Analisis estadistico

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Datos clínicos

```
library(readxl)
clin_data <- read_excel("C:/Users/Usuario1/Desktop/UOC/TFM_CRM/table_jaume_corrected.xlsx")</pre>
dim(clin_data)
## [1] 105 218
nlevels(as.factor(clin_data$redcap_data_access_group))
## [1] 6
Se dispone información sobre 105 subjects, recogida en 6 diferents hospitals, que son:
##
    clinical_hospital hospital_clnic_bar
##
                                             hospital_del_mar hospital_general_d
##
                                        10
## hospital_vall_dheb
                                parc_taul
Descriptiva general
datos <- subset (clin_data, select = c(subject_id, redcap_data_access_group, control_patient, sex, age_enrol
                                    fio2_initial, spo2_initial, bmi,resp_rate_initial, temp_initial, paco
dim(datos)
## [1] 105
#head(datos)
datos <- as. data. frame (datos)
my_data<-datos
datos$sex<-factor(datos$sex, levels=c(1,2),</pre>
                               labels = c("Femenino", "Masculino"))
datos$control_patient<-factor(datos$control_patient, levels=c(1,2),</pre>
                               labels = c("Voluntarios sanos", "Pacientes con COVID-19"))
```

Table 1

```
datos1<-datos[!is.na(datos$sex),]</pre>
library(table1)
## Attaching package: 'table1'
## The following objects are masked from 'package:base':
##
        units, units<-
                           <- "Sexo"
label(datos1$sex)
label(datos1$age_enrollment) <- "Edad"</pre>
#label(datos$control_patient) <- "Ulceration"
label(datos1$T0bs1TSI) <- "St02"</pre>
label(datos1$T0slope1) <- "De02"</pre>
label(datos1$T0slope2) <- "Re02"</pre>
label(datos1$T0minTSI) <- "St02 mínimo"</pre>
label(datos1$T0maxTSI) <- "St02 maximo"</pre>
label(datos1$TOAUC) <- "Hiperemia AUC"</pre>
label(datos1$temp_initial) <- "Temperatura"</pre>
label(datos1$resp rate initial) <- "RR"</pre>
label(datos1$paco2_initial) <- "PacO2"</pre>
label(datos1$ph_initial) <- "PH"</pre>
label(datos1$svco2_initial) <- "Svc02"</pre>
label(datos1$fio2_initial) <- "Fi02"</pre>
label(datos1$spo2_initial) <- "Sp02"</pre>
label(datos1$T0bslTHb) <- "THC"</pre>
###
      UNIDADES
units(datos1$age_enrollment) <- "años"</pre>
units(datos1$T0bslTSI) <- "%"</pre>
units(datos1$T0slope1) <- "%/min"</pre>
units(datos1$T0slope2) <- "%/min"</pre>
units(datos1$T0minTSI) <- "%"</pre>
units(datos1$T0maxTSI) <- "%"</pre>
units(datos1$TOAUC) <- "U"</pre>
units(datos1$temp_initial) <- "°C"</pre>
units(datos1$resp rate initial) <- "resp/min"</pre>
units(datos1$paco2_initial) <- "%"</pre>
units(datos1$svco2_initial) <- "%"</pre>
units(datos1$fio2_initial) <- "%"</pre>
units(datos1$spo2_initial) <- "%"</pre>
units(datos1$T0bslTHb) <- "uM/L"</pre>
table1(~ sex + age_enrollment + bmi +temp_initial + resp_rate_initial + fio2_initial + spo2_initial + T
        overall = "Overall", transpose=FALSE)
```

Get nicer `table1` LaTeX output by simply installing the `kableExtra` package

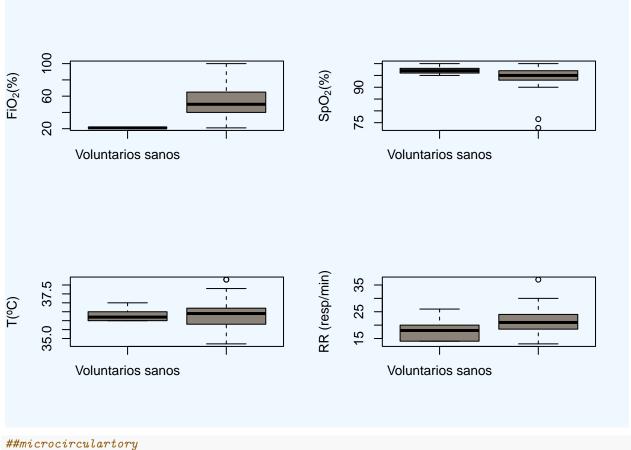
Voluntarios sanos	Pacientes con COVID-19	Overall
(N=32)	(N=72)	(N=104)

	Voluntarios sanos	Pacientes con COVID-19	Overall
Sexo			
Femenino	16 (50.0%)	21~(29.2%)	37 (35.6%)
Masculino	16 (50.0%)	51 (70.8%)	67 (64.4%)
Edad (años)	(
Mean (SD)	33.7 (10.1)	58.8 (13.1)	51.1 (16.9)
Median [Min, Max]	31.5 [24.0, 70.0]	61.5 [23.0, 88.0]	53.5 [23.0, 88.0]
bmi	, ,	, ,	, ,
Mean (SD)	22.7(2.96)	30.0 (5.54)	27.9 (5.94)
Median [Min, Max]	22.5 [18.6, 33.3]	29.4 [19.8, 51.4]	27.4 [18.6, 51.4]
Missing	4 (12.5%)	5 (6.9%)	9 (8.7%)
Temperatura (°C)			
Mean (SD)	36.3 (0.314)	$36.2 \ (0.752)$	36.3 (0.654)
Median [Min, Max]	36.2 [36.0, 37.0]	36.4 [34.7, 38.3]	36.2 [34.7, 38.3]
Missing	5 (15.6%)	6 (8.3%)	11 (10.6%)
RR (resp/min)	- (/)	2 (3.3,0)	(===,=)
Mean (SD)	17.9 (3.33)	21.5 (4.59)	20.5(4.57)
Median [Min, Max]	18.0 [14.0, 26.0]	21.0 [13.0, 37.0]	20.0 [13.0, 37.0]
Missing	10 (31.3%)	12 (16.7%)	22 (21.2%)
FiO2 (%)	10 (01.070)	12 (1011/0)	== (======)
Mean (SD)	21.0 (0)	53.5 (19.6)	42.2 (22.2)
Median [Min, Max]	21.0 [21.0, 21.0]	50.0 [21.0, 100]	40.0 [21.0, 100]
Missing	0 (0%)	12 (16.7%)	12 (11.5%)
SpO2 (%)	0 (070)	12 (101170)	12 (111070)
Mean (SD)	97.3 (1.63)	94.6 (4.15)	95.5 (3.74)
Median [Min, Max]	97.0 [95.0, 100]	95.0 [72.8, 100]	96.0 [72.8, 100]
Missing	0 (0%)	4 (5.6%)	4 (3.8%)
StO2 (%)	(0,0)	- (0.0,0)	= (0.0,0)
Mean (SD)	65.2(3.55)	66.9 (5.94)	66.3 (5.36)
Median [Min, Max]	65.7 [57.7, 71.6]	66.8 [51.1, 84.2]	66.3 [51.1, 84.2]
THC (uM/L)	[,]	[- , -]	[- , -]
Mean (SD)	45.1 (11.2)	43.3 (15.0)	43.8 (13.9)
Median [Min, Max]	44.1 [25.0, 79.9]	43.0 [4.28, 88.8]	43.9 [4.28, 88.8]
DeO2 (%/min)	. , ,	, ,	, ,
Mean (SD)	-9.15 (2.37)	-5.22 (1.97)	-6.43 (2.77)
Median [Min, Max]	-9.17 [-13.5, -3.63]	-5.25 [-12.2, -1.43]	-5.96 [-13.5, -1.43]
ReO2 (%/min)	, ,	, ,	, ,
Mean (SD)	120 (55.5)	77.5 (37.1)	90.6 (47.6)
Median [Min, Max]	103 [33.3, 242]	69.9 [12.4, 223]	79.1 [12.4, 242]
StO2 mínimo (%)	[,]	[, , ,
Mean (SD)	39.9 (7.63)	51.6 (8.55)	48.0 (9.88)
Median [Min, Max]	39.7 [21.6, 55.4]	52.5 [34.1, 68.7]	48.6 [21.6, 68.7]
StO2 màximo (%)	L / / 1	. , 1	r / -1
Mean (SD)	78.2 (4.83)	74.2 (5.66)	75.4 (5.72)
Median [Min, Max]	79.0 [67.6, 86.8]	74.8 [57.2, 90.6]	75.9 [57.2, 90.6]
Hiperemia AUC (U)	[,]	[,]	r /1
Mean (SD)	14.8 (6.85)	8.56 (5.00)	10.5 (6.31)
Median [Min, Max]	14.0 [3.82, 35.8]	7.51 [0.0501, 25.8]	9.52 [0.0501, 35.8]

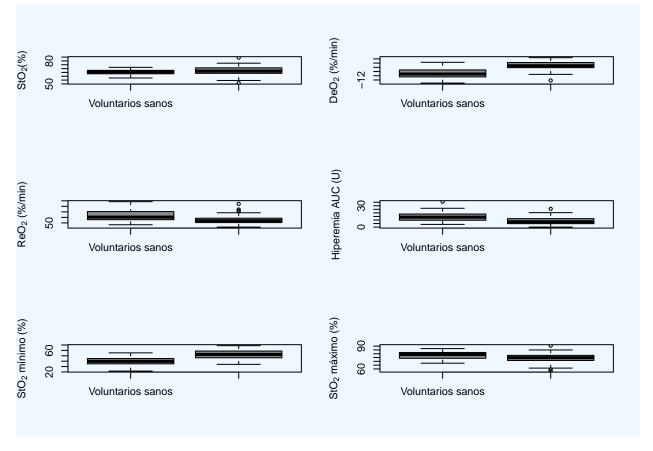
GRAFICAS

- Distribución de las variables en casos y controles
 Estudio de la normalidad (qqplot, residous...)

ESTUDIO DE T0bslTSI, T0slope1, T0slope2, T0minTSI, T0min2max, T0AUC en casos y controles - MICROCIRCULARITY STATUS



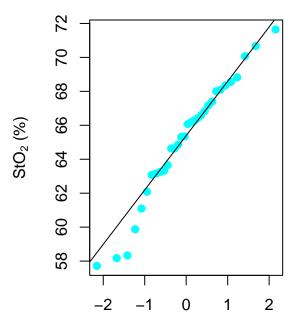
```
ylab=expression(paste('St0'[2],"(%)")),
        xlab="")
boxplot(datos1$T0slope1 ~ datos1$control_patient, col="azure4",
        ylab=expression(paste('DeO'[2], " (%/min)")),
        xlab="")
boxplot(datos1$T0slope2 ~ datos1$control_patient, col="azure4",
        ylab=expression(paste('ReO'[2], " (%/min)")),
        xlab="")
boxplot(datos1$TOAUC ~ datos1$control_patient, col="azure4",
        ylab="Hiperemia AUC (U)",
        xlab="")
boxplot(datos1$T0minTSI ~ datos1$control_patient, col="azure4",
        ylab=expression(paste('StO'[2]," minimo (%)")),
        xlab="")
boxplot(datos1$T0maxTSI ~ datos1$control_patient, col="azure4",
        ylab=expression(paste('StO'[2]," máximo (%)")),
        xlab="")
```



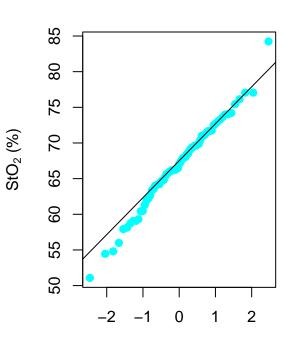
t-test

#normalidad

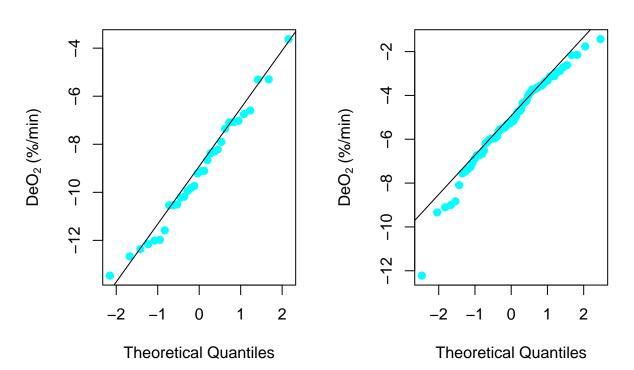
#st02

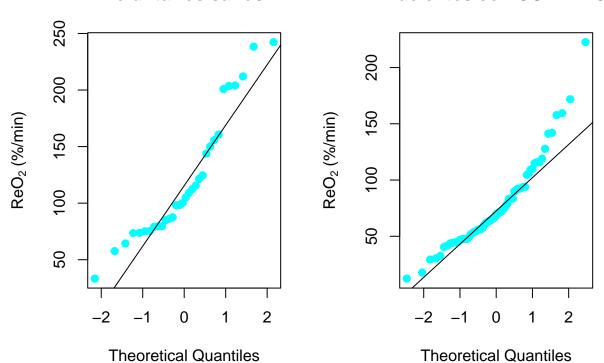


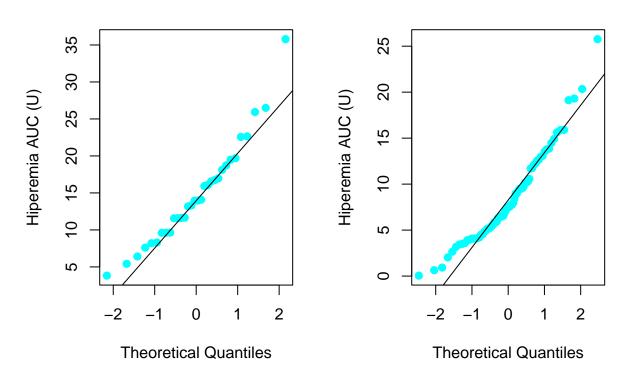


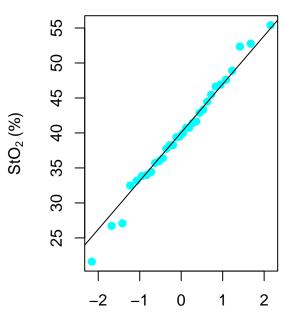


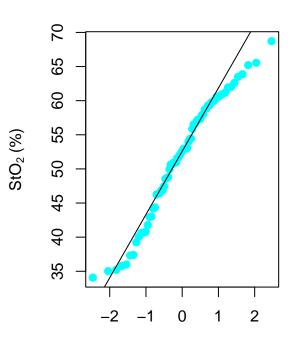
Theoretical Quantiles





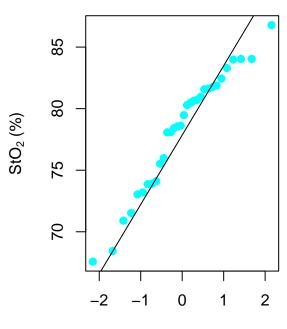


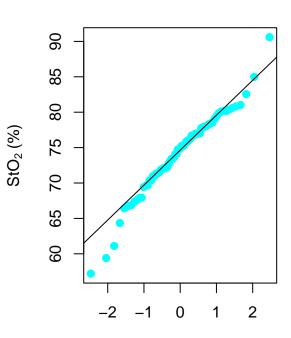




Theoretical Quantiles

Theoretical Quantiles





Theoretical Quantiles

Theoretical Quantiles

```
library(nortest)
st_max<-by(data = datos,INDICES = datos$control_patient,FUN = function(x){ lillie.test(x$TOmaxTSI)})
pv6<-st max$`Voluntarios sanos`$p.value
pc6<-st_max$`Pacientes con COVID-19`$p.value
pv<-rbind(pv1, pv2, pv3, pv4, pv5, pv6)
pc<-rbind(pc1, pc2, pc3, pc4, pc5, pc6)
pvalores<-cbind(pv, pc)</pre>
colnames(pvalores)<-c("Voluntarios sanos", "Pacientes con COVID-19")</pre>
rownames(pvalores)<-c("St02", "De02", "Re02", "AUC", "St02 minimo", "St02 maximo")</pre>
test1 <- t.test(datos$TObslTSI ~ datos$control_patient)</pre>
res_test1<-cbind(test1$estimate[[1]], test1$estimate[[2]], test1$p.value)
test2 <- t.test(datos$TOslope1 ~ datos$control_patient)</pre>
res_test2<-cbind(test2\sestimate[[1]], test2\sestimate[[2]], test2\sestimate[[2]],
test3 <- t.test(datos$TOslope2 ~ datos$control_patient)</pre>
res_test3<-cbind(test3\stimate[[1]], test3\stimate[[2]], test3\sp.value)
test4 <- t.test(datos$TOAUC ~ datos$control patient)</pre>
res_test4<-cbind(test4$estimate[[1]], test4$estimate[[2]], test4$p.value)
test5 <- t.test(datos$TOminTSI ~ datos$control_patient)</pre>
res_test5<-cbind(test5\setimate[[1]], test5\setimate[[2]], test5\sp.value)
```

```
test6 <- t.test(datos$T0maxTSI ~ datos$control_patient)
res_test6<-cbind(test6$estimate[[1]], test6$estimate[[2]], test6$p.value)

ttest_cv<-rbind(res_test1, res_test2, res_test3, res_test4, res_test5, res_test6)
colnames(ttest_cv)<-c("Media en Voluntarios sanos", "Media en Pacientes con COVID-19", "P-Valor")
rownames(ttest_cv)<-c("St02", "De02", "Re02", "AUC", "st02 minimo", "St02 máximo")</pre>
```

En este caso el p-valor es ligeramente inferior al umbral habitual de 0.05, por lo que en muchos contextos se hablaría de rechazar la hipótesis de igualdad de medias efectivamente aceptando diferencia entre medias.

Estudio de Normalidad de T0bslTSI, T0slope1, T0slope2, T0AUC

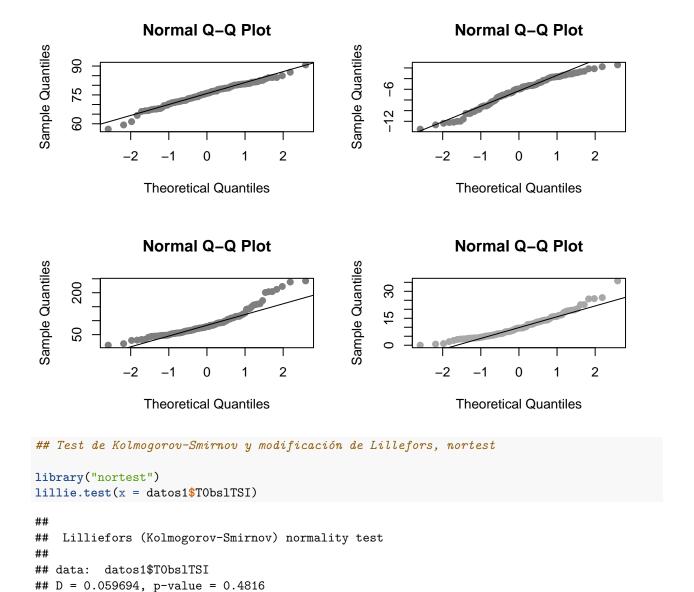
```
library(ggplot2)
require(devtools)

## Loading required package: devtools
## Loading required package: usethis
par(mfrow=c(2,2))

qqnorm(datos1$TOmaxTSI, pch = 19, col = "gray50")
qqline(datos1$TOmaxTSI)

qqnorm(datos1$TOslope1, pch = 19, col = "gray50")
qqline(datos1$TOslope1)

qqnorm(datos1$TOslope2, pch = 19, col = "gray50")
qqline(datos1$TOslope2)
qqnorm(datos1$TOslope2)
qqnorm(datos1$TOslope2)
```



Descriptiva de los casos

MICROCIRCULARITY STATUS EN COVID-19 PATIENTS

- En función de mecanical ventilation microcirculatory status according to the degree of respiratory support
- En función de leve y grave (ARDS) microcirculatory status according to severity of ARDS

mecanical ventilation

```
#head(clin_data)
datos_pac<-subset(clin_data, select=c(subject_id, measurement_identifier, death, redcap_data_access_groupac_cov<-datos_pac[datos_pac$control_patient==2,]</pre>
```

```
pac_cov$sex<-factor(pac_cov$sex, levels=c(1,2),</pre>
                                 labels = c("Femenino", "Masculino"))
pac_cov$invasive_mechanical_ventil<-factor(pac_cov$invasive_mechanical_ventil, levels=c(1,2),
                                 labels = c("IMV", "NRS"))
func_fac<-function(x){</pre>
  x < -factor(x, levels = c(1,2),
         labels = c("Si", "No"))
}
vars_fac<-subset(pac_cov, select = c(hypertension, previous_heart_failure, lung_disease, diabetes, prev</pre>
vars_fac2<-lapply(vars_fac, func_fac)</pre>
pac_cov$ischemic_heart_disease<-as.factor(pac_cov$ischemic_heart_disease)
pac_cov$atrial_fibrillation<-as.factor(pac_cov$atrial_fibrillation)</pre>
pac_cov2<-subset(pac_cov, select=c(invasive_mechanical_ventil, age_enrollment, sex, ischemic_heart_dise
                                      paco2_initial, ph_initial, svco2_initial))
datos_pac<-cbind(pac_cov2, vars_fac2)</pre>
library(table1)
label(datos pac$T0bslTHb) <- "THC"</pre>
label(datos_pac$bmi) <- "BMI"</pre>
label(datos_pac$sex)
                             <- "Sexo"
label(datos_pac$age_enrollment) <- "Edad"</pre>
label(datos_pac$T0bslTSI) <- "St02"</pre>
label(datos_pac$T0slope1) <- "De02"</pre>
label(datos_pac$T0slope2) <- "Re02"</pre>
label(datos_pac$T0minTSI) <- "St02 mínimo"</pre>
label(datos_pac$T0maxTSI) <- "St02 maximo"</pre>
label(datos_pac$TOAUC) <- "Hiperemia AUC"</pre>
label(datos_pac$temp_initial) <- "Temperatura"</pre>
label(datos_pac$resp_rate_initial) <- "RR"</pre>
label(datos_pac$paco2_initial) <- "PacO2"</pre>
label(datos_pac$ph_initial) <- "PH"</pre>
label(datos_pac$svco2_initial) <- "Svc02"</pre>
label(datos_pac$fio2_initial) <- "Fi02"</pre>
label(datos_pac$spo2_initial) <- "Sp02"</pre>
#enfermedades
label(datos_pac$hypertension) <- "Hipertensión"</pre>
label(datos_pac$previous_heart_failure) <- "Insuficiencia cardíaca previa"</pre>
label(datos_pac$lung_disease) <- "Enfermedad pulmonar"</pre>
label(datos_pac$diabetes) <- "Diabetes"</pre>
label(datos_pac$previous_lung_disease) <- "Enfermedad pulmonar previa"</pre>
label(datos_pac$smoker) <- "Fumador/a"</pre>
label(datos_pac$renal_insufficiency) <- "Insuficiencia renal"</pre>
      UNIDADES
units(datos_pac$age_enrollment) <- "años"</pre>
```

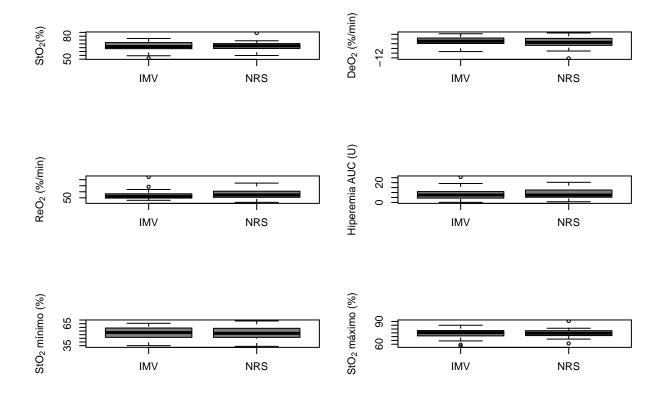
```
units(datos_pac$T0bslTSI) <- "%"</pre>
units(datos_pac$T0slope1) <- "%/min"</pre>
units(datos_pac$T0slope2) <- "%/min"</pre>
units(datos_pac$T0minTSI) <- "%"</pre>
units(datos_pac$T0maxTSI) <- "%"</pre>
units(datos_pac$TOAUC) <- "U"</pre>
units(datos_pac$temp_initial) <- "ºC"
units(datos pac$resp rate initial) <- "resp/min"</pre>
units(datos_pac$paco2_initial) <- "%"</pre>
units(datos_pac$svco2_initial) <- "%"</pre>
units(datos_pac$fio2_initial) <- "%"</pre>
units(datos_pac$spo2_initial) <- "%"</pre>
units(datos_pac$T0bslTHb) <- "uM/L"</pre>
table1(~ sex + age_enrollment + bmi + hypertension + previous_heart_failure +
         lung_disease + diabetes + previous_lung_disease + smoker + renal_insufficiency +
         TObslTSI + TObslTHb + TOslope1 + TOslope2 + TOAUC + temp_initial +
         resp_rate_initial + fio2_initial + spo2_initial + paco2_initial + svco2_initial
       + ischemic_heart_disease + atrial_fibrillation | invasive_mechanical_ventil,
       data=datos_pac, overall = "Overall", transpose=FALSE)
```

Get nicer `table1` LaTeX output by simply installing the `kableExtra` package

	IMV	NRS	Overall	
	(N=39)	(N=34)	(N=73)	
Sexo	, ,	,	, ,	
Femenino	10~(25.6%)	11 (32.4%)	$21\ (28.8\%)$	
Masculino	28 (71.8%)	23~(67.6%)	51 (69.9%)	
Missing	1(2.6%)	0 (0%)	1 (1.4%)	
Edad (años)				
Mean (SD)	59.4 (11.8)	58.5 (14.5)	59.0 (13.0)	
Median [Min, Max]	62.0 [30.0, 79.0]	61.5 [23.0, 88.0]	62.0 [23.0, 88.0]	
BMI	-	•		
Mean (SD)	30.1 (4.76)	29.9(6.41)	30.0(5.54)	
Median [Min, Max]	29.7 [23.1, 47.0]	29.1 [19.8, 51.4]	29.4 [19.8, 51.4]	
Missing	3 (7.7%)	3 (8.8%)	6 (8.2%)	
Hipertensión	,	, ,	,	
Si	18 (46.2%)	15 (44.1%)	33 (45.2%)	
No	$21\ (53.8\%)$	19 (55.9%)	40 (54.8%)	
Insuficiencia cardíaca previa	, ,	,	,	
Si	3(7.7%)	1(2.9%)	4 (5.5%)	
No	36~(92.3%)	33 (97.1%)	69 (94.5%)	
Enfermedad pulmonar	, ,	,	,	
Si	37 (94.9%)	24 (70.6%)	61 (83.6%)	
No	2(5.1%)	10 (29.4%)	12 (16.4%)	
Diabetes	,	,	, ,	
Si	8 (20.5%)	9 (26.5%)	17(23.3%)	
No	31(79.5%)	25(73.5%)	56 (76.7%)	
Enfermedad pulmonar previa	` '	, ,	` /	
Si	2(5.1%)	5 (14.7%)	7(9.6%)	
No	37 (94.9%)	29(85.3%)	66 (90.4%)	
Fumador/a	,	,	, ,	

	IMV	NRS	Overall
Si	7 (17.9%)	11 (32.4%)	18 (24.7%)
No	31(79.5%)	23~(67.6%)	54 (74.0%)
Missing	1(2.6%)	0 (0%)	1 (1.4%)
Insuficiencia renal	(1, 1)	- (-,-)	(, , ,
Si	6 (15.4%)	1(2.9%)	7 (9.6%)
No	33 (84.6%)	33 (97.1%)	66 (90.4%)
StO2 (%)	33 (811070)	33 (01.170)	00 (00.170)
Mean (SD)	67.0 (6.18)	66.9 (5.78)	67.0 (5.96)
Median [Min, Max]	66.5 [51.1, 77.1]	67.5 [54.8, 84.2]	67.0 [51.1, 84.2]
THC (uM/L)	00.9 [91.1, 77.1]	01.0 [04.0, 04.2]	01.0 [01.1, 04.2]
Mean (SD)	41.4 (13.0)	44.2 (18.2)	42.7 (15.6)
Median [Min, Max]	` /	40.8 [4.28, 88.8]	42.7 (13.0) 42.9 [2.31, 88.8]
	43.1 [2.31, 59.3]	40.6 [4.26, 66.6]	42.9 [2.31, 00.0]
DeO2 (%/min)	4.00 (1.75)	F FO (0.14)	F 10 (1 0C)
Mean (SD)	-4.86 (1.75)	-5.58 (2.14)	-5.19 (1.96)
Median [Min, Max]	-4.94 [-9.34, -1.77]	-5.42 [-12.2, -1.43]	-5.23 [-12.2, -1.43]
ReO2 (%/min)	T 0.1 (00.1)	00.0 (07.0)	0 (0- 1)
Mean (SD)	72.1 (36.1)	82.6 (37.9)	77.0 (37.1)
Median [Min, Max]	62.0 [29.3, 223]	73.8 [12.4, 172]	69.5 [12.4, 223]
Hiperemia AUC (U)			, ,
Mean (SD)	8.21 (5.21)	8.83 (4.79)	8.49 (4.99)
Median [Min, Max]	7.50 [0.0501, 25.8]	7.49 [0.637, 20.3]	7.50 [0.0501, 25.8]
Temperatura (°C)			
Mean (SD)	$36.2 \; (0.885)$	$36.3 \ (0.544)$	$36.2 \ (0.747)$
Median [Min, Max]	36.0 [34.7, 38.3]	36.5 [35.2, 37.3]	36.4 [34.7, 38.3]
Missing	3(7.7%)	3~(8.8%)	6 (8.2%)
RR (resp/min)			
Mean (SD)	21.9(4.03)	21.3 (5.57)	21.7(4.68)
Median [Min, Max]	22.0 [14.0, 30.0]	20.0 [13.0, 37.0]	21.0 [13.0, 37.0]
Missing	3(7.7%)	9 (26.5%)	12 (16.4%)
FiO2 (%)			
Mean (SD)	53.0 (20.8)	56.2 (19.9)	54.2 (20.3)
Median [Min, Max]	45.0 [30.0, 100]	60.0 [21.0, 100]	50.0 [21.0, 100]
Missing	2 (5.1%)	10 (29.4%)	12 (16.4%)
SpO2 (%)	,	,	,
Mean (SD)	95.9 (1.91)	93.2 (5.40)	94.6 (4.12)
Median [Min, Max]	96.0 [92.0, 100]	94.0 [72.8, 100]	95.0 [72.8, 100]
Missing	2 (5.1%)	2(5.9%)	4 (5.5%)
PacO2 (%)	= (0.170)	= (0.070)	1 (0.070)
Mean (SD)	45.0 (8.46)	37.5 (4.50)	42.9 (8.26)
Median [Min, Max]	42.4 [33.6, 66.3]	38.2 [28.2, 43.1]	41.4 [28.2, 66.3]
Missing	3 (7.7%)	20 (58.8%)	23 (31.5%)
SvcO2 (%)	3 (1.170)	20 (90.070)	20 (31.070)
Mean (SD)	78.9 (11.8)	74.3 (3.18)	78.2 (11.0)
	` ,		
Median [Min, Max]	76.0 [64.6, 100]	74.3 [72.0, 76.5]	76.0 [64.6, 100]
Missing	28 (71.8%)	32 (94.1%)	60 (82.2%)
ischemic_heart_disease	4 (10 907)	1 (0.007)	r (e oo4)
1	4 (10.3%)	1(2.9%)	5 (6.8%)
2	34 (87.2%)	33 (97.1%)	67 (91.8%)
Missing	1 (2.6%)	0 (0%)	1 (1.4%)
atrial_fibrillation	- /- 00	- (-04)	- /
1	3 (7.7%)	0 (0%)	3 (4.1%)
2	36 (92.3%)	34 (100%)	70~(95.9%)

```
par(mfrow=c(3,2))
boxplot(datos_pac$T0bslTSI~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('StO'[2], "(%)")),
        xlab="")
boxplot(datos_pac$TOslope1~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('DeO'[2], " (%/min)")),
        xlab="")
boxplot(datos_pac$T0slope2~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('ReO'[2], " (%/min)")),
        xlab="")
boxplot(datos_pac$TOAUC~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('Hiperemia AUC', " (U)")),
       xlab="")
boxplot(datos_pac$T0minTSI~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('St0'[2], " minimo (%)")),
       xlab="")
boxplot(datos_pac$T0maxTSI~datos_pac$invasive_mechanical_ventil,
        col = "gray50",
        ylab=expression(paste('St0'[2], " máximo (%)")),
       xlab="")
```



ARDS

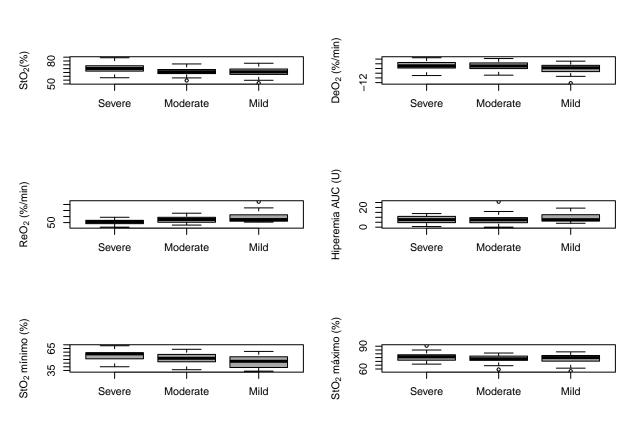
Graficas

Get nicer `table1` LaTeX output by simply installing the `kableExtra` package

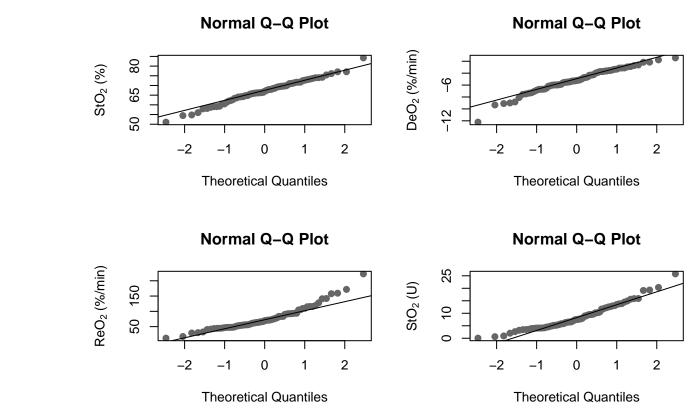
	Severe	Moderate	Mild	Overall
	(N=16)	(N=25)	(N=20)	(N=73)
Sexo				
Femenino	0 (0%)	7(28.0%)	7(35.0%)	$21\ (28.8\%)$
Masculino	15~(93.8%)	18 (72.0%)	13~(65.0%)	51 (69.9%)
Missing	1(6.3%)	0 (0%)	0 (0%)	1 (1.4%)
Edad (años)	,	,	, ,	,
Mean (SD)	60.9(13.3)	$61.4\ (10.6)$	52.6 (13.2)	59.0 (13.0)
Median [Min, Max]	64.0 [27.0, 76.0]	62.0 [38.0, 79.0]	53.0 [23.0, 71.0]	62.0 [23.0, 88.0]
BMI				
Mean (SD)	30.8 (5.61)	29.5(5.59)	29.6(3.72)	30.0(5.54)
Median [Min, Max]	29.0 [24.4, 46.5]	29.4 [19.8, 47.0]	29.9 [21.8, 36.5]	29.4 [19.8, 51.4]
Missing	1 (6.3%)	0 (0%)	0 (0%)	6 (8.2%)
StO2 (%)	,	,	,	,
Mean (SD)	69.9 (6.44)	65.7(4.83)	65.5 (6.50)	67.0 (5.96)
Median [Min, Max]	70.2 [58.1, 84.2]	65.9 [54.5, 76.1]	65.8 [51.1, 77.1]	67.0 [51.1, 84.2]
THC (uM/L)	. , ,	. , ,	. , 1	. , ,
Mean (SD)	43.0 (22.4)	43.9 (12.8)	45.0 (11.1)	42.7 (15.6)
Median [Min, Max]	42.8 [2.31, 88.8]	45.9 [18.2, 74.5]	47.6 [15.7, 60.8]	42.9 [2.31, 88.8]
DeO2 (%/min)	[,]			[,]
Mean (SD)	-4.80 (1.83)	-4.87 (1.83)	-6.17 (2.25)	-5.19 (1.96)
Median [Min, Max]	-4.86 [-9.00, -1.43]	-4.94 [-8.83, -1.77]	-5.71 [-12.2, -2.89]	-5.23 [-12.2, -1.43]
ReO2 (%/min)	1.00 [0.00, 1.10]	1.01[0.00, 1.11]	5[12.2, 2.00]	0.20 [12.2, 11.10]
Mean (SD)	55.7 (23.3)	78.2 (27.2)	93.4 (46.8)	77.0 (37.1)
Median [Min, Max]	57.5 [12.4, 93.9]	75.8 [29.3, 128]	75.1 [53.7, 223]	69.5 [12.4, 223]
StO2 mínimo (%)	37.3 [12.1, 00.0]	10.0 [20.0, 120]	10.1 [00.1, 220]	00.0 [12.1, 220]
Mean (SD)	55.4 (7.61)	51.2 (7.89)	47.4 (9.14)	51.8 (8.64)
Median [Min, Max]	57.6 [40.1, 68.7]	51.6 [36.0, 63.9]	47.5 [34.1, 61.0]	52.6 [34.1, 68.7]
StO2 màximo (%)	91.0 [10.1, 00.1]	01.0 [00.0, 00.0]	11.0 [01.1, 01.0]	02.0 [01.1, 00.1]
Mean (SD)	75.8 (6.45)	73.3 (5.26)	73.6 (6.47)	74.2 (5.64)
Median [Min, Max]	76.0 [66.4, 90.6]	73.4 [59.4, 81.0]	75.0 [57.2, 82.6]	74.9 [57.2, 90.6]
Hiperemia AUC (U)	70.0 [00.4, 50.0]	10.4 [00.4, 01.0]	10.0 [01.2, 02.0]	14.0 [01.2, 00.0]
Mean (SD)	7.60 (3.91)	8.25 (5.58)	9.62 (4.79)	8.49 (4.99)
Median [Min, Max]	7.64 [0.637, 13.8]	7.53 [0.0501,	7.62 [3.96, 19.3]	7.50 [0.0501,
Median [Min, Max]	1.04 [0.037, 13.0]	25.8]	1.02 [5.30, 13.5]	25.8
Temperatura (°C)		29.0]		20.0]
Mean (SD)	26.2 (0.727)	26.1 (0.745)	26.2 (0.855)	26.2 (0.747)
	36.3 (0.727)	36.1 (0.745) 36.0 [34.7, 37.3]	$36.2 \ (0.855)$	36.2 (0.747)
Median [Min, Max]	36.2 [35.5, 38.3]	2 (8.0%)	36.0 [35.0, 38.3]	36.4 [34.7, 38.3]
Missing	0 (0%)	2 (8.0%)	0 (0%)	6 (8.2%)
RR (resp/min)	22.5 (4.16)	22.0 (5.97)	20.7 (2.44)	21 7 (4.69)
Mean (SD)	22.5 (4.16)	22.0 (5.87)	20.7 (3.44)	21.7 (4.68)
Median [Min, Max]	22.0 [14.0, 30.0]	21.5 [13.0, 37.0]	20.5 [13.0, 28.0]	21.0 [13.0, 37.0]
Missing	0 (0%)	1 (4.0%)	0 (0%)	$12 \ (16.4\%)$
FiO2 (%)	01.6 (10.5)	FO 0 (F CF)	04.0 (5.74)	F 4 0 (00 0)
Mean (SD)	81.6 (13.5)	52.8 (7.65)	34.2 (5.74)	54.2 (20.3)
Median [Min, Max]	77.5 [65.0, 100]	50.0 [40.0, 65.0]	35.0 [21.0, 40.0]	50.0 [21.0, 100]
Missing	0 (0%)	0 (0%)	0 (0%)	$12 \ (16.4\%)$
SpO2 (%)	05 0 (0.00)	05 1 (0.05)	05 0 (1 50)	0.4.0 (4.10)
Mean (SD)	95.3 (2.88)	95.1 (2.07)	95.9 (1.79)	94.6 (4.12)
Median [Min, Max]	95.0 [90.0, 100]	95.0 [92.0, 100]	96.0 [91.0, 99.0]	95.0 [72.8, 100]
Missing	0 (0%)	0 (0%)	0 (0%)	4 (5.5%)
PacO2 (%)			44.0.75.75	10.0 (0.5.3)
Mean (SD)	44.5 (11.4)	$44.5 \ (8.66)$	41.0 (5.43)	42.9 (8.26)

	Severe	Moderate	Mild	Overall
Median [Min, Max]	41.7 [28.2, 65.0]	41.5 [35.0, 66.3]	39.5 [33.6, 52.0]	41.4 [28.2, 66.3]
Missing	5 (31.3%)	6 (24.0%)	4 (20.0%)	23 (31.5%)
PH	, ,	,	, ,	, ,
Mean (SD)	$7.40 \ (0.0660)$	$7.38 \ (0.0698)$	$7.41\ (0.0585)$	7.28(0.792)
Median [Min, Max]	7.41 [7.27, 7.53]	7.39 [7.22, 7.48]	7.42 [7.31, 7.55]	7.40 [1.70, 7.55]
Missing	3 (18.8%)	6 (24.0%)	4 (20.0%)	21 (28.8%)
SvcO2 (%)				
Mean (SD)	100(0)	76.0(6.62)	72.7(4.86)	78.2 (11.0)
Median [Min, Max]	100 [100, 100]	76.5 [65.9, 84.5]	73.4 [64.6, 79.0]	76.0 [64.6, 100]
Missing	14 (87.5%)	20 (80.0%)	14 (70.0%)	60 (82.2%)
Hipertensión	, ,	, ,	, ,	, ,
Si	11 (68.8%)	11 (44.0%)	5(25.0%)	33 (45.2%)
No	5 (31.3%)	14 (56.0%)	15(75.0%)	40 (54.8%)
Insuficiencia cardíaca	, ,	, ,	` '	` '
previa				
Si	0 (0%)	2(8.0%)	1(5.0%)	4 (5.5%)
No	16 (100%)	23(92.0%)	19 (95.0%)	69(94.5%)
Enfermedad pulmonar	,	, ,	, ,	, ,
Si	12 (75.0%)	25 (100%)	17 (85.0%)	61~(83.6%)
No	4 (25.0%)	$0 \ (0\%)$	3 (15.0%)	12 (16.4%)
Diabetes	,	,	,	, ,
Si	3 (18.8%)	7(28.0%)	4(20.0%)	17(23.3%)
No	13(81.3%)	18(72.0%)	16 (80.0%)	56~(76.7%)
Enfermedad pulmonar	, ,	, ,	, ,	, ,
previa				
Si	0 (0%)	3(12.0%)	1(5.0%)	7~(9.6%)
No	16 (100%)	22~(88.0%)	19 (95.0%)	66 (90.4%)
Fumador/a	,	, ,	, ,	, ,
Si	3 (18.8%)	4(16.0%)	6(30.0%)	18 (24.7%)
No	13 (81.3%)	21 (84.0%)	13(65.0%)	54 (74.0%)
Missing	$0 \ (0\%)$	0 (0%)	1(5.0%)	1 (1.4%)
Insuficiencia renal	, ,	` '	, ,	, ,
Si	1(6.3%)	4 (16.0%)	2(10.0%)	7(9.6%)
No	15 (93.8%)	21 (84.0%)	18 (90.0%)	66 (90.4%)
ischemic_heart_disease	,	, ,	, ,	, ,
1	0 (0%)	3(12.0%)	1(5.0%)	5(6.8%)
2	16 (100%)	21 (84.0%)	19 (95.0%)	67 (91.8%)
Missing	0 (0%)	1 (4.0%)	0 (0%)	1 (1.4%)
atrial_fibrillation	,	, ,	,	` /
1	0 (0%)	3(12.0%)	0 (0%)	3 (4.1%)
2	16 (100%)	22 (88.0%)	20 (100%)	70 (95.9%)

```
xlab="")
boxplot(datos_pac$T0slope2~datos_pac$ARDS,
        col = "gray66",
        ylab=expression(paste('ReO'[2], " (%/min)")),
        xlab="")
boxplot(datos_pac$TOAUC~datos_pac$ARDS,
        col = "gray66",
        ylab=expression(paste('Hiperemia AUC', " (U)")),
        xlab="")
boxplot(datos_pac$TOminTSI~datos_pac$ARDS,
        col = "gray66",
        ylab=expression(paste('St0'[2], " minimo (%)")),
        xlab="")
boxplot(datos_pac$T0maxTSI~datos_pac$ARDS,
        col = "gray66",
        ylab=expression(paste('St0'[2], " máximo (%)")),
        xlab="")
```



```
par(mfrow=c(2,2))
qqnorm(datos_pac$TObslTSI, pch = 19, col = "gray40",
```



NORMALIDAD

PARAMETROS

- StO2 baseline
- DeO2
- ReO2
- AUC

Correlation matrix

```
library(corrplot)
```

corrplot 0.89 loaded

```
library(ggplot2)
library(PerformanceAnalytics)
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##
       legend
nirs<-subset(datos_pac, select = c(ARDS ,T0bslTSI, T0bslTHb, T0slope1, T0slope2, T0minTSI, T0maxTSI, T0
names(nirs)<-c("ARDS", "St02", "THC", "De02", "Re02", "Valor minimo", "Valor maximo", "Hiperemia AUC")</pre>
nirs$ARDS<-as.numeric(nirs$ARDS)</pre>
mcor<-cor(na.omit(nirs))</pre>
correlacion<-round(mcor, 2)</pre>
corrplot(correlacion, method="number", type="upper")
```

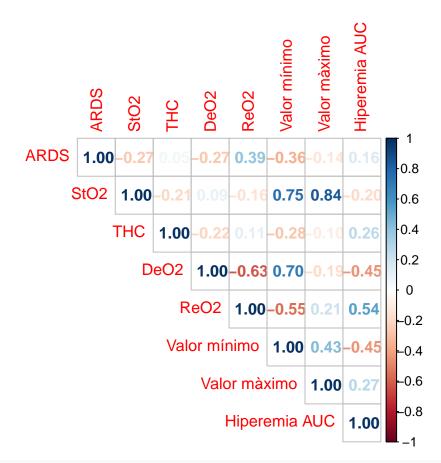
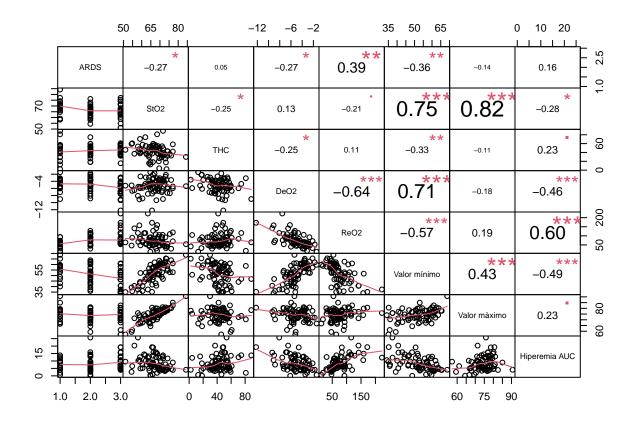


chart.Correlation(nirs, histogram = F, pch = 19)



ANOVA para los parametros de microcirculación ANOVA unidireccional (ANOVA de un solo factor o ANOVA simple)

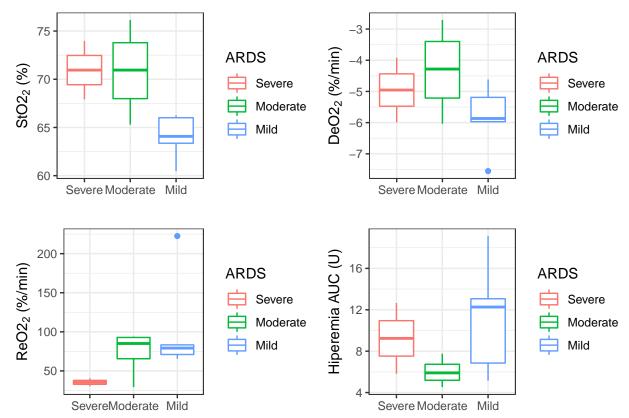
Dado que el número de observaciones por grupo no es constante, se trata de un modelo no equilibrado. Es importante tenerlo en cuenta cuando se comprueben las condiciones de normalidad y homocedasticidad.

```
#ARDS , TObslTSI, TOslope1, TOslope2, TOAUC
a<-ggplot(data = na.omit(datos_pac), aes(x = ARDS, y = TObslTSI, color = ARDS)) +
    geom_boxplot() +
   ylab(expression(paste("St02"[2], " (%)")))+
   xlab("")+
   theme_bw()
b<-ggplot(data = na.omit(datos_pac), aes(x = ARDS, y = TOslope1, color = ARDS)) +
    geom_boxplot() +
   ylab(expression(paste("De02"[2], " (%/min)")))+
   xlab("")+
   theme_bw()
c<-ggplot(data = na.omit(datos_pac), aes(x = ARDS, y = TOslope2, color = ARDS))+</pre>
    geom_boxplot() +
   ylab(expression(paste("ReO2"[2], " (%/min)")))+
   xlab("")+
   theme_bw()
```

```
d<-ggplot(data = na.omit(datos_pac), aes(x = ARDS, y = TOAUC, color = ARDS)) +
    geom_boxplot() +
    ylab(expression(paste("Hiperemia AUC", " (U)")))+
    xlab("")+
    theme_bw()

library(ggpubr)

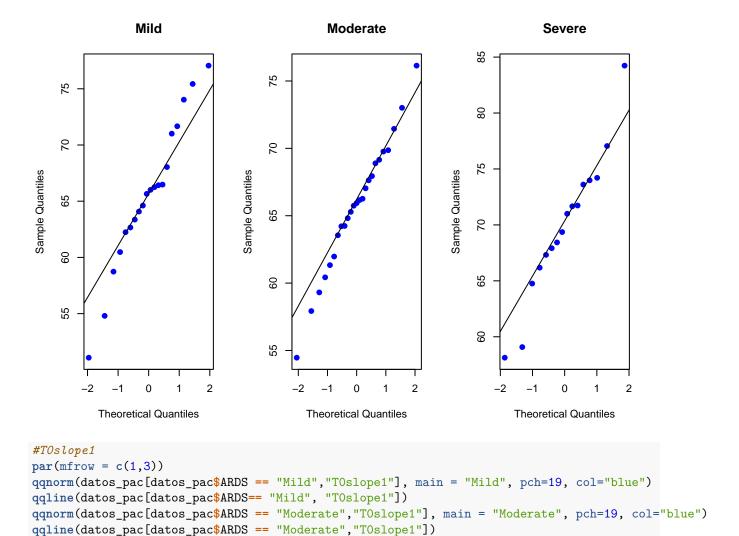
ggarrange(a, b, c, d, ncol = 2, nrow = 2)</pre>
```



verificar condiciones para un ANOVA

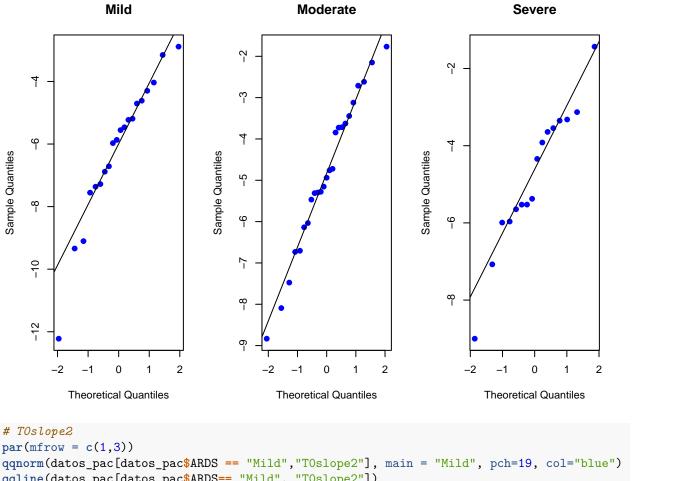
```
#ARDS ,TObslTSI, TOslope1, TOslope2, TOAUC

# TObslTSI
par(mfrow = c(1,3))
qqnorm(datos_pac[datos_pac$ARDS == "Mild","TObslTSI"], main = "Mild", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS== "Mild", "TObslTSI"])
qqnorm(datos_pac[datos_pac$ARDS == "Moderate","TObslTSI"], main = "Moderate", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS == "Moderate","TObslTSI"])
qqnorm(datos_pac[datos_pac$ARDS == "Severe","TObslTSI"], main = "Severe", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS== "Severe","TObslTSI"])
```

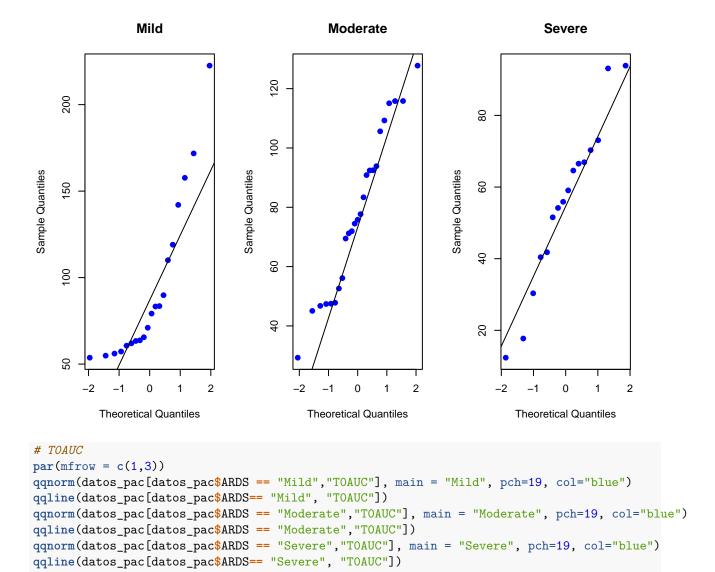


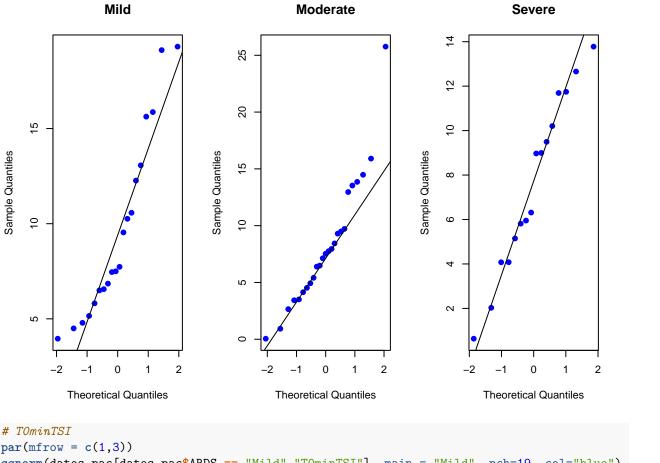
qqnorm(datos_pac(datos_pac(sarDS == "Severe", "TOslope1"], main = "Severe", pch=19, col="blue")

qqline(datos_pac[datos_pac\$ARDS== "Severe", "T0slope1"])

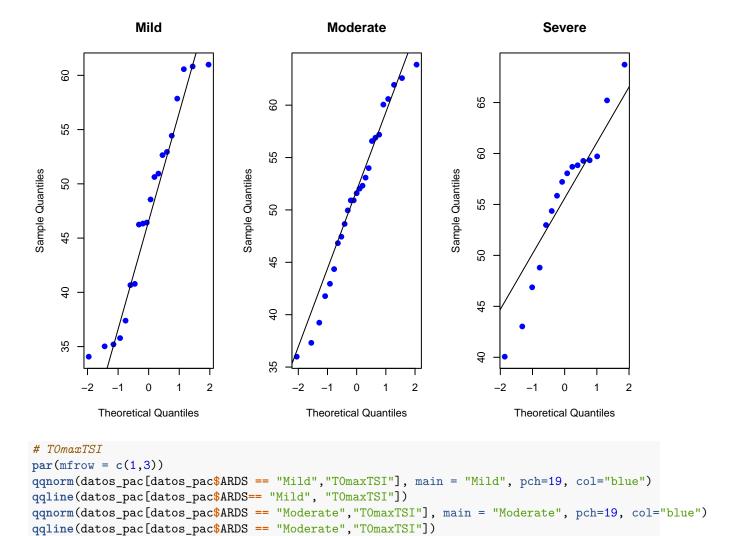


```
# TOslope2
par(mfrow = c(1,3))
qqnorm(datos_pac$ARDS == "Mild","TOslope2"], main = "Mild", pch=19, col="blue")
qqline(datos_pac$ARDS== "Mild", "TOslope2"])
qqnorm(datos_pac[datos_pac$ARDS == "Moderate","TOslope2"], main = "Moderate", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS == "Moderate","TOslope2"])
qqnorm(datos_pac[datos_pac$ARDS == "Severe","TOslope2"])
qqline(datos_pac[datos_pac$ARDS == "Severe","TOslope2"], main = "Severe", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS == "Severe","TOslope2"])
```



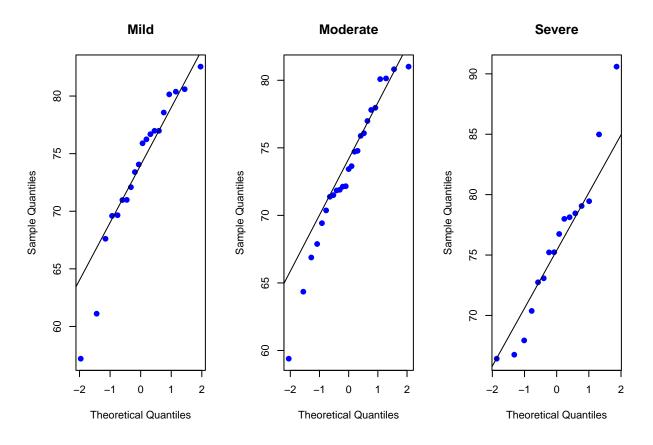


```
# TOminTSI
par(mfrow = c(1,3))
qqnorm(datos_pac[datos_pac$ARDS == "Mild", "TOminTSI"], main = "Mild", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS== "Mild", "TOminTSI"])
qqnorm(datos_pac[datos_pac$ARDS == "Moderate", "TOminTSI"], main = "Moderate", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS == "Moderate", "TOminTSI"])
qqnorm(datos_pac[datos_pac$ARDS == "Severe", "TOminTSI"], main = "Severe", pch=19, col="blue")
qqline(datos_pac[datos_pac$ARDS== "Severe", "TOminTSI"])
```



qqnorm(datos_pac\$ARDS == "Severe", "T0maxTSI"], main = "Severe", pch=19, col="blue")

qqline(datos_pac[datos_pac\$ARDS== "Severe", "T0maxTSI"])



Dado que los grupos tienen menos de 50 eventos se emplea el test de shapiro-wilk.

```
require(nortest)
#TObslTSI
sto2_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ lillie.test(x$T0bslTSI)})
dtest<-rbind(sto2_test$Mild$statistic, sto2_test$Moderate$statistic, sto2_test$Severe$statistic)
ptest<-rbind(sto2_test$Mild$p.value, sto2_test$Moderate$p.value, sto2_test$Severe$p.value)
sto2_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$TObslTSI)
dtest<-rbind(sto2_test$Mild$statistic, sto2_test$Moderate$statistic, sto2_test$Severe$statistic)
ptest<-rbind(sto2_test$Mild$p.value, sto2_test$Moderate$p.value, sto2_test$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
\#colnames(test1) < -c("Estadístic", "P-valor")
#TOslope1
slope1_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$TOslope
dtest<-rbind(slope1_test$Mild$statistic, slope1_test$Moderate$statistic, slope1_test$Severe$statistic)
ptest<-rbind(slope1_test$Mild$p.value, slope1_test$Moderate$p.value, slope1_test$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
```

```
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
slope1_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ lillie.test(x$T0slope1)})</pre>
dtest<-rbind(slope1_test$Mild$statistic, slope1_test$Moderate$statistic, slope1_test$Severe$statistic)
ptest<-rbind(slope1_test$Mild$p.value, slope1_test$Moderate$p.value, slope1_test$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
#TOslope2
slope2_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$TOslope
dtest<-rbind(slope2_test$Mild$statistic, slope2_test$Moderate$statistic, slope2_test$Severe$statistic)
ptest<-rbind(slope2_test$Mild$p.value, slope2_test$Moderate$p.value, slope2_test$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
#TOAUC
auc_test<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$TOAUC)})</pre>
dtest<-rbind(auc_test$Mild$statistic, auc_test$Moderate$statistic, auc_test$Severe$statistic)
ptest<-rbind(auc_test$Mild$p.value, auc_test$Moderate$p.value, auc_test$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
#TOminTSI
sto2_min<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$TOminTSI)}
dtest<-rbind(sto2_min$Mild$statistic, sto2_min$Moderate$statistic, sto2_min$Severe$statistic)
ptest<-rbind(sto2_min$Mild$p.value, sto2_min$Moderate$p.value, sto2_min$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
#TOmaxTSI
st02_max<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ shapiro.test(x = x$T0maxTSI)}
dtest<-rbind(st02_max$Mild$statistic, st02_max$Moderate$statistic, st02_max$Severe$statistic)
ptest<-rbind(st02_max$Mild$p.value, st02_max$Moderate$p.value, st02_max$Severe$p.value)
test1<-cbind(dtest, ptest)</pre>
#rownames(test1)<-c("Mild", "Moderate", "Severe")</pre>
```

```
#colnames(test1)<-c("Estadístic", "P-valor")</pre>
```

Los test de hipótesis no muestran evidencias de falta de normalidad.

Varianza constante entre grupos (homocedasticidad):

- F-test (razón de varianzas): El F-test, también conocido como contraste de la razón de varianzas, contrasta la hipótesis nula de que dos poblaciones normales tienen la misma varianza. Es muy potente, detecta diferencias muy sutiles
- Test de Bartlett :Permite contrastar la igualdad de varianza en 2 o más poblaciones sin necesidad de que el tamaño de los grupos sea el mismo.
- Fligner-Killeen (median) test

```
#TObslTSI
vartest_param1<-fligner.test(TObslTSI ~ ARDS, datos_pac)</pre>
vartest1<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest1)<-c("Estadístic", "P-valor")</pre>
#TOslope1
vartest_param1<-fligner.test(TOslope1 ~ ARDS, datos_pac)</pre>
vartest2<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest2)<-c("Estadístic", "P-valor")</pre>
#TOslope2
vartest_param1<-fligner.test(TOslope2 ~ ARDS, datos_pac)</pre>
vartest3<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest3)<-c("Estadístic", "P-valor")</pre>
#TOAUC
vartest_param1<-fligner.test(TOAUC ~ ARDS, datos_pac)</pre>
vartest4<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest4)<-c("Estadístic", "P-valor")</pre>
#TOminTSI
vartest_param1<-fligner.test(TOminTSI ~ ARDS, datos_pac)</pre>
vartest4<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest4)<-c("Estadístic", "P-valor")</pre>
#TOmaxTSI
vartest_param1<-fligner.test(TOmaxTSI ~ ARDS, datos_pac)</pre>
vartest4<-cbind(vartest_param1$statistic, vartest_param1$p.value)</pre>
colnames(vartest4)<-c("Estadístic", "P-valor")</pre>
```

El test no encuentra diferencias significativas entre las varianzas de los dos grupos. No hay evidencias significativas de falta de homocedasticidad en ninguno de los dos test.

El estudio de las condiciones puede realizarse previo cálculo del ANOVA, puesto que si no se cumplen no tiene mucho sentido seguir adelante. Sin embargo la forma más adecuada de comprobar que se satisfacen las condiciones necesarias es estudiando los residuos del modelo una vez generado el ANOVA. R permite graficar los residuos de forma directa con la función plot(objeto anova).

Análisis de varianza ANOVA

```
library(xtable)
#TObslTSI
anova <- aov(datos_pac$TObslTSI ~ datos_pac$ARDS)
summary(anova)</pre>
```

Df Sum Sq Mean Sq F value Pr(>F)

datos_pac\$ARDS 2 218.8 109.40 3.197 0.0482 * Residuals 58 1985.1 34.23

— Signif. codes: 0 '' 0.001 " 0.01 " 0.05 " 0.1" ' 1 12 observations deleted due to missingness

```
#TOslope1
```

```
anova <- aov(datos_pac$TOslope1 ~ datos_pac$ARDS)
xtable(summary(anova))</pre>
```

% latex table generated in R 4.1.2 by xtable 1.8-4 package % Fri Dec 24 16:57:10 2021

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
datos_pac\$ARDS	2	23.86	11.93	3.05	0.0550
Residuals	58	226.85	3.91		

#TOslope1

```
anova <- aov(datos_pac$TOslope2 ~ datos_pac$ARDS)
xtable(summary(anova))</pre>
```

% latex table generated in R 4.1.2 by x table 1.8-4 package % Fri Dec 24 16:57:10 2021

	Df	$\operatorname{Sum} \operatorname{Sq}$	Mean Sq	F value	Pr(>F)
datos_pac\$ARDS	2	12621.16	6310.58	5.42	0.0069
Residuals	58	67487.84	1163.58		

#TOAUC

```
anova <- aov(datos_pac$TOAUC ~ datos_pac$ARDS)
xtable(summary(anova))</pre>
```

% latex table generated in R 4.1.2 by xtable 1.8-4 package % Fri Dec 24 16:57:10 2021

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
datos_pac\$ARDS	2	39.67	19.84	0.81	0.4477
Residuals	58	1411.71	24.34		

#TOminTSI

```
anova <- aov(datos_pac$T0minTSI ~ datos_pac$ARDS)
xtable(summary(anova))</pre>
```

% latex table generated in R 4.1.2 by x table 1.8-4 package % Fri Dec 24 16:57:10 2021

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
datos_pac\$ARDS	2	572.39	286.20	4.20	0.0197
Residuals	58	3949.61	68.10		

```
#TOmaxTSI
anova <- aov(datos_pac$TOmaxTSI ~ datos_pac$ARDS)
xtable(summary(anova))</pre>
```

% latex table generated in R 4.1.2 by xtable 1.8-4 package % Fri Dec 24 16:57:10 2021

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
datos_pac\$ARDS	2	68.59	34.30	0.95	0.3911
Residuals	58	2084.54	35.94		

Dado que el p-value es inferior a 0.05 no hay evidencias suficientes para considerar que las dos medias no son distintas. La representación gráfica de los residuos no muestra falta de homocedasticidad (gráfico 1) y en el que los residuos se distribuyen muy cercanos a la linea de la normal.

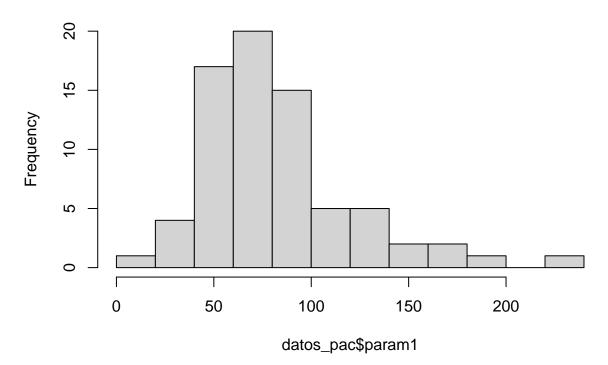
ANOVA para Param1

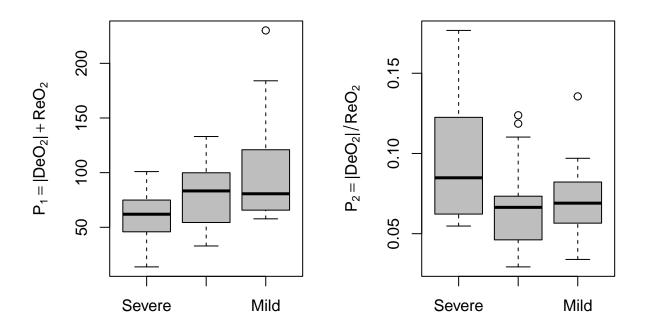
- Suma de los valores en absoluto de los pendientes
- Angulo "v"

Suma de los valores en absoluto de los pendientes

```
datos_pac$param1<-abs(datos_pac$T0slope1) + datos_pac$T0slope2
hist(datos_pac$param1)</pre>
```

Histogram of datos_pac\$param1





ANOVA unidireccional (ANOVA de un solo factor o ANOVA simple)

```
#label(datos_pac$param1)<-""
table1(~ param1 | ARDS, data = datos_pac)</pre>
```

Get nicer `table1` LaTeX output by simply installing the `kableExtra` package

	Severe	Moderate	Mild	Overall
	(N=16)	(N=25)	(N=20)	(N=73)
DeO2 (%/min) Mean (SD) Median [Min, Max]	60.5 (24.4) 62.0 [13.8, 101]	83.1 (28.1) 83.3 [32.9, 133]	99.5 (48.4) 80.7 [57.8, 230]	82.2 (38.4) 73.6 [13.8, 230]

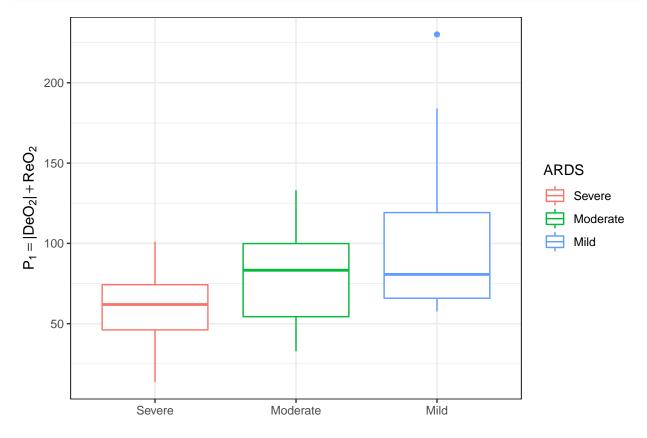
```
table1(~ param2 | ARDS, data = datos_pac)
```

Get nicer `table1` LaTeX output by simply installing the `kableExtra` package

	Severe	Moderate	Mild	Overall
DeO2 (%/min)	(N=16)	(N=25)	(N=20)	(N=73)
Mean (SD) Median [Min,	0.0957 (0.0369) 0.0848 [0.0548,	$0.0664 \ (0.0259)$ $0.0664 \ [0.0293,$	$0.0715 \ (0.0218)$ $0.0690 \ [0.0339,$	$0.0750 \ (0.0287)$ $0.0692 \ [0.0293,$
Max]	0.177]	0.124]	0.136]	0.177]

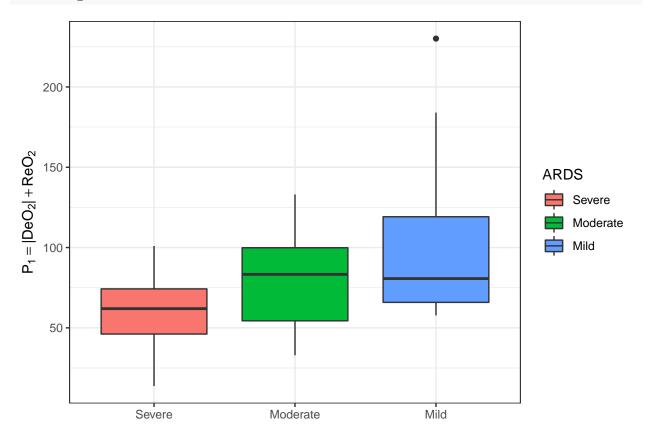
Dado que el número de observaciones por grupo no es constante, se trata de un modelo no equilibrado. Es importante tenerlo en cuenta cuando se comprueben las condiciones de normalidad y homocedasticidad.

```
midatos<-subset(datos_pac, select = c(ARDS, param1))
ggplot(data = na.omit(midatos), aes(x = ARDS, y = param1, color = ARDS)) +
    geom_boxplot() +
    ylab(expression("P"[1] == paste( "|", "De0"[2], "|") + "Re0"[2]))+
    xlab("")+
    theme_bw()</pre>
```

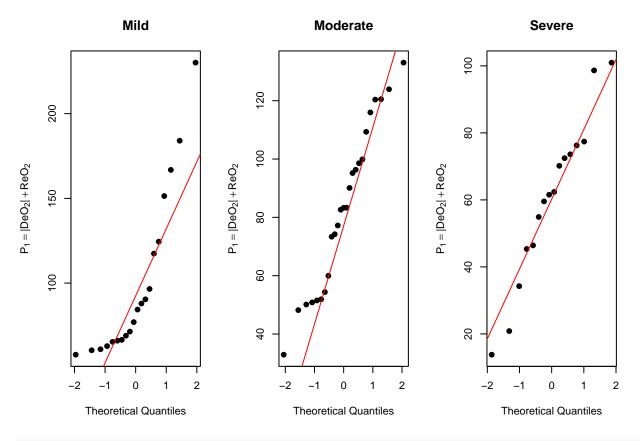


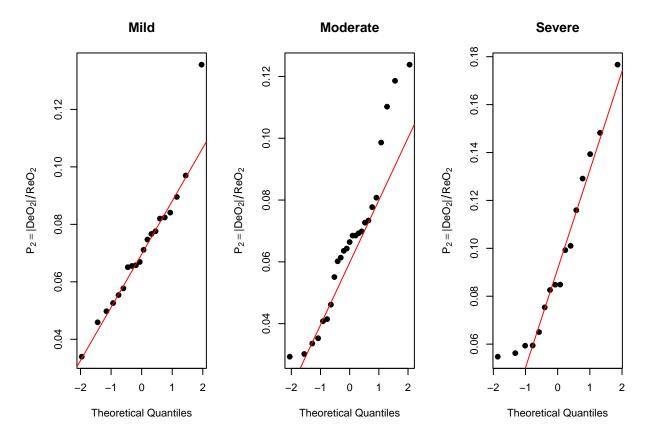
```
ggplot(data = na.omit(midatos), aes(x = ARDS, y = param1, fill = ARDS)) +
    geom_boxplot() +
    ylab(expression("P"[1] == paste( "|", "DeO"[2], "|") + "ReO"[2]))+
    xlab("")+
```

theme_bw()



verificar condiciones para un ANOVA





Dado que los grupos tienen mas de 50 eventos se emplea el test de Kolmogorov-Smirnov con la corrección de Lilliefors.

```
require(nortest)
test_param1<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ lillie.test(x$param1)})</pre>
\#dtest < -rbind(test\_param1\$Mild\$statistic, test\_param1\$Moderate\$statistic, \#test\_param1\$Severe\$statistic
ptest<-cbind(test_param1$Mild$p.value, test_param1$Moderate$p.value, test_param1$Severe$p.value)
#test1<-rbind(dtest, ptest)</pre>
test_param2<-by(data = datos_pac,INDICES = datos_pac$ARDS,FUN = function(x){ lillie.test(x$param2)})
\#dtest2 < -rbind(test\_param2\$Mild\$statistic,\ test\_param2\$Moderate\$statistic,\ \#test\_param2\$Severe\$statistic\}
ptest2<-cbind(test_param2$Mild$p.value, test_param2$Moderate$p.value, test_param2$Severe$p.value)
test2<-rbind(ptest, ptest2)</pre>
#test_norm<-cbind(test1, test2)</pre>
colnames(test2)<-c("Mild", "Moderate", "Severe")</pre>
rownames(test2)<-c("P1", "P2")</pre>
test2
              Mild Moderate
                                 Severe
## P1 0.009407288 0.3864491 0.7812309
## P2 0.487168078 0.1339958 0.1883315
```

Los test de hipótesis muestran evidencias de falta de normalidad en los dos grupos.

Varianza constante entre grupos (homocedasticidad):

```
vartest_param1<-fligner.test(param1 ~ ARDS, datos_pac)
vartest_param2<-fligner.test(param2 ~ ARDS, datos_pac)

vartest<-cbind(vartest_param1$p.value, vartest_param2$p.value)
colnames(vartest)<-c("P1", "P2")

xtable(vartest)</pre>
```

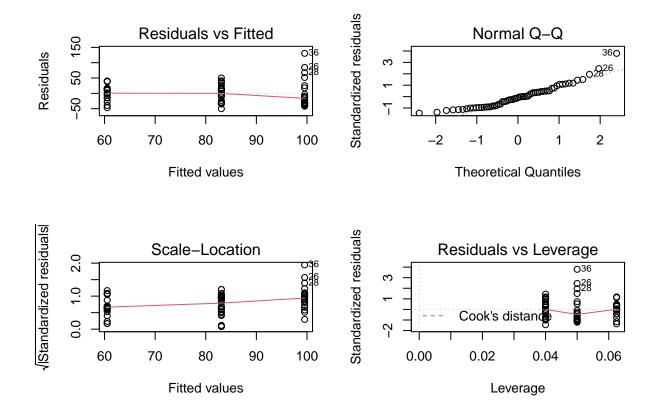
% latex table generated in R 4.1.2 by xtable 1.8-4 package % Fri Dec 24 16:57:11 2021

	P1	P2
1	0.43	0.19

El test no encuentra diferencias significativas entre las varianzas de los dos grupos. No hay evidencias significativas de falta de homocedasticidad en ninguno de los dos test.

Análisis de varianza ANOVA

```
anova <- aov(datos_pac$param1 ~ datos_pac$ARDS)</pre>
summary(anova)
##
                 Df Sum Sq Mean Sq F value Pr(>F)
## datos_pac$ARDS 2 13530
                              6765
                                     5.413 0.00699 **
## Residuals
                 58 72485
                              1250
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## 12 observations deleted due to missingness
anova2 <- aov(datos_pac$param2 ~ datos_pac$ARDS)</pre>
summary(anova2)
##
                 Df Sum Sq Mean Sq F value Pr(>F)
## datos_pac$ARDS 2 0.00896 0.004479
                                        5.71 0.00545 **
                58 0.04549 0.000784
## Residuals
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## 12 observations deleted due to missingness
par(mfrow=c(2,2))
plot(anova)
```



Dado que el p-value es inferior a 0.05 no hay evidencias suficientes para considerar que las dos medias son distintas. La representación gráfica de los residuos no muestra falta de homocedasticidad (gráfico 1) y en el que los residuos se distribuyen muy cercanos a la linea de la normal.

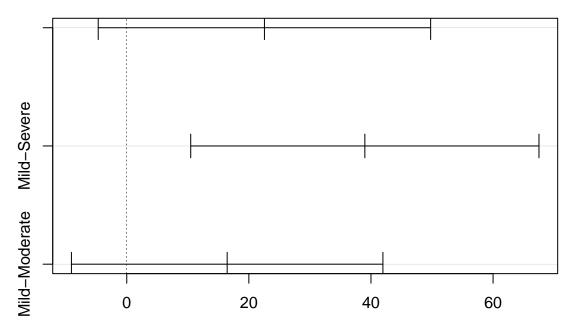
```
pairwise.t.test(x = datos_pac$param1, g = datos_pac$ARDS, p.adjust.method = "holm",
                pool.sd = TRUE, paired = FALSE, alternative = "two.sided")
##
##
    Pairwise comparisons using t tests with pooled SD
##
          datos_pac$param1 and datos_pac$ARDS
##
##
##
            Severe Moderate
## Moderate 0.1021 -
## Mild
            0.0051 0.1264
## P value adjustment method: holm
TukeyHSD (anova)
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = datos_pac$param1 ~ datos_pac$ARDS)
##
## $`datos_pac$ARDS`
##
                       diff
                                   lwr
                                            upr
                                                    p adj
## Moderate-Severe 22.55118 -4.672222 49.77459 0.1231286
```

```
## Mild-Severe 38.99809 10.477614 67.51857 0.0048050

## Mild-Moderate 16.44691 -9.062582 41.95640 0.2751808

plot(TukeyHSD(anova))
```

95% family-wise confidence level

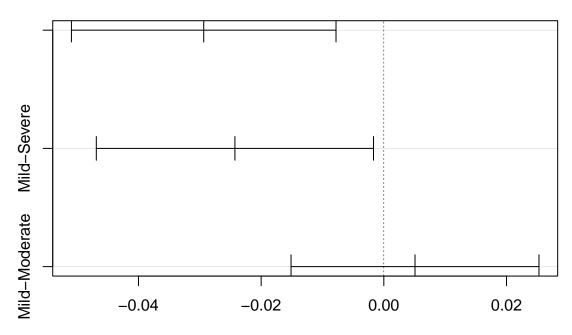


Differences in mean levels of datos_pac\$ARDS

TukeyHSD(anova2)

```
Tukey multiple comparisons of means
##
##
       95% family-wise confidence level
##
## Fit: aov(formula = datos_pac$param2 ~ datos_pac$ARDS)
##
## $`datos_pac$ARDS`
                          diff
                                        lwr
                                                     upr
                                                             p adj
## Moderate-Severe -0.02936198 -0.05092899 -0.007794972 0.0050100
## Mild-Severe
                   -0.02427805 -0.04687263 -0.001683472 0.0324514
## Mild-Moderate
                    0.00508393 -0.01512528 0.025293138 0.8179090
plot(TukeyHSD(anova2))
```

95% family-wise confidence level



Differences in mean levels of datos_pac\$ARDS

MODELOS de regressión

• Modelos con pendientes (DeO2 y ReO2)

```
##
                               Estimate Std. Error
                                                       t value
                                                                   Pr(>|t|)
## (Intercept)
                            5.996946653 1.339044339 4.4785273 4.782989e-05
## age_enrollment
                           -0.014530061 0.007274873 -1.9972941 5.160092e-02
## sexMasculino
                           -0.745179913 0.193658548 -3.8479061 3.582429e-04
## T0slope1
                           -0.110398224 0.045248721 -2.4398087 1.852216e-02
## T0slope2
                            0.004170609 0.002658827
                                                     1.5685899 1.234523e-01
## TOmaxTSI
                           -0.044613093 0.013385785 -3.3328709 1.682048e-03
## hypertensionNo
                            0.299149287 0.175657615
                                                    1.7030249 9.516869e-02
## previous_lung_diseaseNo
                            0.557675145 0.309500136 1.8018575 7.798373e-02
## smokerNo
                           -0.277966899 0.183395710 -1.5156674 1.363007e-01
## ischemic_heart_disease2 -0.036959838 0.297564210 -0.1242079 9.016803e-01
## atrial_fibrillation2
                           -0.634344245 0.344117443 -1.8433946 7.158195e-02
```

- Modelos con param1
- Modelos con param2

Resultados

% latex table generated in R 4.1.2 by xtable 1.8-4 package % Fri Dec 24 16:57:11 2021

	Modelos	R-cuadrado ajustado
1	Modelo 1	47.42
2	Modelo 2	48.63
3	Modelo 3	48.28
4	Modelo 4	49.47
5	Modelo 5	48.56
6	Modelo 6	49.65
7	Modelo 7	49.97
8	Modelo 8	50.83
9	Modelo 9	51.61
10	Modelo 10	47.04
11	Modelo 11	49.35
12	Modelo 12	45.14
13	Modelo 13	39.93
14	Modelo 14	51.14

Table 6: Resultados de los modelos que incluyen valores de las pendientes

% latex table generated in R 4.1.2 by x table 1.8-4 package % Fri Dec 24 16:57:11 2021

	M-1-1	D d d:4- d-
	Modelos	R-cuadrado ajustado
1	Modelo 1	48.66
2	Modelo 2	49.81
3	Modelo 3	49.38
4	Modelo 4	50.28
5	Modelo 5	49.38
6	Modelo 6	47.68
7	Modelo 7	47.91
8	Modelo 8	48.83
9	Modelo 9	47.46
10	Modelo 10	43.54
11	Modelo 11	44.26
12	Modelo 12	40.83
13	Modelo 13	39.93
14	Modelo 14	51.9

Table 7: Resultados de los modelos que incluyen param1

% latex table generated in R 4.1.2 by x table 1.8-4 package % Fri Dec 24 16:57:11 2021

Árbol de regresión

```
datos_pac$log_TSI<-log(datos_pac$T0bslTSI)
datos_pac$log_DeO2<-log(datos_pac$T0slope1)

## Warning in log(datos_pac$T0slope1): Se han producido NaNs
datos_pac$log_ReO2<-log(datos_pac$T0slope2)
datos_pac$log_AUC<-log(datos_pac$T0AUC)
datos_pac$log_edad<-log(datos_pac$age_enrollment)</pre>
```

1 Modelo 1 50.17 2 Modelo 2 51.33 3 Modelo 3 50.11 4 Modelo 4 50.42 5 Modelo 5 49.05 6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63		3.5.1.1	D 1 1 1 1 1
2 Modelo 2 51.33 3 Modelo 3 50.11 4 Modelo 4 50.42 5 Modelo 5 49.05 6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63		Modelos	R-cuadrado ajustado
3 Modelo 3 50.11 4 Modelo 4 50.42 5 Modelo 5 49.05 6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63	1	Modelo 1	50.17
4 Modelo 4 50.42 5 Modelo 5 49.05 6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63	2	Modelo 2	51.33
5 Modelo 5 49.05 6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63	3	Modelo 3	50.11
6 Modelo 6 39.37 7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63	4	Modelo 4	50.42
7 Modelo 7 39.85 8 Modelo 8 39.29 9 Modelo 9 31.63	5	Modelo 5	49.05
8 Modelo 8 39.29 9 Modelo 9 31.63	6	Modelo 6	39.37
9 Modelo 9 31.63	7	Modelo 7	39.85
0 0 00	8	Modelo 8	39.29
	9	Modelo 9	31.63
10 Modelo 10 24.09	10	Modelo 10	24.09
11 Modelo 11 30.92	11	Modelo 11	30.92
12 Modelo 12 24.03	12	Modelo 12	24.03
13 Modelo 13 39.93	13	Modelo 13	39.93
14 Modelo 14 52.38	14	Modelo 14	52.38

Table 8: Resultados de los modelos que incluyen param2

```
datos_pac$log_param1<-log(datos_pac$param1)</pre>
datos pac$log param2<-log(datos pac$param2)</pre>
library(rpart)
library(rpart.plot)
library(table1)
datos_reg<-subset(datos_pac, select = c(age_enrollment, sex, bmi, T0bslTSI, T0bslTHb, T0slope1, T0slope
previous_lung_disease, smoker, renal_insufficiency, ARDS, SF, param1, param2))
datos_pac1<-subset(datos_pac, select = c(age_enrollment, sex, bmi, T0bslTSI, T0slope1,</pre>
                    TOslope2, TOminTSI, TOmaxTSI, TOAUC, ischemic_heart_disease,
                    atrial_fibrillation, hypertension, previous_heart_failure,
                   lung_disease, diabetes, previous_lung_disease, smoker,
                   renal_insufficiency, ARDS, param1, param2, log_TSI,
                   log_DeO2, log_ReO2, log_AUC, log_edad, log_param1, log_param2))
names(datos_pac1)<-c("Edad", "Sexo", "BMI", "TSI baseline", "De02", "Re02",</pre>
                     "TSI_minimo", "TSI_maximo", "AUC", "EIC", "FA", "HTA", "ICP", "EP",
                     "Diabetes", "EPP", "Fumador/a", "IR", "ARDS", "P1", "P2",
                     "log_TSI", "log_DeO2", "log_ReO2", "log_AUC", "log_edad", "log_param1",
                     "log param2")
datos_pac2<-subset(datos_pac, select = c(age_enrollment, sex, bmi, T0bslTSI, T0slope1,</pre>
                    TOslope2, TOminTSI, TOmaxTSI, TOAUC, ischemic_heart_disease,
                    atrial fibrillation, hypertension, previous heart failure,
                   lung_disease, diabetes, previous_lung_disease, smoker,
                   renal_insufficiency, SF, param1, param2, log_TSI,
                   log_DeO2, log_ReO2, log_AUC, log_edad, log_param1, log_param2))
names(datos_pac2)<-c("Edad", "Sexo", "BMI", "TSI baseline", "De02", "Re02",</pre>
                     "TSI_minimo", "TSI_maximo", "AUC", "EIC", "FA", "HTA", "ICP", "EP",
```

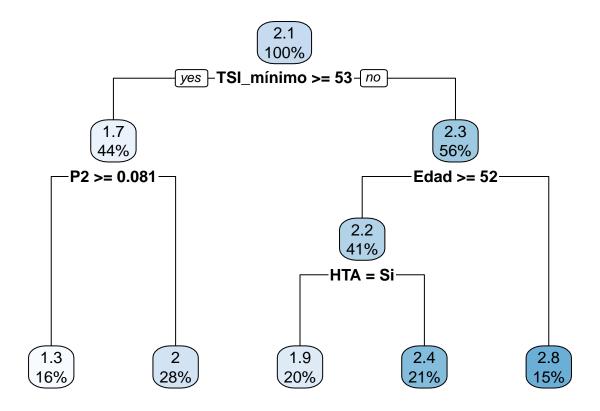
```
"Diabetes", "EPP", "Fumador/a", "IR", "SF","P1", "P2",

"log_TSI","log_DeO2","log_ReO2", "log_AUC","log_edad","log_param1",

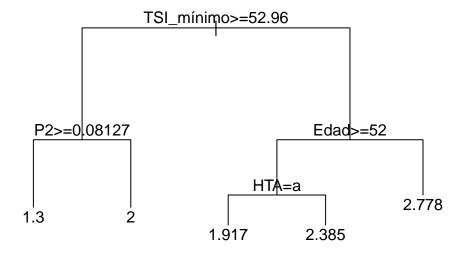
"log_param2")
```

Àrbol de clasificación

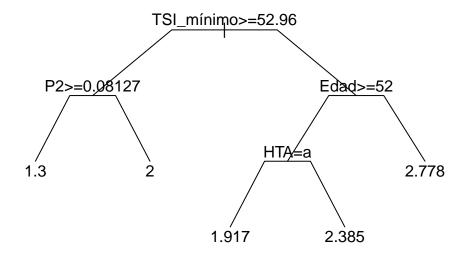
```
#arg_reg
arb_reg <- rpart(ARDS ~ ., data=datos_pac1, method = "anova")
rpart.plot(arb_reg)</pre>
```



plot(arb_reg,margin=0.1); text(arb_reg)

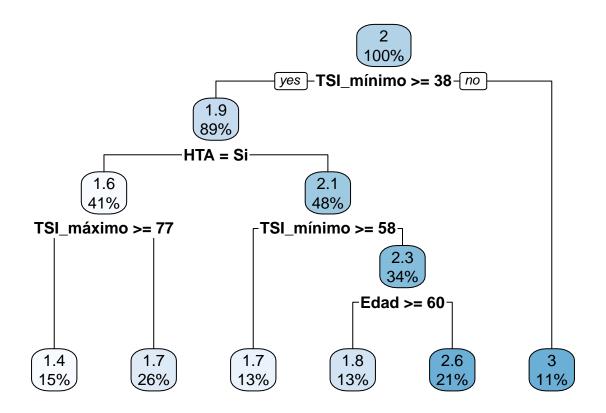


plot(arb_reg,compress=T,uniform=T,branch=0.4,margin=0.1); text(arb_reg)

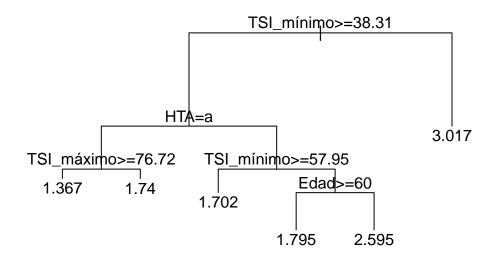


Árbol de regresión

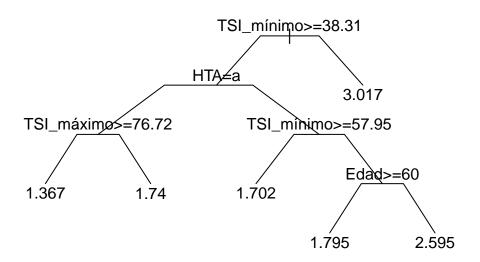
```
#mod_SF
mod1_sf <- rpart(SF ~ ., data=datos_pac2, method = "anova")
rpart.plot(mod1_sf)</pre>
```



plot(mod1_sf,margin=0.1); text(mod1_sf)



plot(mod1_sf,compress=T,uniform=T,branch=0.4,margin=0.1); text(mod1_sf)



Random Forest

##

##

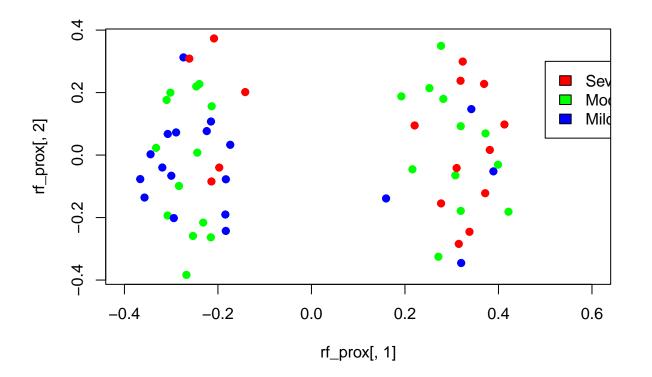
```
library(randomForest)
Categorico
datos2 <- subset(datos_pac, select = c(param2, T0minTSI, age_enrollment, hypertension,</pre>
                         ischemic_heart_disease, previous_lung_disease,
                         atrial_fibrillation, ARDS))
names(datos2)<-c("P2", "TSI_minimo", "Edad", "HTA", "EIC", "EPP", "FA", "ARDS")</pre>
set.seed(12345)
(modelo1_cat <- randomForest(ARDS ~ P2 + TSI_minimo + Edad + HTA</pre>
                              + EIC+EPP+FA,
                             data=na.omit(datos2),
                      ntree=300, mtry=1,
                       proximity=TRUE, importance=TRUE))
##
## Call:
    randomForest(formula = ARDS ~ P2 + TSI_minimo + Edad + HTA +
                                                                        EIC + EPP + FA, data = na.omit(da
##
                  Type of random forest: classification
##
```

Number of trees: 300

OOB estimate of error rate: 56.67%

No. of variables tried at each split: 1

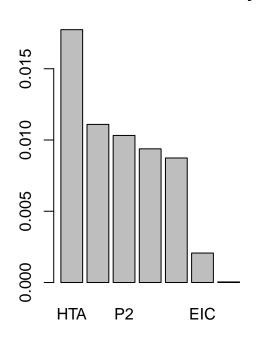
```
## Confusion matrix:
##
            Severe Moderate Mild class.error
                                    0.5625000
## Severe
                 7
                           5
## Moderate
                 3
                          11
                               10
                                    0.5416667
## Mild
                 3
                                    0.6000000
mod1_prox <- modelo1_cat$proximity</pre>
mode(mod1_prox); class(mod1_prox); dim(mod1_prox)
## [1] "numeric"
## [1] "matrix" "array"
## [1] 60 60
#proximity
library(MVA)
## Loading required package: HSAUR2
## Loading required package: tools
rf_prox <- cmdscale(1 - mod1_prox)</pre>
plot(rf_prox[,1],rf_prox[,2],
     col=rainbow(3)[as.integer(na.omit(datos2$ARDS))],pch=19,
     xlim = c(-0.4, 0.6))
legend(0.5,0.3,levels(datos2$ARDS),fill=rainbow(3))
```

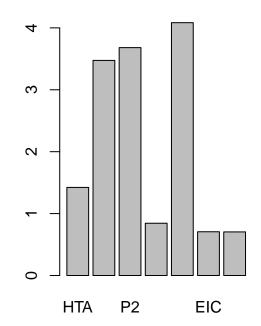


```
#importance
mod_imp<-modelo1_cat$importance[order(modelo1_cat$importance[,"MeanDecreaseAccuracy"], decreasing = T),</pre>
round(mod imp,2)
              Severe Moderate Mild MeanDecreaseAccuracy MeanDecreaseGini
##
## HTA
                0.05
                         0.00 0.02
                                                     0.02
                         0.00 0.02
## Edad
                0.01
                                                     0.01
                                                                       3.47
## P2
                0.02
                         0.02 -0.01
                                                     0.01
                                                                       3.68
## FA
                0.01
                         0.01 0.00
                                                     0.01
                                                                       0.84
## TSI_minimo
                0.03
                         0.00 0.01
                                                     0.01
                                                                       4.08
                         0.00 0.00
## EIC
                0.01
                                                     0.00
                                                                       0.71
                                                                      0.70
## EPP
                0.01
                         0.00 0.00
                                                     0.00
par(mfrow=c(1,2))
for(i in 4:5) {
  barplot(mod_imp[,i],main=colnames(mod_imp)[i])
```

MeanDecreaseAccuracy

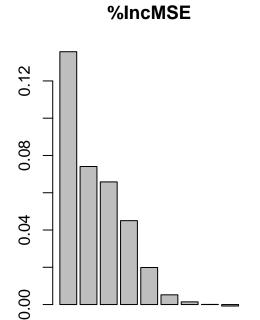
MeanDecreaseGini





$Cont{\'inuo}$

```
set.seed(12345)
(modelo1_sf <- randomForest(SF ~ P1+TSI_min + TSI_max+ Edad + HTA+</pre>
                              Sexo+ EIC + EPP + FA,
                              data=na.omit(datos3),
                      ntree=500, mtry=4,
                      proximity=TRUE, importance=TRUE))
##
## Call:
## randomForest(formula = SF ~ P1 + TSI_min + TSI_max + Edad + HTA + Sexo + EIC + EPP + FA, data
##
                  Type of random forest: regression
                        Number of trees: 500
\#\# No. of variables tried at each split: 4
##
##
             Mean of squared residuals: 0.4029008
                       % Var explained: 29.22
##
#importance
sf_imp<-modelo1_sf$importance[order(modelo1_sf$importance[,"%IncMSE"], decreasing = T),]
round(sf_imp,2)
##
           %IncMSE IncNodePurity
## P1
              0.14
## TSI_min
              0.07
                            7.19
## Sexo
              0.07
                            3.02
## HTA
              0.05
                            1.89
## Edad
              0.02
                            4.84
                            3.86
## TSI_máx
              0.01
## EPP
              0.00
                            0.18
## FA
              0.00
                            0.15
## EIC
              0.00
                            0.09
par(mfrow=c(1,2))
for(i in 1:2) {
  barplot(sf_imp[,i],main=colnames(sf_imp)[i])
```



HTA

EPP

P1

IncNodePurity

```
P1 HTA EPP
```

