Anexo - Código de R

Tesina: Rendimiento pronóstico de subrogados clínicos para estimar la acumulación de tejido adiposo visceral en la predicción de mortalidad por causas cardiovasculares en participantes de la Cohorte de la Ciudad de México

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Bibliotecas de Datos

```
library(dplyr)
library(tidyverse)
library(ggthemes)
library(ggpubr)
library(readr)
library(survival)
library(jtools)
library(gtsummary)
library(data.table)
library(epiR)
library(riskRegression)
library(prodlim)
library(survival)
library(cmprsk)
library(lava)
library(ggplot2)
library(MASS)
library(treemapify)
library("maxstat")
library("survival")
library(survminer)
library("survival")
library(rms)
library(pec)
```

Bibliotecas de Datos

```
base <- read_csv("~/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONALAUTÓNOMADEMÉXICO/PROY
base.mort <- read_csv("/Users/nefoantonio/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONAL
base.mrn <- read_csv("/Users/nefoantonio/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONAL
base.rsv <- read_csv("/Users/nefoantonio/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONAL
base.edu <- read_csv("~/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONALAUTÓNOMADEMÉXICO/
base.ids <- read_csv("~/Library/CloudStorage/OneDrive-UNIVERSIDADNACIONALAUTÓNOMADEMÉXICO/
base <- base %>%
    left_join(base.mort, by="PATID") %>%
    left_join(base.mrn, by="PATID") %>%
    left_join(base.rsv, by="PATID") %>%
    left_join(base.edu, by="PATID") %>%
    left_join(base.edu, by="PATID") %>%
    left_join(base.ids, by="PATID") %>%
```

Recodificación de Variables

```
#Mean SBP
base$PAS_PROM<-rowMeans(as.matrix(base%>%dplyr::select(SBP1,SBP2,SBP3)))
#Mean DBP
base$PAD PROM<-rowMeans(as.matrix(base%%dplyr::select(DBP1,DBP2,DBP3)))</pre>
#Diabetes
base$DIABETES_FINAL<-NULL
base$DIABETES_FINAL[base$BASE_DIABETES==1 |
                      base$DRUG_D1 == 1 |
                      base$DRUG_D2 == 1 |
                      base$DRUG_D3 == 1 |
                      base$DRUG_D4 == 1 |
                      baseBASE_HBA1C >= 6.5 < -1
base$DIABETES_FINAL<-na.tools::na.replace(base$DIABETES_FINAL,0)
#Cardiovascular Disease
base$CVD_BASAL<-NULL
base$CVD_BASAL[base$BASE_HEARTATTACK==1 |
                 base$BASE_ANGINA==1 |
                 base$BASE_STROKE==1]<-1
base$CVD_BASAL<-na.tools::na.replace(base$CVD_BASAL,0)
```

```
#Hypertension Definition
base$HAS_FINAL<-NULL
base$HAS_FINAL[base$BASE_HYPERTENSION==1 |
                                               base$DRUG_A1 == 1 |
                                               base$DRUG A2 == 1 |
                                               base$DRUG_A3 == 1 |
                                               base$DRUG_A4 == 1 |
                                               base$DRUG_A5 == 1 |
                                               base\$DRUG A6 == 1 |
                                               base$DRUG_A7 == 1 |
                                               base$DRUG_A8 == 1 |
                                               base$DRUG_A9 == 1 |
                                               base$DRUG_A10 == 1 |
                                               base$DRUG_A11 == 1 |
                                               base$PAS_PROM >= 140 \mid base$PAD_PROM >= 90] < -1
base$HAS_FINAL<-na.tools::na.replace(base$HAS_FINAL,0)</pre>
#CVD DEATHS
base$VASC_NON_CARD<-NULL
base$VASC_NON_CARD[base$D011==1]<-1
base$VASC_NON_CARD[base$D012==1]<-1
base$VASC_NON_CARD[is.na(base$VASC_NON_CARD)]<-0
## Numeric Variables
base$IMC<-base$WEIGHT/((base$HEIGHT/100)^2)</pre>
base$ICE<-base$WAISTC/base$HEIGHT</pre>
#Transform Laboratories to mg/dl
base$Glc_mgdl<-base$Glc*18;base$Glc_mgdl[base$Glc_mgdl<=0]<-NA
base\$Serum\_TG\_mgd1 < -base\$Serum\_TG*88.57; base\$Serum\_TG\_mgd1 \\ [base\$Serum\_TG\_mgd1 < = 0] < -NASSERUM\_TG\_mgd1 \\ [base$Serum\_TG_mgd1] \\ [base Serum\_TG_mgd1] \\ [base$Serum\_TG_mgd1] \\ [base Serum\_TG_mgd1] \\
## Log Transform Income
base$INCOME_LN<-log(base$INCOME+1)</pre>
#Education Recode
base$EDUGP_2<-NULL
base$EDUGP_2[base$EDUGP==1]<-1
base$EDUGP_2[base$EDUGP==2]<-1
```

```
base$EDUGP_2[base$EDUGP==3]<-1
base$EDUGP_2[base$EDUGP==4]<-2
base$EDUGP_2[base$EDUGP==5] < -2
base$EDUGP_2[base$EDUGP==6]<-2
base$EDUGP_2[base$EDUGP==7]<-3
base$EDUGP_2[base$EDUGP==8]<-3
base$EDUGP_2[base$EDUGP==9]<-3
base$EDUGP_2[base$EDUGP==10]<-4
base$EDUGP_2[base$EDUGP==11]<-4
base$EDUGP_2[base$EDUGP==12]<-4
base$EDUGP_2[base$EDUGP==13]<-4
base$EDAD_CAT<-NULL
base$EDAD_CAT[base$AGE<45]<-1
base$EDAD_CAT[base$AGE>=45 & base$AGE<65]<-2
baseEDAD_CAT[base\\AGE>=65]<-3
## Transform selected numeric to factors
base$EDAD_CAT<-factor(base$EDAD_CAT,labels = c("<45","45-65",">65"))
base$COYOACAN<- factor(base$COYOACAN,levels = c(0,1),labels = c("Iztapalapa","Coyoacan"))</pre>
base$EDU_LEVEL<- factor(base$EDU_LEVEL,levels = c(1:4),labels = c("University/College","Hi
base$EDUGP<- factor(base$EDUGP,levels = c(1:13),labels = c("Illiterate",</pre>
                                                             "Knows how to read",
                                                             "Knows how to read and write",
                                                             "Incomplete elementary",
                                                             "Complete elementary",
                                                             "Tecnical Studies with complete
                                                             "Incomplete high School",
                                                             "Complete high School",
                                                             "Tecnical Studies with complete
                                                             "Collegue",
                                                             "Tecnical Studies with complete
                                                             "Incomplete univesity",
                                                             "Complete univesity"))
base $EDUGP_2 <- factor (base $EDUGP_2, labels = c("Illiterate or Non-proper education",
                                               "Elementary",
```

```
"High School",
                                                 "Collegue"))
## Ocupation Categories
base$OCCUPATION_REC<-NULL
base$OCCUPATION_REC[base$OCCUPATION==10]<-1</pre>
base$OCCUPATION_REC[base$OCCUPATION==11]<-1</pre>
base$OCCUPATION_REC[base$OCCUPATION==1]<-1</pre>
base$OCCUPATION_REC[base$OCCUPATION==13]<-1
base$OCCUPATION_REC[base$OCCUPATION==14]<-1</pre>
base$OCCUPATION_REC[base$OCCUPATION==2]<-2</pre>
base$OCCUPATION_REC[base$OCCUPATION==3]<-2</pre>
base$OCCUPATION_REC[base$OCCUPATION==4]<-2
base$OCCUPATION_REC[base$OCCUPATION==6]<-2
base$OCCUPATION_REC[base$OCCUPATION==16]<-2
base$OCCUPATION_REC[base$OCCUPATION==19]<-2
base$OCCUPATION_REC[base$OCCUPATION==17]<-2</pre>
base$OCCUPATION_REC[base$OCCUPATION==18]<-2</pre>
base$OCCUPATION_REC[base$OCCUPATION==12]<-2</pre>
base$OCCUPATION_REC[base$OCCUPATION==7]<-3
base$OCCUPATION_REC[base$OCCUPATION==8]<-3</pre>
base$OCCUPATION_REC[base$OCCUPATION==5]<-3
base$OCCUPATION_REC[base$OCCUPATION==20]<-3
base$OCCUPATION_REC[base$OCCUPATION==15]<-4
base$OCCUPATION_REC[base$OCCUPATION==21]<-4
base $OCCUPATION_REC <- factor (base $OCCUPATION_REC, levels = c(1:4), labels = c("Private Employ
                                                                                "Blue-Collar Wo
                                                                                "Public Sector
                                                                                "Retired or Une
#Health Provider
base$HEALTH_PROVIDER_2<-NULL
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==1]<-1
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==2]<-1
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==3]<-1
```

base\$HEALTH_PROVIDER_2[base\$HEALTH_PROVIDER==4]<-1

```
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==5]<-1
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==6]<-1
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==7]<-1
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==8]<-2
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==9]<-2
base$HEALTH_PROVIDER_2[base$HEALTH_PROVIDER==10]<-3
base$HEALTH_PROVIDER_2[is.na(base$HEALTH_PROVIDER)]<-4
base$HEALTH_PROVIDER_2<-factor(base$HEALTH_PROVIDER_2, labels = c("Public HC", "Private HC",
#Alcohol Intake
base$ALCGP<-factor(base$ALCGP,labels = c("Never","Former",">3 times a month", ">2 times a
#Physical Activity
base$PHYSGP<-factor(base$PHYSGP,labels = c("None", ">2 times a week",">3 times a week"))
#Visceral Adiposiy Metrics
#METS_IR
base$METS_IR<-((log((2*base$Glc_mgdl)+base$Serum_TG_mgdl)*base$IMC))/(log(base$HDL_C_mgdl)
#METS VF
base METSVF < -(4.466+0.011*(log(base METS_IR)^3) + 3.239*(log(base ICE)^3) + 0.319*(base MALE) + 0.011*(log(base MALE)^3) + 0.01*(log(base MALE)^4) + 0.01*(log(base MALE)^4) + 0.01*(log(base MALE)^4) + 0.01*(log(base MALE)
                                                     +0.594*(log(base$AGE)))
#Grasa Visceral
base$VAT_METS<-exp(base$METSVF)</pre>
#Indices
base<-base %>%
       mutate(DAAT_index = if_else(MALE >= 1,
                                                                                                                 (-382.9+(1.09*base$WEIGHT)+(6.04*base$WAISTC)+(-2.29*base$IMEIGHT)+(-382.9+(1.09*base$WEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGHT)+(-2.29*base$IMEIGH
                                                                                                                 (-278+(-0.86*base\$WEIGHT)+(5.19*base\$WAISTC))),
                                  Depres_index = if_else(MALE >= 1,
                                                                                                                         (-225.39 + (2.125*base$AGE) + (2.843*base$WAISTC)),
                                 VAI_index = if_else(MALE >= 1,
                                                                                                              ((base$WAISTC/(39.68+(1.88*base$IMC)))*(base$Serum_TG/1.03)*(
                                                                                                              ((base$WAISTC/(36.58+(1.89*base$IMC)))*(base$Serum_TG/0.81)*(
                                 VAI_GEA_index = if_else(MALE >= 1,
                                                                                                                             ((base$WAISTC/(22.79+(2.68*base$IMC)))*(base$Serum_TG/1.3
```

Etiqueta de Variables

```
setattr(base$AGE, "label", "Edad, (Años)")
setattr(base$MALE, "label", "Sexo, (%)")
setattr(base$COYOACAN, "label", "Municipio, (%)")
setattr(base$EDU_LEVEL, "label", "Educación, (%)")
setattr(base$INCOME, "label", "Ingreso, (pesos/month)")
setattr(base$HEALTH_PROVIDER_2, "label", "Provedor de Servicios de Salud, (%)")
setattr(base$OCCUPATION_REC, "label", "Ocupación, (%)")
setattr(base$EVER_SMOK, "label", "Tabaquismo, (%)")
setattr(base$ALCGP, "label", "Habitos de Consumo de Alcohol, (%)")
setattr(base$PHYSGP, "label", "Actividad Física, (%)")
setattr(base$IMC, "label", "Indice de Masa Corporal, (kg/m2)")
setattr(base$ICE, "label", "Waist-to-Height Ratio, (%)")
setattr(base$HIPC, "label", "Hip Circunference, (cm)")
setattr(base$PAS_PROM, "label", "PAS, (mmHg)")
setattr(base$PAD_PROM, "label", "TAD, (mmHg)")
setattr(base$BASE_HBA1C, "label", "HbA1c, (%)")
setattr(base$Glc_mgdl, "label", "Glucose, (mg/dl)")
setattr(base$Serum_TG_mgdl, "label", "Trigliceridos, (mg/dl)")
setattr(base$HDL_C_mgdl, "label", "HDL-C, (mg/dl)")
setattr(base$LDL_C_mgdl, "label", "LDL-C, (mg/dl)")
```

Etiqueta de Variables

```
base.2<-base%>%
  dplyr::filter(DIABETES_FINAL!=1)%>%
  dplyr::filter(CVD_BASAL!=1)%>%
  dplyr::filter(CVD_BASAL!=1)%>%
  dplyr::filter(AGE<80)%>%
```

```
dplyr::filter(IMC<40)%>%
    dplyr::filter(IMC>=18.5)%>%
    dplyr::mutate(Depres.meanval = mean(Depres_index,na.rm = T), Depres.stdev = sd(Depres_index)
                  DAAT.meanval = mean(DAAT_index,na.rm = T), DAAT.stdev = sd(DAAT_index,na.rm
                  LAAP.meanval = mean(LAAP_index,na.rm = T), LAAP.stdev = sd(LAAP_index,na.r
                  EVA.meanval = mean(EVA_index,na.rm = T), EVA.stdev = sd(EVA_index,na.rm =
                  METS.meanval = mean(METSVF,na.rm = T), METS.stdev = sd(METSVF,na.rm = T),
                  DAI.meanval = mean(VAI_GEA_index,na.rm = T), DAI.stdev = sd(VAI_GEA_index,
  nrow(base.2)
[1] 117739
  base.2<-base.2%>%
    dplyr::filter(!is.na(VAI_index))%>%
    dplyr::filter(!is.na(EVA_index))%>%
    dplyr::filter(!is.na(PERSON_YEARS))%>%
    dplyr::filter(STATUS!="U")%>%
    dplyr::filter(abs((DAAT_index-DAAT.meanval)/DAAT.stdev)<3)%>%
    dplyr::filter(abs((LAAP_index-LAAP.meanval)/LAAP.stdev)<3)%>%
    dplyr::filter(abs((EVA_index-EVA.meanval)/EVA.stdev)<3)%>%
    dplyr::filter(abs((METSVF-METS.meanval)/METS.stdev)<3)%>%
    dplyr::filter(abs((VAI_GEA_index-DAI.meanval)/DAI.stdev)<3)</pre>
```

[1] 29034

nrow(base.2)

Resultados: Capitulo 3 - Proceso de selección de participantes integrantes de la muestra de estudio

```
#Se corrobora que no se tengan datos perdidos en los estimadores de adiposidad
#visceral
sum(is.na(base.2$Depres_index))
```

[1] 18441

```
sum(is.na(base.2$DAAT_index))
[1] 0
  sum(is.na(base.2$LAAP_index))
[1] 0
  sum(is.na(base.2$VAI_index))
[1] 0
  sum(is.na(base.2$EVA_index))
[1] 0
  sum(is.na(base.2$METSVF))
[1] 0
  sum(is.na(base.2$VAI_GEA_index))
[1] 0
  sum(is.na(base.2$PERSON_YEARS))
[1] 0
```

Resultados: Capitulo 3 - Características descriptivas de la muestra de estudio

```
t.0<-base.2 %>%
  dplyr::select(EDAD_CAT,AGE,COYOACAN,EDUGP_2,
                INCOME,OCCUPATION_REC,HEALTH_PROVIDER_2,
                EVER_SMOK, ALCGP, PHYSGP,
                IMC, ICE, HIPC, PAS_PROM, PAD_PROM, BASE_HBA1C, Glc_mgdl, Serum_TG_mgdl, HDL_C_mgd
  tbl_summary(missing = "no")%>%
  bold_labels()%>%
  modify_spanning_header(all_stat_cols() ~ "**Overall Sample**")%>%
  modify_table_body(
    dplyr::mutate,
    label = ifelse(label == "N missing (% missing)",
                    "Unknown",
                    label))
t1.1<-base.2 %>%
  dplyr::filter(MALE==1)%>%
  dplyr::select(EDAD_CAT,AGE,COYOACAN,EDUGP_2,
                INCOME,OCCUPATION_REC,HEALTH_PROVIDER_2,
                EVER_SMOK, ALCGP, PHYSGP,
                IMC, ICE, HIPC, PAS_PROM, PAD_PROM, BASE_HBA1C, Glc_mgdl, Serum_TG_mgdl, HDL_C_mgd
  tbl_summary(by = EDAD_CAT,missing = "no")%>%
  bold_labels()%>%
  modify_spanning_header(all_stat_cols() ~ "**Overall Sample**")%>%
  modify_table_body(
    dplyr::mutate,
    label = ifelse(label == "N missing (% missing)",
                    "Unknown",
                    label))
t1.2<-base.2 %>%
  dplyr::filter(MALE!=1)%>%
  dplyr::select(EDAD_CAT,AGE,COYOACAN,EDUGP_2,
                INCOME,OCCUPATION_REC,HEALTH_PROVIDER_2,
                EVER_SMOK, ALCGP, PHYSGP,
                IMC, ICE, HIPC, PAS_PROM, PAD_PROM, BASE_HBA1C, Glc_mgdl, Serum_TG_mgdl, HDL_C_mgd
  tbl_summary(by = EDAD_CAT, missing = "no")%>%
  bold_labels()%>%
  modify_spanning_header(all_stat_cols() ~ "**Overall Sample**")%>%
  modify_table_body(
    dplyr::mutate,
```

```
label))
   tbl_merge(
     tbls = list(t.0, t1.1, t1.2),
     tab_spanner = c("Muestra Total (n=29,034)","**Hombres (n=10,593)**", "**Mujeres (n=18,44
                                                                           45-65,
                                          45-65,
                    N =
                              <45, N
                                           N =
                                                     >65, N
                                                               <45, N
                                                                            N =
                                                                                      >65, N
                   29,034
                              =4,314
                                           4,760
                                                     = 1,519
                                                               = 7,737
                                                                                      = 2,319
Characteristic
                                                                            8,385
EDAD_CAT
< 45
                    12,051
                    (42\%)
45-65
                    13,145
                    (45\%)
> 65
                    3,838
                    (13\%)
                                                                                      70 (67,
Edad, (Años)
                   47 (40,
                              39 (37,
                                          53 (48,
                                                     70 (67,
                                                                39 (37,
                                                                           52 (48,
                                            58)
                     57)
                                41)
                                                       74)
                                                                 42)
                                                                             57)
                                                                                        74)
Municipio,
(\%)
Iztapalapa
                                368
                                           409
                                                       162
                                                                 829
                                                                             847
                                                                                        283
                    2,898
                   (10.0\%)
                              (8.5\%)
                                          (8.6\%)
                                                     (11\%)
                                                                (11\%)
                                                                           (10\%)
                                                                                      (12\%)
                                           4,351
                                                                            7,538
                                                                                       2,036
Coyoacan
                    26,136
                               3,946
                                                      1,357
                                                                6,908
                                          (91\%)
                                                                                       (88\%)
                    (90\%)
                               (91\%)
                                                     (89\%)
                                                                (89\%)
                                                                           (90\%)
EDUGP_2
                                                                                        895
Illiterate or
                    3,481
                                88
                                           420
                                                       441
                                                                 317
                                                                            1,320
                    (12\%)
                              (2.1\%)
                                          (8.9\%)
                                                     (29\%)
                                                                (4.1\%)
                                                                           (16\%)
                                                                                       (39\%)
Non-proper
education
                               1,085
                                           2,349
                                                       794
                                                                3,039
                                                                            4,818
                                                                                       1,168
Elementary
                    13,253
                               (26\%)
                                          (50\%)
                                                     (53\%)
                                                                (40\%)
                                                                           (58\%)
                                                                                      (51\%)
                    (46\%)
High School
                                           1,047
                                                       163
                                                                 2,851
                                                                                        186
                    7,339
                               1,559
                                                                            1,533
                    (25\%)
                               (37\%)
                                          (22\%)
                                                     (11\%)
                                                                (37\%)
                                                                           (18\%)
                                                                                      (8.0\%)
Collegue
                    4,735
                               1,516
                                           889
                                                       113
                                                                1,472
                                                                             682
                                                                                        63
                    (16\%)
                               (36\%)
                                          (19\%)
                                                     (7.5\%)
                                                                (19\%)
                                                                           (8.2\%)
                                                                                      (2.7\%)
Ingreso, (pe-
                    1,000
                               2,500
                                           2,000
                                                      1,200
                                                                 0 (0,
                                                                            0 (0,
                                                                                       0 (0,
                                                                                       600)
sos/month)
                     (0,
                              (1,500,
                                          (1,200,
                                                      (500,
                                                                1,600)
                                                                           1,120)
                              4,000)
                                                     2,000)
                    2,354)
                                          3,500)
```

label = ifelse(label == "N missing (% missing)",

"Unknown",

Ocupación,

(%)

		45-65 ,			45-65 ,		
	N =	<45, N	N =	> 65 , N	<45, N	N =	>65, N
Characteristic	29,034	=4,314	4,760	= 1,519	= 7,737	$8,\!385$	= 2,319
Private	3,998	1,242	1,007	119	1,022	576	32
Employers and	(14%)	(29%)	(21%)	(7.8%)	(13%)	(6.9%)	(1.4%)
Professionals							
Blue-Collar	18,874	2,004	$2,\!151$	417	$5,\!541$	6,771	1,990
Workers	(65%)	(47%)	(45%)	(27%)	(72%)	(81%)	(86%)
Public Sector	3,738	925	922	111	1,089	661	30
Workers	(13%)	(21%)	(19%)	(7.3%)	(14%)	(7.9%)	(1.3%)
Retired or	2,390	137	677	871	72	368	265
Unemployed	(8.2%)	(3.2%)	(14%)	(57%)	(0.9%)	(4.4%)	(11%)
Provedor de							
Servicios de							
Salud, (%)							
Public HC	1,350	178	186	73	393	404	116
	(4.6%)	(4.1%)	(3.9%)	(4.8%)	(5.1%)	(4.8%)	(5.0%)
Private HC	407	70	75	11	127	86	38
	(1.4%)	(1.6%)	(1.6%)	(0.7%)	(1.6%)	(1.0%)	(1.6%)
Non-Specified	58	11	8~(0.2%)	2	15	20	2
	(0.2%)	(0.3%)		(0.1%)	(0.2%)	(0.2%)	(<0.1%)
Missing	27,219	4,055	4,491	1,433	7,202	7,875	2,163
	(94%)	(94%)	(94%)	(94%)	(93%)	(94%)	(93%)
Tabaquismo,	$15,\!353$	3,412	$3,\!836$	1,220	3,431	2,850	604
(%)	(53%)	(79%)	(81%)	(80%)	(44%)	(34%)	(26%)
Habitos de							
Consumo de							
Alcohol, (%)							
Never	4,679	204	157	78	1,643	1,890	707
	(16%)	(4.7%)	(3.3%)	(5.1%)	(21%)	(23%)	(31%)
Former	1,639	375	398	105	346	339	76
	(5.6%)	(8.7%)	(8.4%)	(6.9%)	(4.5%)	(4.0%)	(3.3%)
>3 times a	$7,\!483$	1,729	1,964	644	1,241	1,463	442
month	(26%)	(40%)	(41%)	(42%)	(16%)	(17%)	(19%)
>2 times a week	13,195	1,358	1,587	500	4,247	4,475	1,028
	(45%)	(31%)	(33%)	(33%)	(55%)	(53%)	(44%)
>3 times a week	2,026	647	651	190	259	215	64
	(7.0%)	(15%)	(14%)	(13%)	(3.3%)	(2.6%)	(2.8%)
Actividad	. ,	, ,	, ,	. ,	, ,	, ,	. ,
Física, (%)							
None	22,120	2,730	3,360	1,123	6,329	6,758	1,820
	(76%)	(63%)	(71%)	(74%)	(82%)	(81%)	(78%)

			45-65,			45-65,	
	N =	<45, N	N = 1	>65, N	<45, N	N =	>65, N
Characteristic	$29,\!034$	=4,314	4,760	= 1,519	= 7,737	8,385	= 2,319
>2 times a week	2,509	909	570	85	383	409	153
	(8.6%)	(21%)	(12%)	(5.6%)	(5.0%)	(4.9%)	(6.6%)
>3 times a week	4,398	675	829	309	1,023	1,216	346
	(15%)	(16%)	(17%)	(20%)	(13%)	(15%)	(15%)
Indice de	27.9	27.0	27.4	26.7	27.8	29.0	28.2
Masa	(25.3,	(24.8,	(25.2,	(24.4,	(25.2,	(26.3,	(25.5,
Corporal,	30.9)	29.6)	30.0)	29.1)	30.9)	32.1)	31.6)
$(\mathrm{kg/m2})$							
Waist-to-	0.58	0.56	0.58	0.59	0.57	0.61	0.64
Height Ratio,	(0.54,	(0.52,	(0.55,	(0.56,	(0.53,	(0.56,	(0.59,
(%)	0.63)	0.59)	0.62)	0.63)	0.62)	0.66)	0.69)
Hip Circun-	102 (97,	100 (96,	100 (96,	99 (95,	102 (97,	105 (99,	104 (98,
ference, (cm)	108)	104)	105)	104)	108)	111)	112)
PAS, (mmHg)	125	123 (117,	128 (120,	135	120	127 (119,	137
	(117,	130)	137)	(126,	(111,	137)	(127,
	135)			146)	127)		149)
TAD,	83 (77,	83 (78,	85 (80,	85 (80,	80 (73,	83 (79,	86 (80,
(mmHg)	90)	89)	91)	91)	85)	90)	92)
$\mathrm{HbA1c},(\%)$	5.35	5.26	5.35	5.45	5.17	5.35	5.54
	(5.08,	(4.99,	(5.17,	(5.17,	(4.99,	(5.17,	(5.26,
	5.54)	5.45)	5.63)	5.63)	5.35)	5.63)	5.72)
Glucose,	55 (45,	51 (41,	55 (46,	58 (49,	52 (43,	56 (47,	61 (50,
$(\mathrm{mg}/\mathrm{dl})$	65)	61)	67)	71)	61)	67)	73)
Trigliceridos,	128	134 (110,	138 (112,	125	117 (95,	130 (107,	127
$(\mathrm{mg}/\mathrm{dl})$	(104,	164)	166)	(102,	143)	157)	(105,
	156)			151)			154)
HDL-C,	40 (35,	37 (33,	37 (33,	37 (33,	41 (37,	42 (37,	43 (37,
$(\mathrm{mg/dl})$	45)	41)	41)	42)	46)	47)	48)
LDL-C,	50 (40,	50 (39,	50 (39,	48 (39,	48 (39,	52 (41,	52 (41,
(mg/dl)	60)	60)	60)	58)	57)	62)	62)

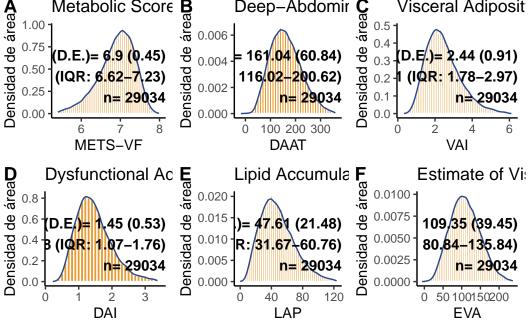
Resultados: Capitulo 3 - Estimación de subrogados clínicos de adiposidad viscera

```
#METS-VF
Sup.Fig.1A<-base.2 %>%
    ggplot(aes(x = METSVF)) +
    geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 0.05) +
```

```
geom_density(col="#29498d")+
    labs(title = "Metabolic Score for Visceral Fat")+
    xlab ("METS-VF") +
    ylab ("Densidad de área") +
    theme_classic()
  Sup.Fig.1A<-egg::tag_facet(Sup.Fig.1A, x = Inf, y = Inf,
                 hjust = 1, open = "", close = "",
                 tag_pool = c(paste0("Media (D.E.)= ", paste0(round(mean(base.2$METSVF),2),"
                                      "\nMediana (R.I.Q.)= ", paste0(round(median(base.2$MET
                                      "\nn= ",nrow(base.2))))
Warning: The dot-dot notation (`..density...`) was deprecated in ggplot2 3.4.0.
i Please use `after_stat(density)` instead.
  #DAAT
  Sup.Fig.1B<-base.2 %>%
    ggplot(aes(x = DAAT_index)) +
    geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 10) +
    geom_density(col="#29498d")+
    labs(title = "Deep-Abdominal-Adipose-Tissue")+
    xlab ("DAAT") +
    ylab ("Densidad de área") +
    theme_classic()
  Sup.Fig.1B<-egg::tag_facet(Sup.Fig.1B, x = Inf, y = Inf,</pre>
                              hjust = 1, open = "", close = "",
                              tag_pool = c(paste0("Media (D.E.) = ", paste0(round(mean(base.2$)
                                                  "\nMediana (R.I.Q.)= ", paste0(round(media
                                                  "\nn= ",nrow(base.2))))
  #VAI
  Sup.Fig.1C<-base.2 %>%
    ggplot(aes(x = VAI_index)) +
    geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 0.1) +
    geom_density(col="#29498d")+
    labs(title = "Visceral Adiposity Index ")+
    xlab ("VAI") +
    ylab ("Densidad de área") +
    theme_classic()
```

```
Sup.Fig.1C\leftarrow-egg::tag_facet(Sup.Fig.1C, x = Inf, y = Inf,
                           hjust = 1, open = "", close = "",
                           tag_pool = c(paste0("Media (D.E.)= ", paste0(round(mean(base.2$)))
                                                "\nMediana (R.I.Q.)= ", paste0(round(media
                                                "\nn= ",nrow(base.2))))
#VAI-GEA
Sup.Fig.1D<-base.2 %>%
  ggplot(aes(x = VAI_GEA_index)) +
  geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 0.1) +
  geom_density(col="#29498d")+
  labs(title = "Dysfunctional Adiposity Index ")+
  xlab ("DAI") +
 ylab ("Densidad de área") +
 theme_classic()
Sup.Fig.1D<-egg::tag_facet(Sup.Fig.1D, x = Inf, y = Inf,
                           hjust = 1, open = "", close = "",
                           tag_pool = c(paste0("Media (D.E.)= ", paste0(round(mean(base.2$)
                                                "\nMediana (R.I.Q.)= ", paste0(round(media
                                                "\nn= ",nrow(base.2))))
#LAAP
Sup.Fig.1E<-base.2 %>%
  ggplot(aes(x = LAAP_index)) +
  geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 2.5) +
  geom_density(col="#29498d")+
 labs(title = "Lipid Accumulation Product")+
 xlab ("LAP") +
 ylab ("Densidad de área") +
  theme_classic()
Sup.Fig.1E<-egg::tag_facet(Sup.Fig.1E, x = Inf, y = Inf,
                           hjust = 1, open = "", close = "",
                           tag_pool = c(paste0("Media (D.E.) = ", paste0(round(mean(base.2$)
                                                "\nMediana (R.I.Q.)= ", paste0(round(media
                                                "\nn= ",nrow(base.2))))
```

```
#EVA
Sup.Fig.1F<-base.2 %>%
  ggplot(aes(x = EVA_index)) +
  geom_histogram(aes(y=..density..),fill="#e79b33",col="white", binwidth = 5) +
  geom_density(col="#29498d")+
  labs(title = "Estimate of Visceral Adipose Tissue Area ")+
  xlab ("EVA") +
  ylab ("Densidad de área") +
  theme_classic()
Sup.Fig.1F<-egg::tag_facet(Sup.Fig.1F, x = Inf, y = Inf,
                           hjust = 1, open = "", close = "",
                           tag_pool = c(paste0("Media (D.E.)= ", paste0(round(mean(base.2$)
                                                "\nMediana (R.I.Q.)= ", paste0(round(media
       "\nn= ",nrow(base.2))))
ggarrange(Sup.Fig.1A,Sup.Fig.1B,Sup.Fig.1C,Sup.Fig.1D,Sup.Fig.1E,Sup.Fig.1F,ncol = 3,nrow
                                 Deep-Abdomir C
                                                        Visceral Adiposit
     1.00
                                                    0.5
                                                    0.4
     0.75
                                 <u>|</u>= 161<mark>.04</mark> (60.84)
                                                        D.E<u>/.)=</u>
     0.50
                                  116,02-200.62)
                                                    0.2
                                                         (IQR: 1.78
```

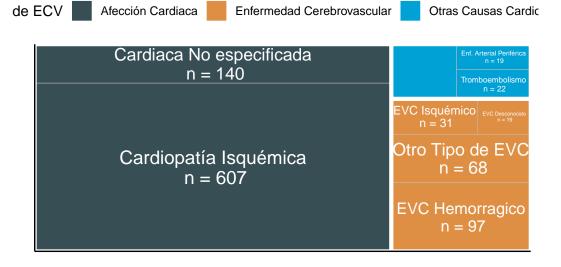


Resultados: Capitulo 3 - Evaluación de causas de muerte cardiovascular

```
df<- data.frame(CLUSTERS=c("Afección Cardiaca",</pre>
                           "Afección Cardiaca",
                           "Enfermedad Cerebrovascular".
                           "Enfermedad Cerebrovascular".
                           "Enfermedad Cerebrovascular",
                           "Enfermedad Cerebrovascular",
                           "Otras Causas Cardiovasculares",
                           "Otras Causas Cardiovasculares",
                           "Otras Causas Cardiovasculares"),
                prevalence=c(table(base.2$D001)[2],
                             table(base.2$D002)[2],
                             table(base.2$D004)[2],
                             table(base.2$D005)[2],
                             table(base.2$D006)[2],
                             table(base.2$D007)[2],
                             table(base.2$D009)[2],
                             table(base.2$D010)[2],
                             table(base.2$D011)[2]),
                hemisphere=c("Afección Cardiaca",
                             "Afección Cardiaca",
                             "Enfermedad Cerebrovascular",
                             "Enfermedad Cerebrovascular",
                             "Enfermedad Cerebrovascular",
                             "Enfermedad Cerebrovascular",
                             "Otras Causas Cardiovasculares",
                             "Otras Causas Cardiovasculares",
                             "Otras Causas Cardiovasculares"),
                labels=c(paste0("Cardiopatía Isquémica\n n = ",table(base.2$D001)[2]),
                         pasteO("Cardiaca No especificada\n n = ",table(base.2$D002)[2]),
                         pasteO("EVC Isquémico n = ",table(base.2$D004)[2]),
                         paste0("EVC Hemorragico\n n = ",table(base.2$D005)[2]),
                         paste0("EVC Desconocido\n n = ",table(base.2$D006)[2]),
                         paste0("Otro Tipo de EVC\n n = ",table(base.2$D007)[2]),
                         paste0("Tromboembolismo\n n = ",table(base.2$D009)[2]),
                         paste0("Enf. Arterial Periférica\n n = ",table(base.2$D010)[2]),
                         pasteO("Otras Muertes No Especificadas\n n = ",table(base.2$D011)
ggplot2::ggplot(df,aes(area=prevalence,
```

Mortalidad por Causas Cardiovasculares

Cohorte Prospectiva de la Ciudad de México



Resultados: Capitulo 3 - Resultados del modelaje de riesgos competitivos

```
#Muerte All Vascular base.2$MUERTE_VASCULAR
```

Warning: Unknown or uninitialised column: `MUERTE_VASCULAR`.

```
base.2$MUERTE_VASCULAR[base.2$STATUS=="A" & base.2$D015==0]<-0
Warning: Unknown or uninitialised column: `MUERTE_VASCULAR`.
  base.2$MUERTE_VASCULAR[base.2$STATUS=="D" & base.2$D015==1]<-1
  base.2$MUERTE_VASCULAR[base.2$STATUS=="D" & base.2$D015==0]<-2
  event<- base.2$MUERTE_VASCULAR</pre>
  event<- factor(event, 0:2, labels=c("censor","Vascular","death_other"))</pre>
  fgfit1<-FGR(Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(METSVF), data=base.2, cause = 1)</pre>
  summary(fgfit1)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(METSVF),
    data = base.2, cause = 1)
              coef exp(coef) se(coef)
                                          z p-value
                        2.59 0.0446 21.3
scale(METSVF) 0.95
              exp(coef) exp(-coef) 2.5% 97.5%
                          0.387 2.37 2.82
scale(METSVF)
                   2.59
Num. cases = 29034
Pseudo Log-likelihood = -10255
Pseudo likelihood ratio test = 636 on 1 df,
  #Deep-Abdominal Adipose Tissue index
  fgfit2<-FGR(Hist(PERSON_YEARS, MUERTE_VASCULAR) ~scale(DAAT_index), data=base.2, cause = 1)</pre>
  summary(fgfit2)
```

Competing Risks Regression

```
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(DAAT_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef)
                                              z p-value
scale(DAAT_index) 0.383
                             1.47
                                    0.0264 14.5
                  exp(coef) exp(-coef) 2.5% 97.5%
                       1.47
                                 0.682 1.39 1.54
scale(DAAT_index)
Num. cases = 29034
Pseudo Log-likelihood = -10490
Pseudo likelihood ratio test = 167 on 1 df,
  #Visceral Adiposity Index
  fgfit3<-FGR(Hist(PERSON_YEARS,MUERTE_VASCULAR)~scale(VAI_index),data=base.2,cause = 1)</pre>
  summary(fgfit3)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(VAI_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef)
                                               z p-value
scale(VAI_index) 0.0273
                         1.03 0.0302 0.903
                                                    0.37
                 exp(coef) exp(-coef) 2.5% 97.5%
                     1.03
                           0.973 0.969 1.09
scale(VAI_index)
Num. cases = 29034
Pseudo Log-likelihood = -10573
Pseudo likelihood ratio test = 0.78 on 1 df,
  #GEA-VAI
  fgfit4<-FGR(Hist(PERSON_YEARS, MUERTE_VASCULAR)~scale(VAI_GEA_index), data=base.2, cause = 1)
  summary(fgfit4)
```

Competing Risks Regression

```
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(VAI_GEA_index),
    data = base.2, cause = 1)
                       coef exp(coef) se(coef)
                                                  z p-value
scale(VAI_GEA_index) 0.0443
                                 1.05
                                      0.0298 1.49
                                                        0.14
                     exp(coef) exp(-coef) 2.5\% 97.5\%
scale(VAI_GEA_index)
                          1.05
                                    0.957 0.986 1.11
Num. cases = 29034
Pseudo Log-likelihood = -10572
Pseudo likelihood ratio test = 2.07 on 1 df,
  #Lipid Accumulation Product
  fgfit5<-FGR(Hist(PERSON_YEARS,MUERTE_VASCULAR)~scale(LAAP_index),data=base.2,cause = 1)</pre>
  summary(fgfit5)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(LAAP_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef) z p-value
scale(LAAP_index) 0.179
                             1.2 0.0267 6.7 2.1e-11
                  exp(coef) exp(-coef) 2.5% 97.5%
                        1.2
                             0.836 1.14 1.26
scale(LAAP_index)
Num. cases = 29034
Pseudo Log-likelihood = -10555
Pseudo likelihood ratio test = 35.9 on 1 df,
  #Eva Index
  fgfit6<-FGR(Hist(PERSON_YEARS,MUERTE_VASCULAR)~scale(EVA_index),data=base.2,cause = 1)</pre>
  summary(fgfit6)
```

Competing Risks Regression

```
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(EVA_index),
    data = base.2, cause = 1)
                  coef exp(coef) se(coef) z p-value
scale(EVA_index) 0.735
                            2.09
                                   0.0273 26.9
                 exp(coef) exp(-coef) 2.5% 97.5%
                      2.09
                                 0.48 1.98 2.2
scale(EVA_index)
Num. cases = 29034
Pseudo Log-likelihood = -10272
Pseudo likelihood ratio test = 602 on 1 df,
  #Depress
  fgfit7<-FGR(Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(Depres_index), data=base.2, cause = 1)</pre>
  summary(fgfit7)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_VASCULAR) ~ scale(Depres_index),
    data = base.2, cause = 1)
                     coef exp(coef) se(coef)
                                                z p-value
scale(Depres_index) 0.907
                               2.48
                                       0.046 19.7
                    exp(coef) exp(-coef) 2.5% 97.5%
                         2.48
                                0.404 2.26 2.71
scale(Depres_index)
Num. cases = 10593
Pseudo Log-likelihood = -3965
Pseudo likelihood ratio test = 378 on 1 df,
  #Any Cardiac Death
  base.2$MUERTE_CARDIO
Warning: Unknown or uninitialised column: `MUERTE_CARDIO`.
```

NULL

```
base.2$MUERTE_CARDIO[base.2$STATUS=="A" & base.2$D003==0]<-0
Warning: Unknown or uninitialised column: `MUERTE_CARDIO`.
  base.2$MUERTE_CARDIO[base.2$STATUS=="D" & base.2$D003==1]<-1
  base.2$MUERTE_CARDIO[base.2$STATUS=="D" & base.2$D003==0]<-2
  #METS-VF
  fg.fit.1.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(METSVF), data=base.2, cause = 1)
  summary(fg.fit.1.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(METSVF),
    data = base.2, cause = 1)
              coef exp(coef) se(coef) z p-value
scale(METSVF) 0.94
                       2.56
                              0.0524 17.9
              exp(coef) exp(-coef) 2.5% 97.5%
scale(METSVF)
                   2.56
                            0.391 2.31 2.84
Num. cases = 29034
Pseudo Log-likelihood = -7370
Pseudo likelihood ratio test = 450 on 1 df,
  #Deep-Abdominal Adipose Tissue index
  fg.fit.2.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(DAAT_index), data=base.2, cause
  summary(fg.fit.2.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(DAAT_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef)
                                              z p-value
scale(DAAT_index) 0.393 1.48 0.0311 12.6
```

```
exp(coef) exp(-coef) 2.5% 97.5%
scale(DAAT_index)
                       1.48
                                 0.675 1.39 1.57
Num. cases = 29034
Pseudo Log-likelihood = -7532
Pseudo likelihood ratio test = 126 on 1 df,
  #Visceral Adiposity Index
  fg.fit.3.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(VAI_index), data=base.2, cause =
  summary(fg.fit.3.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(VAI_index),
    data = base.2, cause = 1)
                     coef exp(coef) se(coef)
                                                  z p-value
scale(VAI_index) -0.00692
                            0.993 0.0363 -0.191
                                                       0.85
                 exp(coef) exp(-coef) 2.5\% 97.5\%
                    0.993
scale(VAI_index)
                                1.01 0.925 1.07
Num. cases = 29034
Pseudo Log-likelihood = -7595
Pseudo likelihood ratio test = 0.04 on 1 df,
  #GEA-VAI
  fg.fit.4.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(VAI_GEA_index), data=base.2, cau
  summary(fg.fit.4.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(VAI_GEA_index),
    data = base.2, cause = 1)
                       coef exp(coef) se(coef)
                                                   z p-value
scale(VAI_GEA_index) 0.0116
                            1.01 0.0357 0.325 0.74
```

```
exp(coef) exp(-coef) 2.5% 97.5%
scale(VAI_GEA_index)
                          1.01
                                    0.988 0.943 1.09
Num. cases = 29034
Pseudo Log-likelihood = -7595
Pseudo likelihood ratio test = 0.1 on 1 df,
  #Lipid Accumulation Product
  fg.fit.5.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(LAAP_index), data=base.2, cause
  summary(fg.fit.5.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(LAAP_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef)
                                              z p-value
scale(LAAP_index) 0.162
                             1.18
                                  0.0313 5.18 2.3e-07
                  exp(coef) exp(-coef) 2.5% 97.5%
scale(LAAP_index)
                       1.18
                                 0.85 1.11 1.25
Num. cases = 29034
Pseudo Log-likelihood = -7584
Pseudo likelihood ratio test = 20.9 on 1 df,
  #Eva Index
  fg.fit.6.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(EVA_index), data=base.2, cause =
  summary(fg.fit.6.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(EVA_index),
    data = base.2, cause = 1)
                  coef exp(coef) se(coef)
                                             z p-value
scale(EVA_index) 0.737
                            2.09 0.0319 23.1
```

```
exp(coef) exp(-coef) 2.5% 97.5%
                      2.09
                              0.479 1.96 2.22
scale(EVA_index)
Num. cases = 29034
Pseudo Log-likelihood = -7377
Pseudo likelihood ratio test = 436 on 1 df,
  #Depress
  fg.fit.7.cardio<-FGR(Hist(PERSON_YEARS, MUERTE_CARDIO)~scale(Depres_index), data=base.2, caus
  summary(fg.fit.7.cardio)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_CARDIO) ~ scale(Depres_index),
    data = base.2, cause = 1)
                     coef exp(coef) se(coef)
                                                z p-value
                               2.39
                                      0.0534 16.3
scale(Depres_index) 0.869
                    exp(coef) exp(-coef) 2.5% 97.5%
scale(Depres_index)
                         2.39
                                0.419 2.15 2.65
Num. cases = 10593
Pseudo Log-likelihood = -2894
Pseudo likelihood ratio test = 255 on 1 df,
  #Any Cerebrovascular Death
  base.2$MUERTE_STROKE
Warning: Unknown or uninitialised column: `MUERTE_STROKE`.
NULL
  base.2$MUERTE_STROKE[base.2$STATUS=="A" & base.2$D008==0]<-0
```

Warning: Unknown or uninitialised column: `MUERTE_STROKE`.

```
base.2$MUERTE_STROKE[base.2$STATUS=="D" & base.2$D008==1]<-1
  base.2$MUERTE_STROKE[base.2$STATUS=="D" & base.2$D008==0]<-2
  #METS-VF
  fg.fit.1.stroke<-FGR(Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(METSVF), data=base.2, cause = 1)
  summary(fg.fit.1.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(METSVF),
    data = base.2, cause = 1)
              coef exp(coef) se(coef)
                                         z p-value
                      2.74 0.0976 10.3
scale(METSVF) 1.01
              exp(coef) exp(-coef) 2.5% 97.5%
                 2.74
                          0.364 2.27 3.32
scale(METSVF)
Num. cases = 29034
Pseudo Log-likelihood = -2127
Pseudo likelihood ratio test = 146 on 1 df,
  #Deep-Abdominal Adipose Tissue index
  fg.fit.2.stroke<-FGR(Hist(PERSON_YEARS, MUERTE_STROKE)~scale(DAAT_index), data=base.2, cause
  summary(fg.fit.2.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(DAAT_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef) z p-value
                          1.39 0.0564 5.87 4.3e-09
scale(DAAT_index) 0.331
                  exp(coef) exp(-coef) 2.5% 97.5%
                      1.39
                                 0.718 1.25 1.56
scale(DAAT_index)
Num. cases = 29034
```

```
Pseudo Log-likelihood = -2187
Pseudo likelihood ratio test = 25.6 on 1 df,
  #Visceral Adiposity Index
  fg.fit.3.stroke<-FGR(Hist(PERSON_YEARS,MUERTE_STROKE)~scale(VAI_index),data=base.2,cause =</pre>
  summary(fg.fit.3.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(VAI_index),
    data = base.2, cause = 1)
                  coef exp(coef) se(coef)
                                             z p-value
                                   0.064 1.89 0.059
scale(VAI_index) 0.121
                            1.13
                 exp(coef) exp(-coef) 2.5\% 97.5\%
                      1.13 0.886 0.995 1.28
scale(VAI_index)
Num. cases = 29034
Pseudo Log-likelihood = -2198
Pseudo likelihood ratio test = 3.33 on 1 df,
  #GEA-VAI
  fg.fit.4.stroke<-FGR(Hist(PERSON_YEARS, MUERTE_STROKE)~scale(VAI_GEA_index), data=base.2, cau
  summary(fg.fit.4.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(VAI_GEA_index),
    data = base.2, cause = 1)
                      coef exp(coef) se(coef)
                                                 z p-value
scale(VAI_GEA_index) 0.114
                               1.12 0.0639 1.78 0.075
                     exp(coef) exp(-coef) 2.5\% 97.5\%
                         1.12
                                  0.892 0.989 1.27
scale(VAI_GEA_index)
Num. cases = 29034
```

```
Pseudo Log-likelihood = -2198
Pseudo likelihood ratio test = 2.94 on 1 df,
  #Lipid Accumulation Product
  fg.fit.5.stroke<-FGR(Hist(PERSON_YEARS, MUERTE_STROKE)~scale(LAAP_index), data=base.2, cause
  summary(fg.fit.5.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(LAAP_index),
    data = base.2, cause = 1)
                                              z p-value
                   coef exp(coef) se(coef)
                         1.31 0.0584 4.64 3.6e-06
scale(LAAP_index) 0.271
                  exp(coef) exp(-coef) 2.5% 97.5%
scale(LAAP_index)
                      1.31 0.763 1.17 1.47
Num. cases = 29034
Pseudo Log-likelihood = -2191
Pseudo likelihood ratio test = 17.6 on 1 df,
  #Eva Index
  fg.fit.6.stroke<-FGR(Hist(PERSON_YEARS,MUERTE_STROKE)~scale(EVA_index),data=base.2,cause =</pre>
  summary(fg.fit.6.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(EVA_index),
    data = base.2, cause = 1)
                  coef exp(coef) se(coef)
                                            z p-value
scale(EVA_index) 0.718
                            2.05 0.0591 12.1
                 exp(coef) exp(-coef) 2.5% 97.5%
                            0.488 1.83
                      2.05
scale(EVA_index)
Num. cases = 29034
```

```
Pseudo Log-likelihood = -2140
Pseudo likelihood ratio test = 120 on 1 df,
  #Depress
  fg.fit.7.stroke<-FGR(Hist(PERSON_YEARS, MUERTE_STROKE)~scale(Depres_index), data=base.2, caus
  summary(fg.fit.7.stroke)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_STROKE) ~ scale(Depres_index),
    data = base.2, cause = 1)
                    coef exp(coef) se(coef)
                                               z p-value
scale(Depres_index) 1.05
                              2.85
                                       0.11 9.49
                    exp(coef) exp(-coef) 2.5% 97.5%
                         2.85
                               0.351 2.29 3.54
scale(Depres_index)
Num. cases = 10593
Pseudo Log-likelihood = -695
Pseudo likelihood ratio test = 88.6 on 1 df,
  #Other CVD
  base.2$MUERTE_OTRAS
Warning: Unknown or uninitialised column: `MUERTE_OTRAS`.
NULL
  base.2\$MUERTE_OTRAS[base.2\$D009==1]<-1
Warning: Unknown or uninitialised column: `MUERTE_OTRAS`.
  base.2$MUERTE_OTRAS[base.2$D010==1]<-1
  base.2$MUERTE_OTRAS[base.2$D011==1]<-1
  base.2$MUERTE_OTRAS[base.2$D012==1]<-1
```

```
base.2$MUERTE_OTRAS<-na.tools::na.replace(base.2$MUERTE_OTRAS,0)
  base.2$MUERTE OTRAS 2[base.2$STATUS=="A" & base.2$MUERTE OTRAS==0]<-0
Warning: Unknown or uninitialised column: `MUERTE_OTRAS_2`.
  base.2$MUERTE_OTRAS_2[base.2$STATUS=="D" & base.2$MUERTE_OTRAS==1]<-1
  base.2$MUERTE_OTRAS_2[base.2$STATUS=="D" & base.2$MUERTE_OTRAS==0]<-2
  #METS-VF
  fg.fit.1.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(METSVF), data=base.2, cause = 1
  summary(fg.fit.1.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(METSVF),
    data = base.2, cause = 1)
               coef exp(coef) se(coef) z p-value
scale(METSVF) 0.734
                         2.08
                                 0.151 4.85 1.2e-06
              exp(coef) exp(-coef) 2.5% 97.5%
scale(METSVF)
                   2.08
                              0.48 1.55 2.8
Num. cases = 29034
Pseudo Log-likelihood = -771
Pseudo likelihood ratio test = 30.7 on 1 df,
  #Deep-Abdominal Adipose Tissue index
  fg.fit.2.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(DAAT_index), data=base.2, cause
  summary(fg.fit.2.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(DAAT_index),
    data = base.2, cause = 1)
```

```
coef exp(coef) se(coef)
                                              z p-value
scale(DAAT_index) 0.392
                             1.48
                                     0.102 3.83 0.00013
                  exp(coef) exp(-coef) 2.5% 97.5%
                      1.48
                                 0.676 1.21 1.81
scale(DAAT_index)
Num. cases = 29034
Pseudo Log-likelihood = -780
Pseudo likelihood ratio test = 12.9 on 1 df,
  #Visceral Adiposity Index
  fg.fit.3.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(VAI_index), data=base.2, cause
  summary(fg.fit.3.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(VAI_index),
    data = base.2, cause = 1)
                   coef exp(coef) se(coef)
                                             z p-value
scale(VAI_index) 0.0813
                             1.08
                                     0.102 0.8
                                                  0.42
                 exp(coef) exp(-coef) 2.5\% 97.5\%
scale(VAI_index)
                      1.08
                              0.922 0.889 1.32
Num. cases = 29034
Pseudo Log-likelihood = -786
Pseudo likelihood ratio test = 0.53 on 1 df,
  #GEA-VAI
  fg.fit.4.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(VAI_GEA_index), data=base.2, ca
  summary(fg.fit.4.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(VAI_GEA_index),
    data = base.2, cause = 1)
```

```
coef exp(coef) se(coef)
                                                z p-value
scale(VAI_GEA_index) 0.15
                               1.16
                                      0.0979 1.53
                                                     0.13
                     exp(coef) exp(-coef) 2.5% 97.5%
                        1.16
                                0.861 0.959 1.41
scale(VAI_GEA_index)
Num. cases = 29034
Pseudo Log-likelihood = -786
Pseudo likelihood ratio test = 1.85 on 1 df,
  #Lipid Accumulation Product
  fg.fit.5.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(LAAP_index), data=base.2, cause
  summary(fg.fit.5.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(LAAP_index),
    data = base.2, cause = 1)
                    coef exp(coef) se(coef)
                                                z p-value
scale(LAAP_index) 0.0641
                              1.07
                                      0.103 0.619
                                                     0.54
                  exp(coef) exp(-coef) 2.5% 97.5%
scale(LAAP_index)
                      1.07
                                 0.938 0.87 1.31
Num. cases = 29034
Pseudo Log-likelihood = -787
Pseudo likelihood ratio test = 0.32 on 1 df,
  #Eva Index
  fg.fit.6.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(EVA_index), data=base.2, cause
  summary(fg.fit.6.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(EVA_index),
    data = base.2, cause = 1)
```

```
coef exp(coef) se(coef) z p-value
                                     0.101 6.38 1.8e-10
scale(EVA_index) 0.642
                             1.9
                 \exp(\texttt{coef}) \exp(-\texttt{coef}) 2.5\% 97.5\%
                            0.526 1.56 2.32
scale(EVA_index)
                       1.9
Num. cases = 29034
Pseudo Log-likelihood = -770
Pseudo likelihood ratio test = 34.3 on 1 df,
  #Depress
  fg.fit.7.others<-FGR(Hist(PERSON_YEARS, MUERTE_OTRAS_2)~scale(Depres_index), data=base.2, cau
  summary(fg.fit.7.others)
Competing Risks Regression
Call:
FGR(formula = Hist(PERSON_YEARS, MUERTE_OTRAS_2) ~ scale(Depres_index),
    data = base.2, cause = 1)
                     coef exp(coef) se(coef)
                                                 z p-value
                               2.13
scale(Depres_index) 0.757
                                       0.128 5.91 3.4e-09
                    exp(coef) exp(-coef) 2.5% 97.5%
scale(Depres_index)
                         2.13
                                 0.469 1.66 2.74
Num. cases = 10593
Pseudo Log-likelihood = -384
Pseudo likelihood ratio test = 25.7 on 1 df,
```

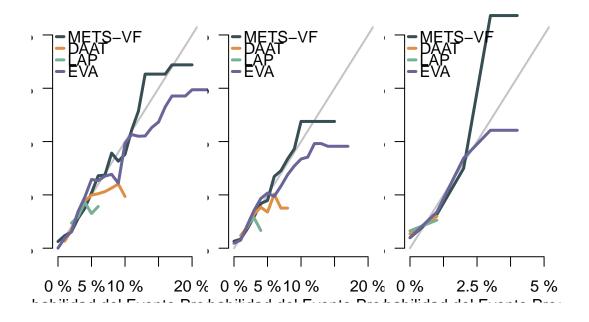
Resultados: Capitulo 3 - Métricas de calibración

```
xlim=c(0,0.20),
                                 ylim=c(0,0.20),
                                 xlab = c("Probabilidad del Evento Predicho"),
                                 ylab = c("Frecuencia de Eventos Observados")))+
    ggtitle("Todas las Causas Cardiovasculares")
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  Figure 4B < - ggplotify::as.ggplot(~calPlot(list("METS-VF"=fg.fit.1.cardio,
               "DAAT"=fg.fit.2.cardio,
               "LAP"=fg.fit.5.cardio,
               "EVA"=fg.fit.6.cardio),
          time=c(20), cores = 3, col = c(ggsci::pal_jama("default")(6))[c(1:2,5:6)],
          showPseudo=FALSE,
          type="risk",
          data=base.2,
          xlim=c(0,0.20),
          vlim=c(0,0.20),
          xlab = c("Probabilidad del Evento Predicho"),
          ylab = c("Frecuencia de Eventos Observados")))+
    ggtitle("Afecciones Cardíacas")
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  Figure4C<-ggplotify::as.ggplot(~calPlot(list("METS-VF"=fg.fit.1.stroke,
               "DAAT"=fg.fit.2.stroke,
               "LAP"=fg.fit.5.stroke,
               "EVA"=fg.fit.6.stroke),
          time=c(15), cores = 3, col = c(ggsci::pal_jama("default")(6))[c(1:2,5:6)],
          showPseudo=FALSE,
          type="risk",
          data=base.2,
          xlim=c(0,0.05),
```

```
ylim=c(0,0.05),
    xlab = c("Probabilidad del Evento Predicho"),
    ylab = c("Frecuencia de Eventos Observados")))+
    ggtitle("Enfermedades Cerebrovasculares")

[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"

ggarrange(Figure4A,Figure4B,Figure4C,ncol = 3,nrow = 1)
```

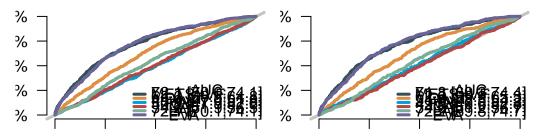
Resultados: Capitulo 3 - Metricas de rendimiento general

```
"LAP"=fgfit5,
                   "EVA"=fgfit6),
              formula=Hist(PERSON_YEARS,MUERTE_VASCULAR)~1,
              data=base.2,
              conf.int=TRUE,
              summary="risks",
              metrics="auc",
              plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Ischemic Heart Disease)
  ROC.df.2<-Score(list("METS-VF"=fg.fit.1.cardio,</pre>
                        "DAAT"=fg.fit.2.cardio,
                        "VAI"=fg.fit.3.cardio,
                        "DAI"=fg.fit.4.cardio,
                        "LAP"=fg.fit.5.cardio,
                        "EVA"=fg.fit.6.cardio),
                   formula=Hist(PERSON_YEARS, MUERTE_CARDIO)~1,
                   data=base.2,
                   conf.int=TRUE,
                   summary="risks",
                   metrics="auc",
                   plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Stroke)
  ROC.df.3<-Score(list("METS-VF"=fg.fit.1.stroke,</pre>
```

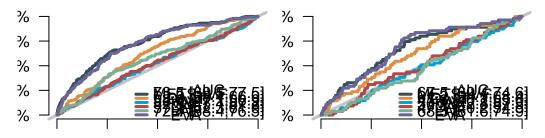
```
"DAAT"=fg.fit.2.stroke,
                        "VAI"=fg.fit.3.stroke,
                        "DAI"=fg.fit.4.stroke,
                        "LAP"=fg.fit.5.stroke,
                        "EVA"=fg.fit.6.stroke),
                   formula=Hist(PERSON_YEARS, MUERTE_STROKE)~1,
                   data=base.2,
                   conf.int=TRUE,
                   summary="risks",
                   metrics="auc",
                   plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT index)"
[1] "scale(VAI_index)"
[1] "scale(VAI GEA index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Others)
  ROC.df.4<-Score(list("METS-VF"=fg.fit.1.others,</pre>
                        "DAAT"=fg.fit.2.others,
                        "VAI"=fg.fit.3.others,
                        "DAI"=fg.fit.4.others,
                        "LAP"=fg.fit.5.others,
                        "EVA"=fg.fit.6.others),
                   formula=Hist(PERSON_YEARS, MUERTE_OTRAS_2)~1,
                   data=base.2,
                   conf.int=TRUE,
                   summary="risks",
                   metrics="auc",
                   plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
```

```
#ROC CURVE
Figure 2A < - ggplotify::as.ggplot(~plotROC(ROC.df.1,col = c(ggsci::pal_jama("default")(6)),
                              xlab = "Tasa de Falsos Positivos (1-Especificidad)",
                              ylab = "Tasa de Verdaderos Positivos (Sensibilidad)"))+
  ggtitle("Todas las Causas Cardiovasculares")
Figure 2B < - ggplotify::as.ggplot(~plotROC(ROC.df.2,col = c(ggsci::pal_jama("default")(6)),
                                         xlab = "Tasa de Falsos Positivos (1-Especificidad)
                                         ylab = "Tasa de Verdaderos Positivos (Sensibilidad
  ggtitle("Afecciones Cardíacas")
Figure 2C <- ggplotify::as.ggplot(~plotROC(ROC.df.3,col = c(ggsci::pal_jama("default")(6)),
                                         xlab = "Tasa de Falsos Positivos (1-Especificidad)
                                         ylab = "Tasa de Verdaderos Positivos (Sensibilidad
  ggtitle("Enfermedades Cerebrovasculares")
Figure 2D < - ggplotify::as.ggplot(~plotROC(ROC.df.4,col = c(ggsci::pal_jama("default")(6)),
                                         xlab = "Tasa de Falsos Positivos (1-Especificidad)
                                         ylab = "Tasa de Verdaderos Positivos (Sensibilidad
  ggtitle("Otras Causas Cardiovasculares")
ggarrange(Figure2A,Figure2B,Figure2C,Figure2D,ncol = 2,nrow = 2)
```

Todas las Causas Cardiovasculare Afecciones Cardíacas



Enfermedades Cerebrovasculares Otras Causas Cardiovasculares



Resultados: Capitulo 3 - Estimación del rendimiento pronóstico durante el tiempo de seguimiento

- [1] "scale(METSVF)"
- [1] "scale(DAAT_index)"
- [1] "scale(VAI_index)"

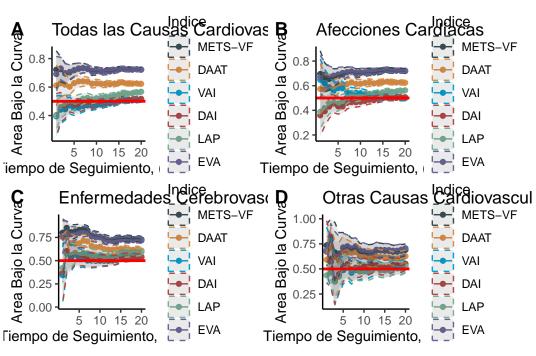
```
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Cardiac Causes)
  ROC.time.df.2<-Score(list("METS-VF"=fg.fit.1.cardio,</pre>
                           "DAAT"=fg.fit.2.cardio,
                           "VAI"=fg.fit.3.cardio,
                           "DAI"=fg.fit.4.cardio,
                           "LAP"=fg.fit.5.cardio,
                           "EVA"=fg.fit.6.cardio),
                      formula=Hist(PERSON_YEARS,MUERTE_CARDIO)~1,
                      data=base.2,
                      times = seq(1,20,1),
                      conf.int=TRUE,
                      summary="risks",
                      metrics="auc",
                      plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Cerebrovascular Causes)
  ROC.time.df.3<-Score(list("METS-VF"=fg.fit.1.stroke,</pre>
                             "DAAT"=fg.fit.2.stroke,
                             "VAI"=fg.fit.3.stroke,
                             "DAI"=fg.fit.4.stroke,
                             "LAP"=fg.fit.5.stroke,
                             "EVA"=fg.fit.6.stroke),
                        formula=Hist(PERSON_YEARS, MUERTE_STROKE) ~ 1,
                        data=base.2,
                        times = seq(1,20,1),
                        conf.int=TRUE,
                        summary="risks",
                        metrics="auc",
                        plots="roc")
```

```
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP index)"
[1] "scale(EVA_index)"
  #ROC Metrics (Other Causes)
  ROC.time.df.4<-Score(list("METS-VF"=fg.fit.1.others,</pre>
                             "DAAT"=fg.fit.2.others,
                             "VAI"=fg.fit.3.others,
                             "DAI"=fg.fit.4.others,
                             "LAP"=fg.fit.5.others,
                             "EVA"=fg.fit.6.others),
                        formula=Hist(PERSON_YEARS, MUERTE_OTRAS_2)~1,
                        data=base.2,
                        times = seq(1,20,1),
                        conf.int=TRUE,
                        summary="risks",
                        metrics="auc",
                        plots="roc")
[1] "scale(METSVF)"
[1] "scale(DAAT_index)"
[1] "scale(VAI_index)"
[1] "scale(VAI_GEA_index)"
[1] "scale(LAAP_index)"
[1] "scale(EVA_index)"
  Figure3A<-ggplot(ROC.time.df.1$AUC$score,
         aes(times,AUC,colour=model))+
    geom point()+
    geom_line()+
    geom_ribbon(aes(ymin=lower,
                     ymax=upper,
                     colour=model),
                 alpha=0.1,linetype=2)+
    ggsci::scale_colour_jama()+
    labs(colour="Indice")+
    theme_classic()+
```

Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0. i Please use `linewidth` instead.

```
Figure3B<-ggplot(ROC.time.df.2$AUC$score,
                 aes(times,AUC,colour=model))+
  geom point()+
  geom_line()+
  geom_ribbon(aes(ymin=lower,
                  ymax=upper,
                  colour=model),
              alpha=0.1,linetype=2)+
  ggsci::scale_colour_jama()+
  labs(colour="Indice")+
  theme_classic()+
  ylab("Area Bajo la Curva")+
  xlab("Tiempo de Seguimiento, (Años)")+
  ggtitle("Afecciones Cardíacas")+
  geom_hline(yintercept = 0.5, linetype=1,
             color = "red", size=1)
Figure3C<-ggplot(ROC.time.df.3$AUC$score,
                 aes(times,AUC,colour=model))+
  geom_point()+
  geom_line()+
  geom_ribbon(aes(ymin=lower,
                  ymax=upper,
                  colour=model),
              alpha=0.1,linetype=2)+
  ggsci::scale_colour_jama()+
  labs(colour="Indice")+
  theme_classic()+
  ylab("Area Bajo la Curva")+
  xlab("Tiempo de Seguimiento, (Años)")+
  ggtitle("Enfermedades Cerebrovasculares")+
  geom_hline(yintercept = 0.5, linetype=1,
```

```
color = "red", size=1)
Figure3D<-ggplot(ROC.time.df.4$AUC$score,
                 aes(times,AUC,colour=model))+
  geom_point()+
  geom_line()+
  geom_ribbon(aes(ymin=lower,
                  ymax=upper,
                  colour=model),
              alpha=0.1,linetype=2)+
  ggsci::scale_colour_jama()+
  labs(colour="Indice")+
  theme_classic()+
  ylab("Area Bajo la Curva")+
  xlab("Tiempo de Seguimiento, (Años)")+
  ggtitle("Otras Causas Cardiovasculares")+
  geom_hline(yintercept = 0.5, linetype=1,
             color = "red", size=1)
ggarrange(Figure3A, Figure3B, Figure3C, Figure3D, ncol = 2, nrow = 2, labels = LETTERS[1:4])
                                            Afecciones Cardia
                                                               METS-VF
```



Resultados: Capitulo 3 - Evaluación de puntos de corte específicos

```
#Muerte All Vascular
  #METS-VF
  res.cut.all.1 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.1)
       cutpoint statistic
METSVF 7.211766 24.29035
  #Deep-Abdominal Adipose Tissue index
  res.cut.all.2 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.2)
           cutpoint statistic
DAAT_index 145.206 13.38629
  #Visceral Adiposity Index
  res.cut.all.3 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.3)
          cutpoint statistic
VAI_index 2.012154 1.991938
  #GEA-VAI
  res.cut.all.4 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.4)
              cutpoint statistic
VAI_GEA_index 1.237504
                         2.61297
  #Lipid Accumulation Product
  res.cut.all.5 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.5)
           cutpoint statistic
LAAP_index 45.253 7.466332
```

```
#Eva Index
  res.cut.all.6 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.6)
          cutpoint statistic
EVA_index 133.2737 23.48369
  #Depress
  res.cut.all.7 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D015", va
  summary(res.cut.all.7)
             cutpoint statistic
Depres_index 174.9466
                       19.0239
  ###Cardiac Deaths
  #METS-VF
  res.cut.cardiac.1 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003
  summary(res.cut.cardiac.1)
       cutpoint statistic
METSVF 7.300217 20.86636
  #Deep-Abdominal Adipose Tissue index
  res.cut.cardiac.2 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003
  summary(res.cut.cardiac.2)
           cutpoint statistic
DAAT_index 150.8149 11.95281
  #Visceral Adiposity Index
  res.cut.cardiac.3 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003
  summary(res.cut.cardiac.3)
          cutpoint statistic
VAI_index 3.436877 1.631837
```

```
#GEA-VAI
  res.cut.cardiac.4 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003"
  summary(res.cut.cardiac.4)
               cutpoint statistic
VAI_GEA_index 0.9975464 2.342943
  #Lipid Accumulation Product
  res.cut.cardiac.5 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003
  summary(res.cut.cardiac.5)
           cutpoint statistic
LAAP index 45.078 6.363569
  #Eva Index
  res.cut.cardiac.6 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003
  summary(res.cut.cardiac.6)
          cutpoint statistic
EVA_index 135.2184 20.89256
  #Depress
  res.cut.cardiac.7 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D003"
  summary(res.cut.cardiac.7)
             {\tt cutpoint \ statistic}
Depres_index 179.2111 16.55898
  ##Any Cerebrovascular Deaths
  #METS-VF
  res.cut.cerebrovasc.1 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.1)
       cutpoint statistic
METSVF 7.231185 12.21993
```

```
#Deep-Abdominal Adipose Tissue index
  res.cut.cerebrovasc.2 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.2)
           cutpoint statistic
DAAT_index 153.9097 5.540596
  #Visceral Adiposity Index
  res.cut.cerebrovasc.3 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.3)
          cutpoint statistic
VAI_index 3.090636 2.674579
  #GEA-VAI
  res.cut.cerebrovasc.4 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.4)
              cutpoint statistic
VAI_GEA_index 1.383923 2.489902
  #Lipid Accumulation Product
  res.cut.cerebrovasc.5 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.5)
           cutpoint statistic
LAAP_index 57.4 5.370069
  res.cut.cerebrovasc.6 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.6)
          cutpoint statistic
EVA_index 133.1655 11.0112
```

```
#Depress
  res.cut.cerebrovasc.7 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "
  summary(res.cut.cerebrovasc.7)
             cutpoint statistic
Depres_index 200.509 8.930644
  #Other CVD
  #METS-VF
  res.cut.other.1 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.1)
       cutpoint statistic
METSVF 6.963835 3.596831
  #Deep-Abdominal Adipose Tissue index
  res.cut.other.2 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.2)
           cutpoint statistic
DAAT_index 122.0835 2.566053
  #Visceral Adiposity Index
  res.cut.other.3 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.3)
          cutpoint statistic
VAI_index 3.697909 1.287855
  #GEA-VAI
  res.cut.other.4 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.4)
              cutpoint statistic
VAI_GEA_index 1.798452 1.976538
```

```
#Lipid Accumulation Product
  res.cut.other.5 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.5)
           cutpoint statistic
LAAP_index 32.8032 1.456717
  #Eva Index
  res.cut.other.6 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.6)
          cutpoint statistic
EVA_index 117.4862 4.116698
  #Depress
  res.cut.other.7 <- survminer::surv_cutpoint(base.2, time = "PERSON_YEARS", event = "D009",
  summary(res.cut.other.7)
             cutpoint statistic
Depres_index 171.593 3.501927
  ##Muerte All Vascular
  res.cat.all.1 <- surv_categorize(res.cut.all.1)</pre>
  res.cat.all.2 <- surv_categorize(res.cut.all.2)</pre>
  res.cat.all.3 <- surv_categorize(res.cut.all.3)</pre>
  res.cat.all.4 <- surv_categorize(res.cut.all.4)</pre>
  res.cat.all.5 <- surv_categorize(res.cut.all.5)</pre>
  res.cat.all.6 <- surv_categorize(res.cut.all.6)</pre>
  cox.cat.all.1<-coxph(Surv(PERSON_YEARS, D015)~METSVF, data=res.cat.all.1)</pre>
  cox.cat.all.2<-coxph(Surv(PERSON_YEARS, D015)~DAAT_index, data=res.cat.all.2)</pre>
  cox.cat.all.3<-coxph(Surv(PERSON_YEARS, D015)~VAI_index, data=res.cat.all.3)</pre>
  cox.cat.all.4<-coxph(Surv(PERSON_YEARS, D015)~VAI_GEA_index, data=res.cat.all.4)</pre>
  cox.cat.all.5<-coxph(Surv(PERSON_YEARS, D015)~LAAP_index, data=res.cat.all.5)
  cox.cat.all.6<-coxph(Surv(PERSON_YEARS, D015)~EVA_index, data=res.cat.all.6)
  c.stat.all.1<-paste0(round(cox.cat.all.1$concordance[6],2)," (",round(cox.cat.all.1$concordance
  c.stat.all.2<-paste0(round(cox.cat.all.2$concordance[6],2)," (",round(cox.cat.all.2$concordance</pre>
```

```
c.stat.all.3<-paste0(round(cox.cat.all.3$concordance[6],2)," (",round(cox.cat.all.3$concordance
c.stat.all.4<-paste0(round(cox.cat.all.4$concordance[6],2)," (",round(cox.cat.all.4$concordance</pre>
c.stat.all.5<-paste0(round(cox.cat.all.5$concordance[6],2)," (",round(cox.cat.all.5$concor</pre>
c.stat.all.6<-paste0(round(cox.cat.all.6$concordance[6],2)," (",round(cox.cat.all.6$concordance
##Any Cardiac Death
res.cat.cardiac.1 <- surv_categorize(res.cut.cardiac.1)</pre>
res.cat.cardiac.2 <- surv_categorize(res.cut.cardiac.2)</pre>
res.cat.cardiac.3 <- surv_categorize(res.cut.cardiac.3)</pre>
res.cat.cardiac.4 <- surv_categorize(res.cut.cardiac.4)</pre>
res.cat.cardiac.5 <- surv_categorize(res.cut.cardiac.5)</pre>
res.cat.cardiac.6 <- surv_categorize(res.cut.cardiac.6)</pre>
cox.cat.cardiac.1<-coxph(Surv(PERSON_YEARS, D003)~METSVF, data=res.cat.cardiac.1)</pre>
cox.cat.cardiac.2<-coxph(Surv(PERSON_YEARS, D003)~DAAT_index, data=res.cat.cardiac.2)</pre>
cox.cat.cardiac.3<-coxph(Surv(PERSON_YEARS, D003)~VAI_index, data=res.cat.cardiac.3)
cox.cat.cardiac.4<-coxph(Surv(PERSON_YEARS, D003)~VAI_GEA_index, data=res.cat.cardiac.4)
cox.cat.cardiac.5<-coxph(Surv(PERSON_YEARS, D003)~LAAP_index, data=res.cat.cardiac.5)</pre>
cox.cat.cardiac.6<-coxph(Surv(PERSON_YEARS, D003)~EVA_index, data=res.cat.cardiac.6)
c.stat.cardiac.1<-paste0(round(cox.cat.cardiac.1$concordance[6],2)," (",round(cox.cat.cardiac.1$concordance[6],2)," (",
c.stat.cardiac.2<-paste0(round(cox.cat.cardiac.2$concordance[6],2)," (",round(cox.cat.card</pre>
c.stat.cardiac.3<-paste0(round(cox.cat.cardiac.3$concordance[6],2)," (",round(cox.cat.card</pre>
c.stat.cardiac.4<-paste0(round(cox.cat.cardiac.4$concordance[6],2)," (",round(cox.cat.cardiac.4$concordance[6],2)," (",
c.stat.cardiac.5<-paste0(round(cox.cat.cardiac.5$concordance[6],2)," (",round(cox.cat.card</pre>
c.stat.cardiac.6<-paste0(round(cox.cat.cardiac.6$concordance[6],2)," (",round(cox.cat.card</pre>
#Cerebrovascular
res.cat.cerebrovasc.1 <- surv_categorize(res.cut.cerebrovasc.1)</pre>
res.cat.cerebrovasc.2 <- surv_categorize(res.cut.cerebrovasc.2)</pre>
res.cat.cerebrovasc.3 <- surv_categorize(res.cut.cerebrovasc.3)</pre>
res.cat.cerebrovasc.4 <- surv_categorize(res.cut.cerebrovasc.4)</pre>
res.cat.cerebrovasc.5 <- surv_categorize(res.cut.cerebrovasc.5)</pre>
res.cat.cerebrovasc.6 <- surv_categorize(res.cut.cerebrovasc.6)</pre>
cox.cat.cerebrovasc.1<-coxph(Surv(PERSON_YEARS, D008)~METSVF, data=res.cat.cerebrovasc.1)</pre>
cox.cat.cerebrovasc.2<-coxph(Surv(PERSON_YEARS, D008)~DAAT_index, data=res.cat.cerebrovasc
cox.cat.cerebrovasc.3<-coxph(Surv(PERSON_YEARS, D008)~VAI_index, data=res.cat.cerebrovasc.
cox.cat.cerebrovasc.4<-coxph(Surv(PERSON_YEARS, D008)~VAI_GEA_index, data=res.cat.cerebrov
cox.cat.cerebrovasc.5<-coxph(Surv(PERSON_YEARS, D008)~LAAP_index, data=res.cat.cerebrovasc
```

cox.cat.cerebrovasc.6<-coxph(Surv(PERSON_YEARS, D008)~EVA_index, data=res.cat.cerebrovasc.

```
c.stat.cerebrovasc.1<-paste0(round(cox.cat.cerebrovasc.1$concordance[6],2)," (",round(cox.
c.stat.cerebrovasc.2<-paste0(round(cox.cat.cerebrovasc.2$concordance[6],2)," (",round(cox.</pre>
c.stat.cerebrovasc.3<-paste0(round(cox.cat.cerebrovasc.3$concordance[6],2)," (",round(cox.
c.stat.cerebrovasc.4<-paste0(round(cox.cat.cerebrovasc.4$concordance[6],2)," (",round(cox.</pre>
c.stat.cerebrovasc.5<-paste0(round(cox.cat.cerebrovasc.5$concordance[6],2)," (",round(cox.
c.stat.cerebrovasc.6<-paste0(round(cox.cat.cerebrovasc.6$concordance[6],2)," (",round(cox.</pre>
#Other
res.cat.other.1 <- surv_categorize(res.cut.other.1)</pre>
res.cat.other.2 <- surv_categorize(res.cut.other.2)</pre>
res.cat.other.3 <- surv_categorize(res.cut.other.3)</pre>
res.cat.other.4 <- surv_categorize(res.cut.other.4)</pre>
res.cat.other.5 <- surv_categorize(res.cut.other.5)</pre>
res.cat.other.6 <- surv_categorize(res.cut.other.6)</pre>
cox.cat.other.1<-coxph(Surv(PERSON_YEARS, D009)~METSVF, data=res.cat.other.1)
cox.cat.other.2<-coxph(Surv(PERSON_YEARS, D009)~DAAT_index, data=res.cat.other.2)</pre>
cox.cat.other.3<-coxph(Surv(PERSON_YEARS, D009)~VAI_index, data=res.cat.other.3)
cox.cat.other.4<-coxph(Surv(PERSON_YEARS, D009)~VAI_GEA_index, data=res.cat.other.4)
cox.cat.other.5<-coxph(Surv(PERSON_YEARS, D009)~LAAP_index, data=res.cat.other.5)</pre>
cox.cat.other.6<-coxph(Surv(PERSON_YEARS, D009)~EVA_index, data=res.cat.other.6)
c.stat.other.1<-paste0(round(cox.cat.other.1$concordance[6],2)," (",round(cox.cat.other.1$
c.stat.other.2<-paste0(round(cox.cat.other.2$concordance[6],2)," (",round(cox.cat.other.2$
c.stat.other.3<-paste0(round(cox.cat.other.3$concordance[6],2)," (",round(cox.cat.other.3$
c.stat.other.4<-paste0(round(cox.cat.other.4$concordance[6],2)," (",round(cox.cat.other.4$
c.stat.other.5<-paste0(round(cox.cat.other.5$concordance[6],2)," (",round(cox.cat.other.5$</pre>
c.stat.other.6<-paste0(round(cox.cat.other.6$concordance[6],2)," (",round(cox.cat.other.6$
##Table of CutOffs
df.cut.1<-data.frame(index_1=c("METS-VF","DAAT","VAI","DAI","LAAP","EVA"),</pre>
                      cutoff_1=c(round(as.numeric(res.cut.all.1$cutpoint[1]),2),
                                 round(as.numeric(res.cut.all.2$cutpoint[1]),2),
                                 round(as.numeric(res.cut.all.3$cutpoint[1]),2),
                                 round(as.numeric(res.cut.all.4$cutpoint[1]),2),
                                 round(as.numeric(res.cut.all.5$cutpoint[1]),2),
                                 round(as.numeric(res.cut.all.6$cutpoint[1]),2)),
                      statistic_1=c(round(as.numeric(res.cut.all.1$cutpoint[2]),2),
                                    round(as.numeric(res.cut.all.2$cutpoint[2]),2),
                                    round(as.numeric(res.cut.all.3$cutpoint[2]),2),
```

```
round(as.numeric(res.cut.all.6$cutpoint[2]),2)),
                     CI_95_1=c(rbind(c.stat.all.1,
                                   c.stat.all.2,
                                   c.stat.all.3,
                                   c.stat.all.4,
                                   c.stat.all.5,
                                    c.stat.all.6)))
df.cut.2<-data.frame(index_2=c("METS-VF","DAAT","VAI","DAI","LAAP","EVA"),</pre>
                     cutoff_2=c(round(as.numeric(res.cut.cardiac.1$cutpoint[1]),2),
                                round(as.numeric(res.cut.cardiac.2$cutpoint[1]),2),
                                round(as.numeric(res.cut.cardiac.3$cutpoint[1]),2),
                                round(as.numeric(res.cut.cardiac.4$cutpoint[1]),2),
                                round(as.numeric(res.cut.cardiac.5$cutpoint[1]),2),
                                round(as.numeric(res.cut.cardiac.6$cutpoint[1]),2)),
                     statistic_2=c(round(as.numeric(res.cut.cardiac.1$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cardiac.2$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cardiac.3$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cardiac.4$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cardiac.5$cutpoint[2]),2),
                                    round(as.numeric(res.cut.cardiac.6$cutpoint[2]),2)),
                     CI_95_2=c(rbind(c.stat.cardiac.1,
                                   c.stat.cardiac.2,
                                   c.stat.cardiac.3,
                                    c.stat.cardiac.4,
                                    c.stat.cardiac.5,
                                    c.stat.cardiac.6)))
df.cut.3<-data.frame(index_3=c("METS-VF","DAAT","VAI","DAI","LAAP","EVA"),</pre>
                     cutoff_3=c(round(as.numeric(res.cut.cerebrovasc.1$cutpoint[1]),2),
                                round(as.numeric(res.cut.cerebrovasc.2$cutpoint[1]),2),
                                round(as.numeric(res.cut.cerebrovasc.3$cutpoint[1]),2),
                                round(as.numeric(res.cut.cerebrovasc.4$cutpoint[1]),2),
                                round(as.numeric(res.cut.cerebrovasc.5$cutpoint[1]),2),
                                round(as.numeric(res.cut.cerebrovasc.6$cutpoint[1]),2)),
                     statistic_3=c(round(as.numeric(res.cut.cerebrovasc.1$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cerebrovasc.2$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cerebrovasc.3$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cerebrovasc.4$cutpoint[2]),2),
```

round(as.numeric(res.cut.all.4\$cutpoint[2]),2),
round(as.numeric(res.cut.all.5\$cutpoint[2]),2),

```
round(as.numeric(res.cut.cerebrovasc.5$cutpoint[2]),2),
                                   round(as.numeric(res.cut.cerebrovasc.6$cutpoint[2]),2))
                     CI_95_3=c(rbind(c.stat.cerebrovasc.1,
                                   c.stat.cerebrovasc.2,
                                   c.stat.cerebrovasc.3,
                                   c.stat.cerebrovasc.4,
                                   c.stat.cerebrovasc.5,
                                   c.stat.cerebrovasc.6)))
df.cut.4<-data.frame(index_4=c("METS-VF","DAAT","VAI","DAI","LAAP","EVA"),</pre>
                     cutoff_4=c(round(as.numeric(res.cut.other.1$cutpoint[1]),2),
                                round(as.numeric(res.cut.other.2$cutpoint[1]),2),
                                round(as.numeric(res.cut.other.3$cutpoint[1]),2),
                                round(as.numeric(res.cut.other.4$cutpoint[1]),2),
                                round(as.numeric(res.cut.other.5$cutpoint[1]),2),
                                round(as.numeric(res.cut.other.6$cutpoint[1]),2)),
                     statistic_4=c(round(as.numeric(res.cut.other.1$cutpoint[2]),2),
                                   round(as.numeric(res.cut.other.2$cutpoint[2]),2),
                                   round(as.numeric(res.cut.other.3$cutpoint[2]),2),
                                   round(as.numeric(res.cut.other.4$cutpoint[2]),2),
                                   round(as.numeric(res.cut.other.5$cutpoint[2]),2),
                                   round(as.numeric(res.cut.other.6$cutpoint[2]),2)),
                     CI_95_4=c(rbind(c.stat.other.1,
                                   c.stat.other.2,
                                   c.stat.other.3,
                                   c.stat.other.4,
                                   c.stat.other.5,
                                   c.stat.other.6)))
df.cut < -cbind(df.cut.1, df.cut.2[,-1], df.cut.3[,-1], df.cut.4[,-1])
df.cut<-df.cut %>%
  rename("Indice de Grasa Visceral" = "index_1",
         "Punto de Corte\n(Cualquier Causa Cardiovasculares)" = "cutoff_1",
         "Estadístico-Log-Rank\n(Cualquier Causa Cardiovasculares)" = "statistic_1",
         "Estadístico-C\n(Cualquier Causa Cardiovasculares)" = "CI_95_1",
         "Punto de Corte\n(Afecciones Cardiacas)" = "cutoff_2",
         "Estadístico-Log-Rank\n(Afecciones Cardiacas)" = "statistic_2",
         "Estadístico-C\n(Afecciones Cardiacas)" = "CI_95_2",
         "Punto de Corte\n(Enfermedades Cerebrovasculares)" = "cutoff_3",
         "Estadístico-Log-Rank\n(Enfermedades Cerebrovasculares)" = "statistic_3",
```

```
"Estadístico-C\n(Enfermedades Cerebrovasculares)" = "CI_95_3",
"Punto de Corte\n(Otras Causas Cardiovasculares)" = "cutoff_4",
"Estadístico-Log-Rank\n(Otras Causas Cardiovasculares)" = "statistic_4",
"Estadístico-C\n(Otras Causas Cardiovasculares)" = "CI_95_4")
```

flextable::align(flextable::flextable(df.cut,cwidth=7),align="center",part="all")%>%flextable:

Indice de Grasa Visceral	Punto de Corte (Cualquier Causa Cardiovasculares)	Estadístico-Log-Rank (Cualquier Causa Cardiovasculares)
METS-VF	7.21	24.29
DAAT	145.21	13.39
VAI	2.01	1.99
DAI	1.24	2.61
LAAP	45.25	7.47
EVA	133.27	23.48