

# Ejercicio 3.3

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## Carga de Datos y Librerías

```
library(naniar)
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.1      v stringr    1.5.2
v ggplot2    4.0.0      v tibble     3.3.0
v lubridate  1.9.4      v tidyr      1.3.1
v purrr      1.1.0
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
library(rio)

data <- read.csv("students_FP.csv",
  na.strings = c("", "NA", "NaN", "NULL"),
  stringsAsFactors = TRUE
)
```

## Imputación múltiple

```
library(mice)
```

Attaching package: 'mice'

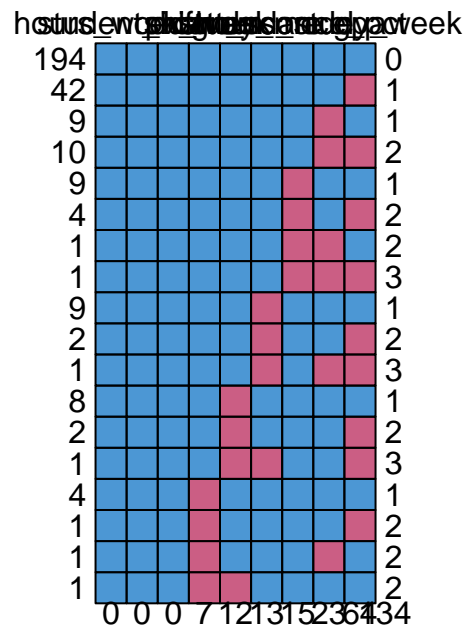
The following object is masked from 'package:stats':

filter

The following objects are masked from 'package:base':

cbind, rbind

```
md.pattern(data)
```



	student_id	hours_work_week	shift	program	exam_score	study_mode
194	1	1	1	1	1	1
42	1	1	1	1	1	1
9	1	1	1	1	1	1
10	1	1	1	1	1	1
9	1	1	1	1	1	1
4	1	1	1	1	1	1
1	1	1	1	1	1	1

1	1	1	1	1	1	1
9	1	1	1	1	1	0
2	1	1	1	1	1	0
1	1	1	1	1	1	0
8	1	1	1	1	0	1
2	1	1	1	1	0	1
1	1	1	1	1	0	0
4	1	1	1	0	1	1
1	1	1	1	0	1	1
1	1	1	1	0	1	1
1	1	1	1	0	0	1
	0	0	0	7	12	13
	attendance_pct	hours_study_week	gpa			
194	1	1	1	0		
42	1	1	0	1		
9	1	0	1	1		
10	1	0	0	2		
9	0	1	1	1		
4	0	1	0	2		
1	0	0	1	2		
1	0	0	0	3		
9	1	1	1	1		
2	1	1	0	2		
1	1	0	0	3		
8	1	1	1	1		
2	1	1	0	2		
1	1	1	0	3		
4	1	1	1	1		
1	1	1	0	2		
1	1	0	1	2		
1	1	1	1	2		
	15	23	64	134		

No hay ninguna observación con todas las variables missing.

```
impData <- mice(select(data,-student_id),m=5,maxit=50,seed=500)
```

iter	imp	variable
1	1	hours_study_week attendance_pct gpa exam_score program study_mode
1	2	hours_study_week attendance_pct gpa exam_score program study_mode
1	3	hours_study_week attendance_pct gpa exam_score program study_mode

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]



```

44 4 hours_study_week attendance_pct gpa exam_score program study_mode
44 5 hours_study_week attendance_pct gpa exam_score program study_mode
45 1 hours_study_week attendance_pct gpa exam_score program study_mode
45 2 hours_study_week attendance_pct gpa exam_score program study_mode
45 3 hours_study_week attendance_pct gpa exam_score program study_mode
45 4 hours_study_week attendance_pct gpa exam_score program study_mode
45 5 hours_study_week attendance_pct gpa exam_score program study_mode
46 1 hours_study_week attendance_pct gpa exam_score program study_mode
46 2 hours_study_week attendance_pct gpa exam_score program study_mode
46 3 hours_study_week attendance_pct gpa exam_score program study_mode
46 4 hours_study_week attendance_pct gpa exam_score program study_mode
46 5 hours_study_week attendance_pct gpa exam_score program study_mode
47 1 hours_study_week attendance_pct gpa exam_score program study_mode
47 2 hours_study_week attendance_pct gpa exam_score program study_mode
47 3 hours_study_week attendance_pct gpa exam_score program study_mode
47 4 hours_study_week attendance_pct gpa exam_score program study_mode
47 5 hours_study_week attendance_pct gpa exam_score program study_mode
48 1 hours_study_week attendance_pct gpa exam_score program study_mode
48 2 hours_study_week attendance_pct gpa exam_score program study_mode
48 3 hours_study_week attendance_pct gpa exam_score program study_mode
48 4 hours_study_week attendance_pct gpa exam_score program study_mode
48 5 hours_study_week attendance_pct gpa exam_score program study_mode
49 1 hours_study_week attendance_pct gpa exam_score program study_mode
49 2 hours_study_week attendance_pct gpa exam_score program study_mode
49 3 hours_study_week attendance_pct gpa exam_score program study_mode
49 4 hours_study_week attendance_pct gpa exam_score program study_mode
49 5 hours_study_week attendance_pct gpa exam_score program study_mode
50 1 hours_study_week attendance_pct gpa exam_score program study_mode
50 2 hours_study_week attendance_pct gpa exam_score program study_mode
50 3 hours_study_week attendance_pct gpa exam_score program study_mode
50 4 hours_study_week attendance_pct gpa exam_score program study_mode
50 5 hours_study_week attendance_pct gpa exam_score program study_mode

```

```
summary(impData)
```

Class: mids

Number of multiple imputations: 5

Imputation methods:

hours_work_week	hours_study_week	attendance_pct	gpa
" "	"pmm"	"pmm"	"pmm"
exam_score	program	study_mode	shift
"pmm"	"polyreg"	"polyreg"	" "

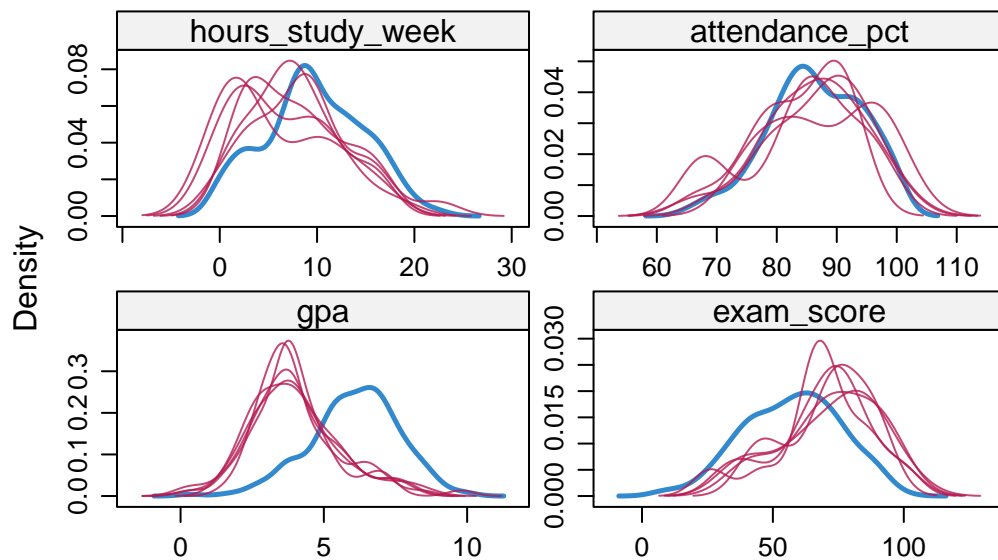
PredictorMatrix:

	hours_work_week	hours_study_week	attendance_pct	gpa	exam_score
hours_work_week	0	1	1	1	1
hours_study_week	1	0	1	1	1
attendance_pct	1	1	0	1	1
gpa	1	1	1	0	1
exam_score	1	1	1	1	0
program	1	1	1	1	1

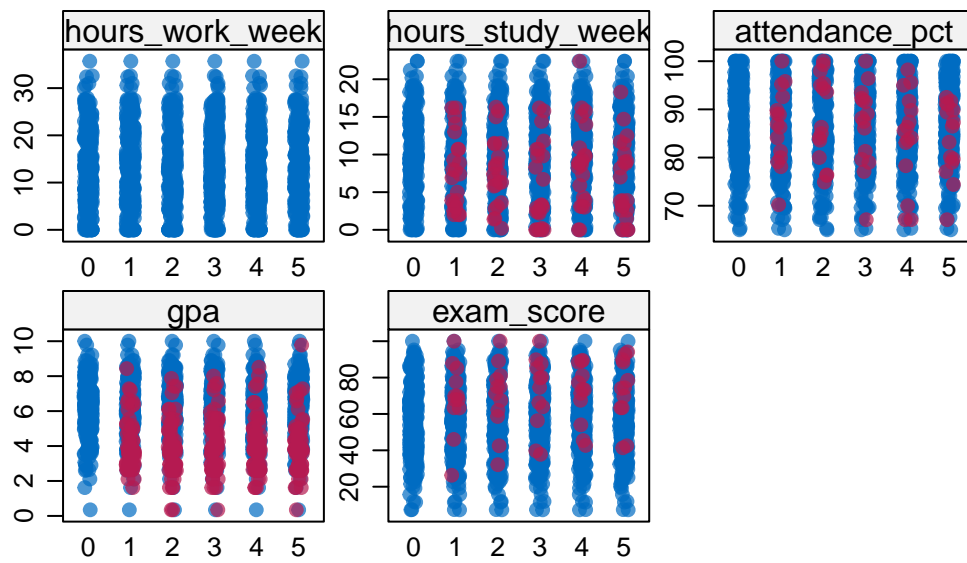
  

	program	study_mode	shift
hours_work_week	1	1	1
hours_study_week	1	1	1
attendance_pct	1	1	1
gpa	1	1	1
exam_score	1	1	1
program	0	1	1

```
###Visualizar las variables cuantitativas
densityplot(impData)
```

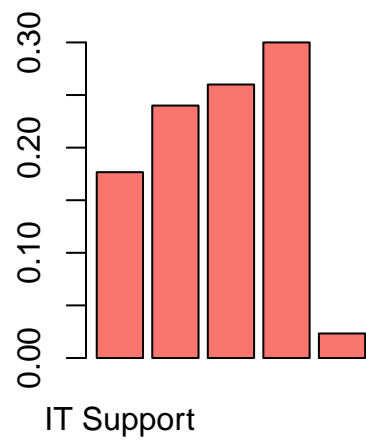


```
stripplot(impData, pch = 20, cex = 1.2)
```

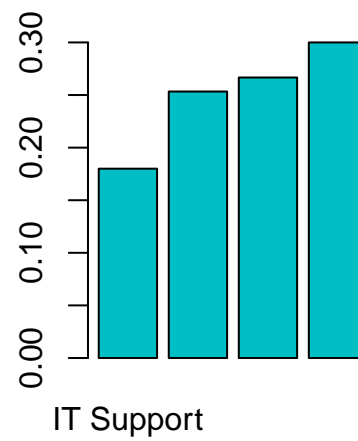


```
###Visualizar las variables cualitativas
completedData <- complete(impData,1)
par(mfrow = c(1, 2))
barplot(prop.table(table(data$program, useNA = "ifany")),
        main = "Antes de la imputación", col = "#F8766D" )
barplot(prop.table(table(completedData$program)),
        main = "Después de la imputación",col = "#00BFC4")
```

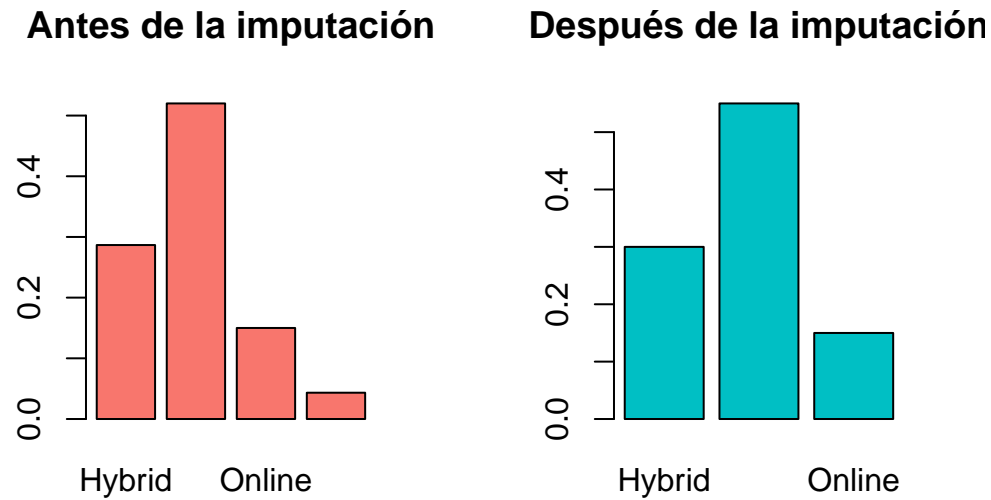
**Antes de la imputación**



**Después de la imputación**



```
par(mfrow = c(1, 2))
barplot(prop.table(table(data$study_mode, useNA = "ifany")),
        main = "Antes de la imputación", col = "#F8766D" )
barplot(prop.table(table(completedData$study_mode)),
        main = "Después de la imputación", col = "#00BFC4")
```



Excepto la variable `attendance_pct`, ninguna variable se ha imputado demasiado bien, pero la que se ha imputado muy mal es la de `gpa` que todos los valores se han imputado por debajo del resto de la distribución, esto no necesariamente está mal, precisamente una hipótesis era que los datos faltantes de `gpa` era MNAR porque podrías estar relacionados con valores bajos de `gpa`, pero no explicado por el resto de variables. En este caso, al tener en cuenta de forma multivariante toda la base de datos, ha imputado los valores de la variable como bajos desplazando así toda la distribución.

Por otro lado las dos variables cualitativas parece que se imputa de forma correcta.