little"	"moderate"	6948	36
9	8	61	"male"	1	"own"	"rich"	"NA"	3059	12
10	9	28	"male"	3	"own"	"little"	"moderate"	5234	30
mo	re								
1000	999	27	"male"	2	"own"	"moderate"	"moderate"	4576	45

```
begin
    german_ref =CSV.read(joinpath(dirname(pathof(DataFrames)),
                                            '..", "docs", "srć", "assets",
"german.csv"),
                                 DataFrame)
    #obs joinpath(dirname(pathof(DataFrames)),"..", "docs", "src", "assets",
"german.csv")
    #é equivalente a:
#C:\\Users\\gabri\\.julia\\packages\\DataFrames\\vuMM8\\src\\..\\docs\\src\\assets\\
german.csv"
    german = copy(german_ref)
    #obs: também é possivel
    german_ref2 = DataFrame(CSV.File(
"C:\\Users\\gabri\\.julia\\packages\\DataFrames\\vuMM8\\src\\..\\docs\\src\\assets\\
german.csv"))
    #leitura excel
    #df = DataFrame(XLSX.readtable("myfile.xlsx", "mysheet")...)
end
```

# **Exportar CSV**

# Vizualização

	id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Du
1	0	67	"male"	2	"own"	"NA"	"little"	1169	6
2	1	22	"female"	2	"own"	"little"	"moderate"	5951	48
3	2	49	"male"	1	"own"	"little"	"NA"	2096	12
4	3	45	"male"	2	"free"	"little"	"little"	7882	42
5	4	53	"male"	2	"free"	"little"	"little"	4870	24
6	5	35	"male"	1	"free"	"NA"	"NA"	9055	36
7	6	53	"male"	2	"own"	"quite rich"	"NA"	2835	24
8	7	35	"male"	3	"rent"	"little"	"moderate"	6948	36
9	8	61	"male"	1	"own"	"rich"	"NA"	3059	12
10	9	28	"male"	3	"own"	"little"	"moderate"	5234	30
moı	re								
1000	999	27	"male"	2	"own"	"moderate"	"moderate"	4576	45

```
begin
first(german,5) #mostra as primeiras 5 linhas do DataFrame
last(german, 6) #mostra as ultimas 6 linhas do DataFrame

### view(DataFrame, array_linhas, array_colunas)
view(german, 2, 2) #mostra uma unica célula
view(german, : , :) #mostra as linhas e colunas de algum DataFrame
end
```

	Sex	Saving accounts
1	"male"	"NA"
2	"female"	"little"
3	"male"	"little"
4	"male"	"little"
5	"male"	"NA"
6	"male"	"quite rich"
7	"male"	"little"
8	"male"	"rich"
9	"male"	"little"
10	"female"	"little"
mo	ore	
999	"male"	"moderate"

```
begin
    #usando betewwn, cols, not
    german[:, Not(:Age)] #retorna todo a DataFrame menos a coluna :Age
    german[:, Between(:Sex, :Housing)] #Retorna as colunas Sex, Job, Housing
    german[:,Cols("Age",Between("Sex","Job"))]#Retorna as colunas Age,Sex,Job,
    Housing
    german[Not(5), r"S"] #retorna colunas com a letra S menos a linhas 5
    end

["Sex", "Housing", "Saving accounts", "Checking account", "Purpose"]

begin
    names(german) #retorna uma lista de nomes das colunas
    names(german, String) #retorna uma lista de nomes das Colunas do tipo String
    end

[:id, :Age, :Sex, :Job, :Housing, Symbol("Saving accounts"), Symbol("Checking account"),

propertynames(german) #retorna uma lista de simbolos de l=nomes das colunas

[Int64, Int64, String, Int64, String, String, String, Int64, Int64, String]
    eltype.(eachcol(german)) #retorna uma lista de tipos de cada coluna
```

### empty x empty!

id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Duration	Purpo
----	-----	-----	-----	---------	-----------------	------------------	---------------	----------	-------

```
    begin
    a = empty(german) #cria um novo DataFrame com o formato de german, mas vazio
    #b = empty!(german) #exclui as linhas de german
    end
```

### Infos Basicas do DataFrame

```
10
```

```
begin
size(german) #(1000,100)
size(german,1) #1000
size(german,2) #10
nrow(german) #1000
ncol(german) #10
end
```

	variable	mean	min	median	max	nmis
1	:id	499.5	0	499.5	999	0
2	:Age	35.546	19	33.0	75	0
3	:Sex	nothing	"female"	nothing	"male"	0
4	:Job	1.904	0	2.0	3	0
5	:Housing	nothing	"free"	nothing	"rent"	0
6	Symbol("Saving accounts")	nothing	"NA"	nothing	"rich"	0
7	<pre>Symbol("Checking account")</pre>	nothing	"NA"	nothing	"rich"	0
8	<pre>Symbol("Credit amount")</pre>	3271.26	250	2319.5	18424	0
9	:Duration	20.903	4	18.0	72	0
10	:Purpose	nothing	"business"	nothing	"vacation/others"	0

```
describe(german)
```

	variable	mean	min	median	max	nmissing	eltype
1	:id	499.5	0	499.5	999	0	Int64
2	:Age	35.546	19	33.0	75	0	Int64
3	:Sex	nothing	"female"	nothing	"male"	0	String

- describe(german, cols=1:3) # define quais colunas são analisadas
- #função show(german, allrows=True) -> imprimirá todas as linhas

### Análises de dados

#### 35.546

- begin
- mean(skipmissing(german."Age"))
- end

	id	Age	Sex	Job	Housing	Saving accounts	Che
1	0	4489	"malemale"	4	"ownown"	"NANA"	"litt
2	1	484	"femalefemale"	4	"ownown"	"littlelittle"	"mode
3	4	2401	"malemale"	1	"ownown"	"littlelittle"	"NANA
4	9	2025	"malemale"	4	"freefree"	"littlelittle"	"litt
5	16	2809	"malemale"	4	"freefree"	"littlelittle"	"litt
6	25	1225	"malemale"	1	"freefree"	"NANA"	"NANA
7	36	2809	"malemale"	4	"ownown"	"quite richquite rich"	"NANA
8	49	1225	"malemale"	9	"rentrent"	"littlelittle"	"mode
9	64	3721	"malemale"	1	"ownown"	"richrich"	"NANA
10	81	784	"malemale"	9	"ownown"	"littlelittle"	"mode
mon	re						
1000	998001	729	"malemale"	4	"ownown"	"moderatemoderate"	"mode

- mapcols(id -> id .^ 2, german)
- #realia uma função a todas coluna. Deve passar a coluna como argumento da função
- #não altera a função original german

### Renomear Colunas

md" ### Renomear Colunas"

	id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Tin
1	0	67	"male"	2	"own"	"NA"	"little"	1169	6
2	1	22	"female"	2	"own"	"little"	"moderate"	5951	48
3	2	49	"male"	1	"own"	"little"	"NA"	2096	12
4	3	45	"male"	2	"free"	"little"	"little"	7882	42
5	4	53	"male"	2	"free"	"little"	"little"	4870	24
6	5	35	"male"	1	"free"	"NA"	"NA"	9055	36
7	6	53	"male"	2	"own"	"quite rich"	"NA"	2835	24
8	7	35	"male"	3	"rent"	"little"	"moderate"	6948	36
9	8	61	"male"	1	"own"	"rich"	"NA"	3059	12
10	9	28	"male"	3	"own"	"little"	"moderate"	5234	30
mo	re								
1000	999	27	"male"	2	"own"	"moderate"	"moderate"	4576	45

```
rename!(german, "Duration" => "Time")
```

### Acessando coluna

PooledArrays.PooledVector{String, UInt32, Vector{UInt32}}: ["male", "female", "male", "i

```
begin

#acessando a coluna Sex do DataFrame

german.Sex

german."Sex"

german[!,"Sex"]

col_name = "Sex"

german[!,col_name]
end
```

PooledArrays.PooledVector{String, UInt32, Vector{UInt32}}: ["male", "female", "male", "i

```
    begin
    #acessando a coluna Sex do DataFrame
    german[:,col_name]
    #dessa forma é criada uma cópia da coluna que não irá afetar a coluna original
    end
```

#### false

```
begin
german.Sex === german[!, :Sex] #True
german.Sex === german[:, :Sex] #False
end
```

### Acessando Index

### data\_frame[selected\_rows, selected\_columns]

data\_frame[array , array] -> Retorna um DataFrame

data\_frame[array , Symbol/Strin] -> Retorna um Array

pode-se usar data\_frame[condicao,coluna]

É sempre necessário passar linhas e colunas. Operado4: [:] indica todos os elementos da linha ou coluna

#### select(DataFrame, colunas)

tem a vantagem de poder ter vários métodos de seleção de colunas

ex. select(DF, 1, :Sex, r'a')

	id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Tin
1	0	67	"male"	2	"own"	"NA"	"little"	1169	6
2	1	22	"female"	2	"own"	"little"	"moderate"	5951	48
3	2	49	"male"	1	"own"	"little"	"NA"	2096	12
4	3	45	"male"	2	"free"	"little"	"little"	7882	42
5	4	53	"male"	2	"free"	"little"	"little"	4870	24
6	5	35	"male"	1	"free"	"NA"	"NA"	9055	36
7	6	53	"male"	2	"own"	"quite rich"	"NA"	2835	24
8	7	35	"male"	3	"rent"	"little"	"moderate"	6948	36
9	8	61	"male"	1	"own"	"rich"	"NA"	3059	12
10	9	28	"male"	3	"own"	"little"	"moderate"	5234	30
mo	re								
1000	999	27	"male"	2	"own"	"moderate"	"moderate"	4576	45

```
    begin
    german[1:5,1:2] #indicies de linha e coluna
    german[1:5, [:Sex , :Age]] #indices de linha e nomes por Symbol das colunas
    german[1:5,:] #apenas as 5 primeiras linhas, mas todas as colunas
    german[[1,6,15] , :]
    german[! , :] # Retorna todas as linhas e colunas do próprio DataFrame (!)
    end
```

PooledArrays.PooledVector{String, UInt32, Vector{UInt32}}: ["male", "female", "male", "i

```
    begin
    german[:, [:Sex]] #retorna um DataFrame
    german[:, :Sex] #retorna um Array
    end
```

	Job	Housing	Saving accounts	Checking account	Credit amount	Time	Purpose
1	2	"own"	"NA"	"little"	1169	6	"radio/TV"
2	2	"own"	"little"	"moderate"	5951	48	"radio/TV"
3	1	"own"	"little"	"NA"	2096	12	"education"
4	2	"free"	"little"	"little"	7882	42	"furniture/equipment'
5	2	"free"	"little"	"little"	4870	24	"car"
6	1	"free"	"NA"	"NA"	9055	36	"education"
7	2	"own"	"quite rich"	"NA"	2835	24	"furniture/equipment'
8	3	"rent"	"little"	"moderate"	6948	36	"car"
9	1	"own"	"rich"	"NA"	3059	12	"radio/TV"
10	3	"own"	"little"	"moderate"	5234	30	"car"
mo	re						
1000	2	"own"	"moderate"	"moderate"	4576	45	"car"

```
begin
select(german, r"S")
select(german, Not(["Sex","id","Age"]))
end
```

# Acesso condicional

data\_frame[condicao,coluna]

	id	Age	Sex	Job	Housing	Saving accounts	Checking account	Credit amount	Time
1	0	67	"male"	2	"own"	"NA"	"little"	1169	6
2	2	49	"male"	1	"own"	"little"	"NA"	2096	12
3	3	45	"male"	2	"free"	"little"	"little"	7882	42
4	4	53	"male"	2	"free"	"little"	"little"	4870	24
5	5	35	"male"	1	"free"	"NA"	"NA"	9055	36
6	6	53	"male"	2	"own"	"quite rich"	"NA"	2835	24
7	7	35	"male"	3	"rent"	"little"	"moderate"	6948	36
8	8	61	"male"	1	"own"	"rich"	"NA"	3059	12
9	13	60	"male"	1	"own"	"little"	"little"	1199	24
10	16	53	"male"	2	"own"	"NA"	"NA"	2424	24
m	ore								
461	997	38	"male"	2	"own"	"little"	"NA"	804	12

```
begin
    ### ACESSO CONDICIONAL ###

#Retorna apenas as linhas com Age> que 30
german[german."Age" .> 30,:] #necessário o .> para realizar comparação célula a celula
german[(german."Age" .> 30) .& (german."Sex" .== "male"),:]
end
```

# Manipulação de dados

df1 =

	Age	Sex	Job
1	67	"male"	2
2	22	"female"	2
3	49	"male"	1
4	45	"male"	2
5	53	"male"	2
6	35	"male"	1

```
- df1 = german[1:6, 2:4]
```

	Age	Sex	Job
1	80	"male"	2
2	85	"female"	2
3	98	"male"	1
4	95	"male"	2
5	78	"male"	2
6	89	"male"	1

```
    begin
    val = [80, 85, 98, 95, 78, 89]
    df1.Age = val #alterar todos os valores da coluna
    view(df1, :,:)
    end
```

	Age	Sex	Job
1	80	"male"	10
2	85	"female"	10
3	98	"male"	10
4	95	"male"	2
5	78	"male"	2
6	89	"male"	1

```
• begin
• df1[1:3, :Job] = [10, 10, 10] #muda apenas os valores acessados
• view(df1,:,:)
• end
```

	Age	Sex	Job
1	80	"male"	10
2	85	"female"	10
3	78	"male"	4
4	95	"transgender"	2
5	78	"female"	2
6	89	"male"	1

```
begin
df1[!, :Sex] = ["male", "female", "transgender", "female", "male"]
df1[3, 1:3] = [78, "male", 4] #muda apenas os valores acessados
view(df1,:,:)
end
```

#### DataFrameRow (3 columns)

	Age	Sex	Job
	Int64	String	Int64
2	100	male	2

```
begin
df2 = df1[2, :] #não cria uma cópia de df1. Toda alteração em df2, altera df1
df2.Age = 100
df2[2:3] = ["male", 2]
df2
end
```

	Age	Sex	Job	Customers	City
1	80	"male"	4	"Rohit"	"Kanpur"
2	100	"male"	4	"Akshat"	"Lucknow"
3	78	"male"	4	"Rahul"	"Bhuvneshwar"
4	95	"transgender"	4	"Aayush"	"Jaipur"
5	78	"female"	4	"Prateek"	"Ranchi"
6	89	"male"	4	"Anam"	"Dehradoon"

```
begin
df1[!, :Customers] = ["Rohit", "Akshat", "Rahul", "Aayush", "Prateek", "Anam"]
df1[:, :City] = ["Kanpur", "Lucknow", "Bhuvneshwar", "Jaipur", "Ranchi",
"Dehradoon"]
df1[:, 3] .= 4 #precisa do .= para colocar o valor em todo o array
view(df1,:,:)
end
```

	Age	Sex	Job	Customers	City
1	"Economics"	"male"	4	"Rohit"	"Kanpur"
2	"Economics"	"male"	4	"Akshat"	"Lucknow"
3	"Economics"	"male"	4	"Rahul"	"Bhuvneshwar"
4	"Economics"	"transgender"	4	"Aayush"	"Jaipur"
5	"Economics"	"female"	4	"Prateek"	"Ranchi"
6	"Economics"	"male"	4	"Anam"	"Dehradoon"

```
begin
  #df1[:, :Age] .= "Economics"
  #ERROR: não pode converter a cópia de df1 em Economics

df1[!, :Age] .= "Economics" #não dá erro
  view(df1,:,:)
end
```

# Inserção de colunas

insertcols!( DataFrame, index\_da\_coluna, Symbol\_Name => valor)

	Country	Age	Sex	Job	Customers	City
1	"India"	"Economics"	"male"	4	"Rohit"	"Kanpur"
2	"India"	"Economics"	"male"	4	"Akshat"	"Lucknow"
3	"India"	"Economics"	"male"	4	"Rahul"	"Bhuvneshwar"
4	"India"	"Economics"	"transgender"	4	"Aayush"	"Jaipur"
5	"India"	"Economics"	"female"	4	"Prateek"	"Ranchi"
6	"India"	"Economics"	"male"	4	"Anam"	"Dehradoon"

```
• insertcols!(df1, 1, :Country => "India")
```

# Transformações

tranformacao( DataFrameSource, coluna => transformations => coluna\_alvo\_nome)

### tranformações:

- combine -> cria um novo DataFrame populado com a transformação
- select -> cria um novo DataFrame com o mesmo número de linhas do Source, e populado com a transformação
- select! -> altera a Source
- transform -> cria um novo DataFrame com o mesmo número de Linhas e Colunas do Source, e popupado com a transformação
- transform! -> altera o Source

#### transformations:

- mean
- unique
- uppercase
- sqrt
- exp
- sin

mean\_age

**1** 35.546

combine(german, :Age => mean => :mean\_age)

	mean_age
1	35.546
2	35.546
3	35.546
4	35.546
5	35.546
6	35.546
7	35.546
8	35.546
9	35.546
10	35.546
mor	^e
1000	35.546

select(german, :Age => mean => :mean\_age)

	mean_age	housing
1	35.546	"own"
2	35.546	"free"
3	35.546	"rent"

```
combine(german, :Age => mean => :mean_age, :Housing => unique => :housing)
```

```
Sex
                  Sex_UpperCase
       "male"
                  "MALE"
 1
       "female"
                  "FEMALE"
 2
       "male"
                  "MALE"
 3
       "male"
                  "MALE"
 4
       "male"
                  "MALE"
 5
       "male"
                  "MALE"
 6
       "male"
                  "MALE"
 7
       "male"
                  "MALE"
 8
       "male"
                  "MALE"
 9
       "male"
                  "MALE"
 10
  more
       "male"
                  "MALE"
1000
```

```
begin
    #necessita da função lambda quando acessa os valores da coluna 1 a 1
select(german, :Sex => (x -> uppercase.(x)) => :Sex)

#ByRow realiza a transformação célula a celula
select(german, :Sex ,:Sex => ByRow(uppercase) => :Sex_UpperCase)
end
```

	X1	Х2
1	"male"	67
2	"female"	22
3	"male"	49
4	"male"	45
5	"male"	53
6	"male"	35
7	"male"	53
8	"male"	35
9	"male"	61
<b>10</b> "male"		28
moı	^e	
1000	"male"	27

 select(german, :Sex => :x1, :Age => :x2) #apenas crria um dataset igual com nomes diferentes

	Age	Job	res
1	67	2	69
2	22	2	24
3	49	1	50
4	45	2	47
5	53	2	55
6	35	1	36
7	53	2	55
8	35	3	38
9	61	1	62
10	28	3	31
more			
1000	27	2	29

```
select(german, :Age, :Job, [:Age, :Job] => (+) => :res) #soma de coluna
```

	id	Age	Sex	Job	Housing
1	0	67	"male"	2	"own"
2	1	67	"female"	2	"own"
3	2	67	"male"	1	"own"
4	3	67	"male"	2	"free"
5	4	67	"male"	2	"free"
6	5	67	"male"	1	"free"
7	6	67	"male"	2	"own"
8	7	67	"male"	3	"rent"

```
begin
df = german_ref[1:8, 1:5]
transform(df, :Age => maximum => "Age")
end
```

	id	Age	Sex	Job	Housing
1	0	"male"	67	2	"own"
2	1	"female"	22	2	"own"
3	2	"male"	49	1	"own"
4	3	"male"	45	2	"free"
5	4	"male"	53	2	"free"
6	5	"male"	35	1	"free"
7	6	"male"	53	2	"own"
8	7	"male"	35	3	"rent"

• transform(df, :Age => :Sex, :Sex => :Age)

	a	b	c
1	0.41124	0.434895	0.675803
2	0.551547	0.110778	0.558085
3	0.0387148	0.175463	0.865529
4	0.00766722	0.55863	0.437181
5	0.799352	0.868233	0.933774
6	0.870858	0.363279	0.819898
7	0.884659	0.880002	0.72473
8	0.476225	0.879316	0.531207
9	0.942984	0.215322	0.0769115
10	0.265046	0.696807	0.154383

random\_tab

	a	b	C	prediction
1	0.41124	0.434895	0.675803	:c
2	0.551547	0.110778	0.558085	<b>:</b> C
3	0.0387148	0.175463	0.865529	<b>:</b> C
4	0.00766722	0.55863	0.437181	:b
5	0.799352	0.868233	0.933774	:c
6	0.870858	0.363279	0.819898	<b>:</b> a
7	0.884659	0.880002	0.72473	<b>:</b> a
8	0.476225	0.879316	0.531207	<b>:</b> b
9	0.942984	0.215322	0.0769115	<b>:</b> a
10	0.265046	0.696807	0.154383	<b>:</b> b

```
transform(random_tab, AsTable(:) => ByRow(argmax) => :prediction)
```

# Limpeza de dados

dropmissing(DataFrame, colunas,

	i	Х	у
1	1	missing	missing
2	2	4	missing
3	3	missing	"c"
4	4	2	"d"
5	5	1	"e"

	i	х	у
1	2	4	missing
2	4	2	"d"
3	5	1	"e"

 dropmissing!(data\_missing, ["x"], disallowmissing=true ) #remove linhas com valores missing

# **Juntar DataFrames**

innerjoin(DF1, DF2, on = coluna\_de\_juncao)

- innerjoin : Retona apenas os valores que tem correspondência entre os 2 dataframes
- leftjoin : Retorna todo o Dataframe da esquerda e apenas as correspondências da direita
- rightjoin : Retorna todo o Dataframe da direita e apenas as correspondências da esquerda
- **outerjoin**: Retorna todos os valores dos 2 dataframes, mas preenche o que não corresponder com Missing
- semijoin: Igual innerjoy, mas apenas com os valores do dataframe da esquerda
- antijoin : Retorna apenas os valores do dataframe da esquerda que não tem correspondência com algum dataframe da direita

Caso a correspondencia seja em colunas com nomes diferentes em cada dataframe, usa-se:

innerjoin(a, b, on = coluna\_a => coluna\_b)

innerjoin(a, b, on = :ID => :IDNew)

	ID	Job
1	20	"Lawyer"
2	60	"Doctor"

```
• jobs = DataFrame(ID = [20, 60], Job = ["Lawyer", "Doctor"])
```

🔾 DataFrames.jl — Pluto.jl

people =

ID Name

- **1** 20 "John Doe"
- **2** 40 "Jane Doe"
- people = DataFrame(ID = [20, 40], Name = ["John Doe", "Jane Doe"])

ID Name Job

1 20 "John Doe" "Lawyer"

• innerjoin(people,jobs,on ="ID")

ID Name Job

1 20 "John Doe" "Lawyer"
2 40 "Jane Doe" missing

• leftjoin(people, jobs, on= :ID)

ID Name

**1** 20 "John Doe"

semijoin(people, jobs, on=:ID)

ID Name

**1** 40 "Jane Doe"

antijoin(people, jobs, on=:ID)

Junção por várias colunas:

#### A:

5 rows × 3 columns

	City	Job	Category
	String	String	Int64
1	Amsterdam	Lawyer	1
2	London	Lawyer	2
3	London	Lawyer	3
4	New York	Doctor	4
5	New York	Doctor	5

#### В

5 rows × 3 columns

	Location	Work	Name
	String	String	String
1	Amsterdam	Lawyer	а
2	London	Lawyer	b
3	London	Lawyer	С
4	New York	Doctor	d
5	New York	Doctor	е

```
md"""
Junção por várias colunas:

**A**:
$(A)

**B**
$B
```

	City	Job	Category	Name
1	"Amsterdam"	"Lawyer"	1	"a"
2	"London"	"Lawyer"	2	"b"
3	"London"	"Lawyer"	3	"b"
4	"London"	"Lawyer"	2	"c"
5	"London"	"Lawyer"	3	"c"
6	"New York"	"Doctor"	4	"d"
7	"New York"	"Doctor"	5	"d"
8	"New York"	"Doctor"	4	"e"
9	"New York"	"Doctor"	5	"e"

```
begin
A = DataFrame(City = ["Amsterdam", "London", "London", "New York", "New York"],
Job = ["Lawyer", "Lawyer", "Lawyer", "Doctor", "Doctor"],
Category = [1, 2, 3, 4, 5])
B = DataFrame(Location = ["Amsterdam", "London", "London", "New York", "New York"],
Work = ["Lawyer", "Lawyer", "Lawyer", "Doctor", "Doctor"],
Name = ["a", "b", "c", "d", "e"])
innerjoin(A,B, on= [:City=>:Location, :Job=>:Work])
end
```

# Separação - Transformação - Combinação

Agrupamentos

iris =		SepalLength	SepalWidth	PetalLength	PetalWidth	Species
	1	5.1	3.5	1.4	0.2	"Iris-setosa"
	2	4.9	3.0	1.4	0.2	"Iris-setosa"
	3	4.7	3.2	1.3	0.2	"Iris-setosa"
	4	4.6	3.1	1.5	0.2	"Iris-setosa"
	5	5.0	3.6	1.4	0.2	"Iris-setosa"
	6	5.4	3.9	1.7	0.4	"Iris-setosa"
	7	4.6	3.4	1.4	0.3	"Iris-setosa"
	8	5.0	3.4	1.5	0.2	"Iris-setosa"
	9	4.4	2.9	1.4	0.2	"Iris-setosa"
	10	4.9	3.1	1.5	0.1	"Iris-setosa"
	m	ore				
	150	5.9	3.0	5.1	1.8	"Iris-virginica"

### **GROUPBY**

Separa o dataframe em subgrupos dependendo do agrupamento que fizer. Cada grupo terá um dos valores da coluna especificada.

groupby(DataFrame, coluna)

gdf =
GroupedDataFrame with 3 groups based on key: Species

First Group (50 rows): Species = "Iris-setosa"

	SepalLength Float64	SepalWidth Float64	PetalLength Float64	PetalWidth Float64	Species String
1	5.1	3.5	1.4	0.2	Iris-setosa
2	4.9	3.0	1.4	0.2	Iris-setosa
3	4.7	3.2	1.3	0.2	Iris-setosa
4	4.6	3.1	1.5	0.2	Iris-setosa
5	5.0	3.6	1.4	0.2	Iris-setosa
6	5.4	3.9	1.7	0.4	Iris-setosa
7	4.6	3.4	1.4	0.3	Iris-setosa
8	5.0	3.4	1.5	0.2	Iris-setosa
9	4.4	2.9	1.4	0.2	Iris-setosa
10	4.9	3.1	1.5	0.1	Iris-setosa
11	5.4	3.7	1.5	0.2	Iris-setosa
12	4.8	3.4	1.6	0.2	Iris-setosa
13	4.8	3.0	1.4	0.1	Iris-setosa
14	4.3	3.0	1.1	0.1	Iris-setosa
15	5.8	4.0	1.2	0.2	Iris-setosa
16	5.7	4.4	1.5	0.4	Iris-setosa
17	5.4	3.9	1.3	0.4	Iris-setosa
18	5.1	3.5	1.4	0.3	Iris-setosa
:	:	:	:	:	:

:

Last Group (50 rows): Species = "Iris-virginica"

	SepalLength Float64	SepalWidth Float64	PetalLength Float64	PetalWidth Float64	Species String
1	6.3	3.3	6.0	2.5	Iris-virginica
2	5.8	2.7	5.1	1.9	Iris-virginica
3	7.1	3.0	5.9	2.1	Iris-virginica
4	6.3	2.9	5.6	1.8	Iris-virginica
5	6.5	3.0	5.8	2.2	Iris-virginica
6	7.6	3.0	6.6	2.1	Iris-virginica
7	4.9	2.5	4.5	1.7	Iris-virginica
8	7.3	2.9	6.3	1.8	Iris-virginica
9	6.7	2.5	5.8	1.8	Iris-virginica

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
		•	· ·		•
	Float64	Float64	Float64	Float64	String
10	7.2	3.6	6.1	2.5	Iris-virginica
11	6.5	3.2	5.1	2.0	Iris-virginica
12	6.4	2.7	5.3	1.9	Iris-virginica
13	6.8	3.0	5.5	2.1	Iris-virginica
14	5.7	2.5	5.0	2.0	Iris-virginica
15	5.8	2.8	5.1	2.4	Iris-virginica
16	6.4	3.2	5.3	2.3	Iris-virginica
17	6.5	3.0	5.5	1.8	Iris-virginica
18	7.7	3.8	6.7	2.2	Iris-virginica
÷	:	:	:	:	:

gdf = groupby(iris, :Species)

### GroupedDataFrame with 2 groups based on key: Species

First Group (50 rows): Species = "Iris-setosa"

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
	Float64	Float64	Float64	Float64	String
1	5.1	3.5	1.4	0.2	Iris-setosa
2	4.9	3.0	1.4	0.2	Iris-setosa
3	4.7	3.2	1.3	0.2	Iris-setosa
4	4.6	3.1	1.5	0.2	Iris-setosa
5	5.0	3.6	1.4	0.2	Iris-setosa
6	5.4	3.9	1.7	0.4	Iris-setosa
7	4.6	3.4	1.4	0.3	Iris-setosa
8	5.0	3.4	1.5	0.2	Iris-setosa
9	4.4	2.9	1.4	0.2	Iris-setosa
10	4.9	3.1	1.5	0.1	Iris-setosa
11	5.4	3.7	1.5	0.2	Iris-setosa
12	4.8	3.4	1.6	0.2	Iris-setosa
13	4.8	3.0	1.4	0.1	Iris-setosa
14	4.3	3.0	1.1	0.1	Iris-setosa
15	5.8	4.0	1.2	0.2	Iris-setosa
16	5.7	4.4	1.5	0.4	Iris-setosa
17	5.4	3.9	1.3	0.4	Iris-setosa
18	5.1	3.5	1.4	0.3	Iris-setosa
:	:	:	:	:	:

:

Last Group (50 rows): Species = "Iris-virginica"

	SepalLength Float64	SepalWidth Float64	PetalLength Float64	PetalWidth Float64	Species String
1	6.3	3.3	6.0	2.5	Iris-virginica
2	5.8	2.7	5.1	1.9	Iris-virginica
3	7.1	3.0	5.9	2.1	Iris-virginica
4	6.3	2.9	5.6	1.8	Iris-virginica
5	6.5	3.0	5.8	2.2	Iris-virginica
6	7.6	3.0	6.6	2.1	Iris-virginica
7	4.9	2.5	4.5	1.7	Iris-virginica
8	7.3	2.9	6.3	1.8	Iris-virginica
9	6.7	2.5	5.8	1.8	Iris-virginica

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
	Float64	Float64	Float64	Float64	String
10	7.2	3.6	6.1	2.5	Iris-virginica
11	6.5	3.2	5.1	2.0	Iris-virginica
12	6.4	2.7	5.3	1.9	Iris-virginica
13	6.8	3.0	5.5	2.1	Iris-virginica
14	5.7	2.5	5.0	2.0	Iris-virginica
15	5.8	2.8	5.1	2.4	Iris-virginica
16	6.4	3.2	5.3	2.3	Iris-virginica
17	6.5	3.0	5.5	1.8	Iris-virginica
18	7.7	3.8	6.7	2.2	Iris-virginica
:	:	:	:	:	:

• gdf[ [("Iris-setosa",) , ("Iris-virginica",) ]] #pega apenas 2 dos 3 grupos

# Tranformações em grupos

### Combine

	Species	PetalLength_mean
1	"Iris-setosa"	1.464
2	"Iris-versicolor"	4.26
3	"Iris-virginica"	5.552

combine(gdf, :PetalLength=>mean)

	Species	nrow
1	"Iris-setosa"	50
2	"Iris-versicolor"	50
3	"Iris-virginica"	50

• combine(gdf, nrow) #retorna o número de linhas por grupo

	Species	NumRows	Mean
1	"Iris-setosa"	50	1.464
2	"Iris-versicolor"	50	4.26
3	"Iris-virginica"	50	5.552

```
combine(gdf, nrow=>:NumRows , :PetalLength =>mean => :Mean)
```

	Species	a	b
1	"Iris-setosa"	0.292449	73.2
2	"Iris-versicolor"	0.717655	213.0
3	"Iris-virginica"	0.842744	277.6

```
    combine(gdf, [:PetalLength, :SepalLength] => ((p, s) -> (a=mean(p)/mean(s), b=sum(p))) => AsTable)
    #poderia ser passado para um vetor de colunas também
    #combine(gdf, [:PetalLength, :SepalLength] => ((p, s) -> (a=mean(p)/mean(s), b=sum(p))) => [:ColA, :ColB])
```

	Species	PetalLength_SepalLength_function
1	"Iris-setosa"	0.492245
2	"Iris-versicolor"	0.910378
3	"Iris-virginica"	0.867923

```
#valor AsTable, passa um dataframe como parâmetro
combine(gdf,
AsTable([:PetalLength, :SepalLength]) =>
x -> std(x.PetalLength) / std(x.SepalLength))
```

	Species	min	max
1	"Iris-setosa"	1.0	1.9
2	"Iris-versicolor"	3.0	5.1
3	"Iris-virginica"	4.5	6.9

```
    combine(gdf, :PetalLength => (x -> [extrema(x)]) => [:min, :max])
    #retorna os minimos e máximos na forma tuple para cada grupo
```

	Species	PetalWidth_unique
1	"Iris-setosa"	0.2
2	"Iris-setosa"	0.4
3	"Iris-setosa"	0.3
4	"Iris-setosa"	0.1
5	"Iris-setosa"	0.5
6	"Iris-setosa"	0.6
7	"Iris-versicolor"	1.4
8	"Iris-versicolor"	1.5
9	"Iris-versicolor"	1.3
10	"Iris-versicolor"	1.6
ı	nore	
27	"Iris-virginica"	1.4

combine(gdf, :PetalWidth=>unique)

## Select e Transform

• md" #### Select e Transform"

	Species	SepalLength_SepalWidth_cor
1	"Iris-setosa"	0.74678
2	"Iris-setosa"	0.74678
3	"Iris-setosa"	0.74678
4	"Iris-setosa"	0.74678
5	"Iris-setosa"	0.74678
6	"Iris-setosa"	0.74678
7	"Iris-setosa"	0.74678
8	"Iris-setosa"	0.74678
9	"Iris-setosa"	0.74678
10	"Iris-setosa"	0.74678
m	ore	
150	"Iris-virginica"	0.457228

select(gdf, 1:2 => cor)

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species	Species_func
1	5.1	3.5	1.4	0.2	"Iris-setosa"	"setosa"
2	4.9	3.0	1.4	0.2	"Iris-setosa"	"setosa"
3	4.7	3.2	1.3	0.2	"Iris-setosa"	"setosa"
4	4.6	3.1	1.5	0.2	"Iris-setosa"	"setosa"
5	5.0	3.6	1.4	0.2	"Iris-setosa"	"setosa"
6	5.4	3.9	1.7	0.4	"Iris-setosa"	"setosa"
7	4.6	3.4	1.4	0.3	"Iris-setosa"	"setosa"
8	5.0	3.4	1.5	0.2	"Iris-setosa"	"setosa"
9	4.4	2.9	1.4	0.2	"Iris-setosa"	"setosa"
10	4.9	3.1	1.5	0.1	"Iris-setosa"	"setosa"
m	ore					
150	5.9	3.0	5.1	1.8	"Iris-virginica"	"virginica"

```
    transform(gdf, :Species => x -> chop.(x, head=5, tail=0))
    #função chop(string, head=n, tail = m)
    #remove o primeiros n caracteres da string, e os ultimos m caracteres da string
```

Pode-se selecionar grupos desejados a partir de tuples com o valor de separação dos grupos

### Stack

Tranforma uma tabela curta em tablea longa a partir das colunas selecionada.

Cada linha terá o valor da coluna

```
md" ### Stack
Tranforma uma tabela curta em tablea longa a partir das colunas selecionada.
Cada linha terá o valor da coluna
```

```
• stacked = stack(iris, 1:4) #faz um empilhamento das coluna 1:4 (todas menos :value);
```

#### medias\_stacked =

	Species	variable	media
1	"Iris-setosa"	"SepalLength"	5.006
2	"Iris-setosa"	"SepalWidth"	3.418
3	"Iris-setosa"	"PetalLength"	1.464
4	"Iris-setosa"	"PetalWidth"	0.244
5	"Iris-versicolor"	"SepalLength"	5.936
6	"Iris-versicolor"	"SepalWidth"	2.77
7	"Iris-versicolor"	"PetalLength"	4.26
8	"Iris-versicolor"	"PetalWidth"	1.326
9	"Iris-virginica"	"SepalLength"	6.588
10	"Iris-virginica"	"SepalWidth"	2.974
11	"Iris-virginica"	"PetalLength"	5.552
12	"Iris-virginica"	"PetalWidth"	2.026

• medias\_stacked = combine(groupby(stacked, [:Species,:variable]), :value=>mean=>
:media)

	variable	Iris-setosa	Iris-versicolor	Iris-virginica
1	"SepalLength"	5.006	5.936	6.588
2	"SepalWidth"	3.418	2.77	2.974
3	"PetalLength"	1.464	4.26	5.552
4	"PetalWidth"	0.244	1.326	2.026

unstack(medias\_stacked,:Species,:media)

## **Ordenamento**

md" ### Ordenamento"

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
1	4.3	3.0	1.1	0.1	"Iris-setosa"
2	4.4	2.9	1.4	0.2	"Iris-setosa"
3	4.4	3.0	1.3	0.2	"Iris-setosa"
4	4.4	3.2	1.3	0.2	"Iris-setosa"
5	4.5	2.3	1.3	0.3	"Iris-setosa"
6	4.6	3.1	1.5	0.2	"Iris-setosa"
7	4.6	3.2	1.4	0.2	"Iris-setosa"
8	4.6	3.4	1.4	0.3	"Iris-setosa"
9	4.6	3.6	1.0	0.2	"Iris-setosa"
10	4.7	3.2	1.3	0.2	"Iris-setosa"
m	ore				
150	7.9	3.8	6.4	2.0	"Iris-virginica"

• sort(iris)

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
1	7.9	3.8	6.4	2.0	"Iris-virginica"
2	7.7	3.8	6.7	2.2	"Iris-virginica"
3	7.7	3.0	6.1	2.3	"Iris-virginica"
4	7.7	2.8	6.7	2.0	"Iris-virginica"
5	7.7	2.6	6.9	2.3	"Iris-virginica"
6	7.6	3.0	6.6	2.1	"Iris-virginica"
7	7.4	2.8	6.1	1.9	"Iris-virginica"
8	7.3	2.9	6.3	1.8	"Iris-virginica"
9	7.2	3.6	6.1	2.5	"Iris-virginica"
10	7.2	3.2	6.0	1.8	"Iris-virginica"
m	ore				
150	4.3	3.0	1.1	0.1	"Iris-setosa"

<sup>•</sup> sort(iris, rev=true) #ordena ao contrário