

ANEXO

I) Frecuencia de aparición de cada uno de los valores de las variables categóricas

Comprobación de la frecuencia de aparición de los valores de las variables categóricas

```
lapply(Filter(is.factor, datos), FUN = freq)
```

```
## $Genero
##      n      % val%
## Female 2963 46.6 49.4
## Male   3038 47.8 50.6
## NA      352  5.5  NA
##
## $Mayor65
##      n      % val%
## 0   5045 79.4 84.1
## 1    956 15.0 15.9
## NA   352  5.5  NA
##
## $Conyuge
##      n      % val%
## 0  3287 51.7 51.7
## 1  3066 48.3 48.3
##
## $PersCargo
##      n      % val%
## 0  4439 69.9 69.9
## 1  1914 30.1 30.1
##
## $Telf_serv
##      n      % val%
## 0    605  9.5  9.7
## 1   5656 89.0 90.3
## NA    92  1.4  NA
##
## $VariasLineas
##      n      % val%
## 0  3687  58   58
## 1  2666  42   42
##
## $Int_serv
##      n      % val%
## DSL      2122 33.4 34.4
## Fiber optic 2702 42.5 43.8
## No       1339 21.1 21.7
## NA       190  3.0  NA
##
## $Seguridad
##      n      % val%
## 0  4532 71.3 71.3
## 1  1821 28.7 28.7
##
## $CopiaSeguridad
##      n      % val%
## 0  4180 65.8 65.8
## 1  2173 34.2 34.2
##
## $Antivirus_disp
##      n      % val%
## 0  4180 65.8 65.8
## 1  2173 34.2 34.2
##
```

```
## $Soporte_tecnico
##      n      % val%
## 0 4509 71      71
## 1 1844 29      29
##
## $TV_streaming
##      n      % val%
## 0 3936 62      62
## 1 2417 38      38
##
## $Peliculas
##      n      % val%
## 0 3902 61.4 61.4
## 1 2451 38.6 38.6
##
## $Contrato
##              n      % val%
## Month-to-month 3285 51.7 55.1
## One year      1261 19.8 21.2
## Two year      1413 22.2 23.7
## NA              394  6.2   NA
##
## $Fact_sinPapel
##      n      % val%
## 0  2483 39.1 40.8
## 1  3609 56.8 59.2
## NA  261  4.1   NA
##
## $MetodoPago
##              n      % val%
## Bank transfer (automatic) 1283 20.2 21.8
## Credit card (automatic)   1290 20.3 21.9
## Electronic check           1972 31.0 33.5
## Mailed check               1336 21.0 22.7
## NA                         472  7.4   NA
##
## $Fuga
##      n      % val%
## 0 4667 73.5 73.5
## 1 1686 26.5 26.5
```

II) Observación de los estadísticos básicos de las variables predictoras e inspección gráfica de los datos depurados

```
# Observamos a detalle los valores de las variables
```

```
summary(input)
```

##	Genero	Mayor65	Conyuge	PersCargo	Antig.fc.edad	Telf_serv
##	Female:2963	0:5397	0:3287	0:4439	Min. : 0.00	0: 605
##	Male :3390	1: 956	1:3066	1:1914	1st Qu.:10.00	1:5748
##					Median :32.00	
##					Mean :32.38	
##					3rd Qu.:54.00	
##					Max. :72.00	
##	VariasLineas	Int_serv	Seguridad	CopiaSeguridad	Antivirus_disp	
##	0:3687	DSL	:2122	0:4532	0:4180	0:4180
##	1:2666	Fiber optic:	2892	1:1821	1:2173	1:2173
##		No	:1339			
##						
##						
##						
##	Soporte_tecnico	TV_streaming	Peliculas	Contrato	Fact_sinPapel	
##	0:4509	0:3936	0:3902	Month-to-month:3679	0:2483	
##	1:1844	1:2417	1:2451	One year :1261	1:3870	

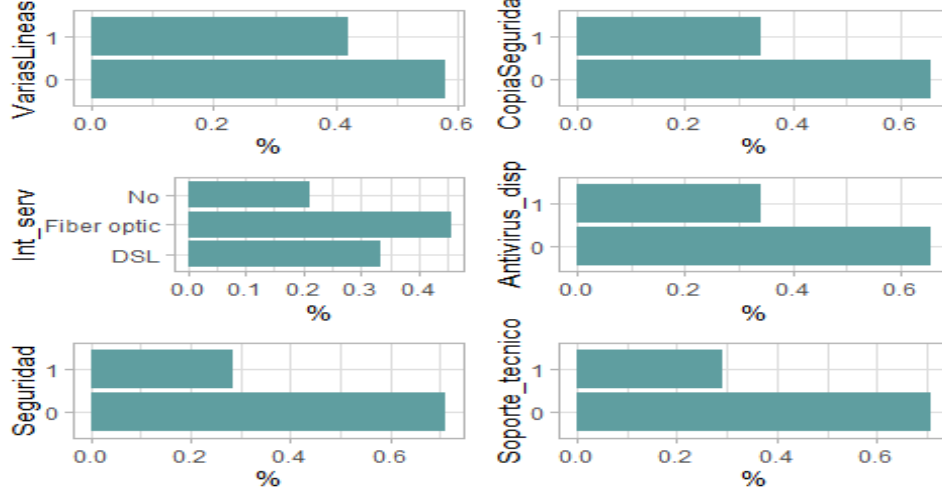
```
##
##
##
##
##      MetodoPago      FacturaMes      FacturaTotal
## Bank transfer (automatic):1283 Min. : 18.25 Min. : 18.8
## Credit card (automatic) :1290 1st Qu.: 41.10 1st Qu.: 396.3
## Electronic check :2444 Median : 68.75 Median :1387.3
## Mailed check :1336 Mean : 64.68 Mean :2268.7
##
##      3rd Qu.: 89.00 3rd Qu.:3778.0
##      Max. :118.75 Max. :8672.5
```

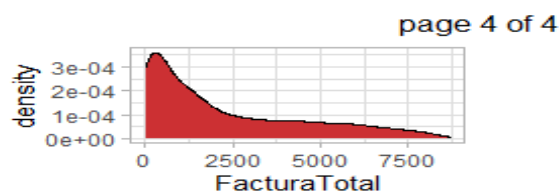
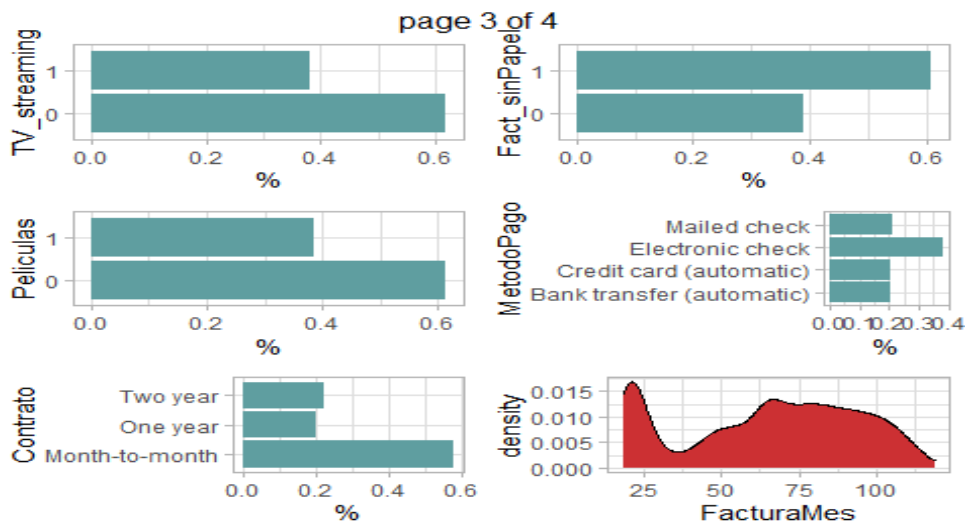
```
# Graficar el Dataset
par(mfrow = c(3,3))
lista_his <- dfplot_his(input)
gridExtra::marrangeGrob(lista_his, nrow=3, ncol=2)
```

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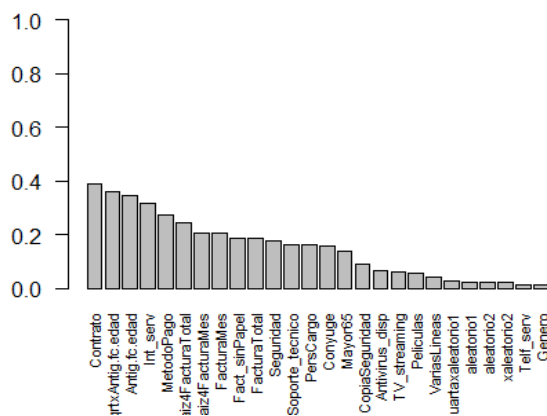




III) Influencia de las variables originales y transformadas sobre la variable objetivo

Evaluación sobre la importancia de las variables predictoras sobre la variable objetivo

```
graficoVcramer(input_obj, varObj)
```



IV) Modelo de referencia en el modelado manual de variables originales

Primer modelo sin las variables transformadas

```
modeloInicial <- glm(varObj~., data = data_train[,c(1:21, 27)], family = binomial)
summary(modeloInicial)
```

```
##
## Call:
## glm(formula = varObj ~ ., family = binomial, data = data_train[,
##      c(1:21, 27)])
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9365  -0.6693  -0.2951   0.7121   3.2456
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -3.556e-01  2.537e-01  -1.402  0.161037
## GeneroMale    2.993e-02  7.657e-02   0.391  0.695855
## Mayor651     1.425e-01  1.021e-01   1.395  0.162938
## Conyuge1     -7.541e-02  9.119e-02  -0.827  0.408290
## PersCargo1   -1.943e-01  1.052e-01  -1.846  0.064887 .
## Antig.fc.edad -3.055e-02  4.703e-03  -6.495  8.29e-11 ***
## Telf_serv1    -5.350e-01  1.750e-01  -3.057  0.002234 **
## VariasLineas1 1.565e-01  9.651e-02   1.621  0.104999
## Int_servFiber optic 9.115e-01  1.465e-01   6.223  4.87e-10 ***
## Int_servNo    -8.179e-01  1.906e-01  -4.291  1.78e-05 ***
## Seguridad1   -4.153e-01  1.034e-01  -4.015  5.95e-05 ***
## CopiaSeguridad1 -1.737e-01  9.457e-02  -1.837  0.066238 .
## Antivirus_disp1 -6.495e-02  9.633e-02  -0.674  0.500196
## Soporte_tecnico1 -3.417e-01  1.027e-01  -3.328  0.000875 ***
## TV_streaming1 2.551e-01  1.056e-01   2.415  0.015731 *
## Peliculas1    2.294e-01  1.043e-01   2.200  0.027804 *
## ContratoOne year -7.581e-01  1.247e-01  -6.079  1.21e-09 ***
## ContratoTwo year -1.619e+00  2.162e-01  -7.490  6.91e-14 ***
## Fact_sinPapell 3.923e-01  8.711e-02   4.503  6.69e-06 ***
## MetodoPagoCredit card (automatic) -6.882e-02  1.401e-01  -0.491  0.623264
## MetodoPagoElectronic check 3.815e-01  1.124e-01   3.393  0.000690 ***
## MetodoPagoMailed check 1.213e-01  1.364e-01   0.889  0.373942
## FacturaMes     2.083e-03  4.697e-03   0.444  0.657361
## FacturaTotal   -6.154e-06  5.616e-05  -0.110  0.912741
## aleatorio1     1.437e-01  1.316e-01   1.091  0.275139
## aleatorio2    -7.623e-02  1.316e-01  -0.579  0.562316
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 5882.3  on 5082  degrees of freedom
## Residual deviance: 4233.1  on 5057  degrees of freedom
## AIC: 4285.1
##
## Number of Fisher Scoring iterations: 6
```

```

# Volver a ajustar el modelo inicial
modeloInicial <- glm(varObj~., data = data_train[,c(1:21, 27)], family=binomial)
summary(modeloInicial)

##
## Call:
## glm(formula = varObj ~ ., family = binomial, data = data_train[,
##      c(1:21, 27)])
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9314  -0.6691  -0.2952   0.7208   3.2513
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -4.488e-01  2.408e-01  -1.864 0.062351 .
## GeneroMale       2.750e-02  7.647e-02   0.360 0.719122
## Mayor651        1.552e-01  1.019e-01   1.523 0.127772
## Conyuge1       -6.675e-02  9.102e-02  -0.733 0.463352
## PersCargo1     -2.051e-01  1.050e-01  -1.952 0.050920 .
## Antig.fc.edad  -3.008e-02  4.701e-03  -6.397 1.58e-10 ***
## Telf_serv1     -5.533e-01  1.743e-01  -3.173 0.001507 **
## VariasLineas1   1.615e-01  9.641e-02   1.676 0.093819 .
## Int_servFiber optic  9.316e-01  1.459e-01   6.386 1.70e-10 ***
## Int_servNo     -8.580e-01  1.896e-01  -4.526 6.00e-06 ***
## Seguridad1     -4.273e-01  1.032e-01  -4.142 3.45e-05 ***
## CopiaSeguridad1 -1.735e-01  9.448e-02  -1.837 0.066260 .
## Antivirus_disp1 -7.233e-02  9.613e-02  -0.752 0.451781
## Soporte_tecnico1 -3.579e-01  1.024e-01  -3.495 0.000474 ***
## TV_streaming1   2.599e-01  1.054e-01   2.465 0.013685 *
## Peliculas1      2.404e-01  1.040e-01   2.310 0.020876 *
## ContratoOne year -7.623e-01  1.247e-01  -6.115 9.67e-10 ***
## ContratoTwo year -1.627e+00  2.160e-01  -7.530 5.09e-14 ***
## Fact_sinPapel1   4.128e-01  8.664e-02   4.765 1.89e-06 ***
## MetodoPagoNot Automatic Payments 3.449e-01  8.637e-02   3.994 6.50e-05 ***
## FacturaMes       2.758e-03  4.674e-03   0.590 0.555214
## FacturaTotal    -1.052e-05  5.612e-05  -0.187 0.851283
## aleatorio1       1.386e-01  1.313e-01   1.055 0.291324
## aleatorio2      -7.133e-02  1.314e-01  -0.543 0.587106
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 5882.3  on 5082  degrees of freedom
## Residual deviance: 4238.8  on 5059  degrees of freedom
## AIC: 4286.8
##
## Number of Fisher Scoring iterations: 6

# Segundo modelo
modelo2 <- update(modeloInicial,.~.-FacturaTotal-aleatorio2-Genero-FacturaMes-
Conyuge-Antivirus_disp-aleatorio1)

# Tercer modelo
modelo3 <- update(modelo2,.~.-Mayor65)

# Cuarto modelo
modelo4 <- update(modelo3,.~.-CopiaSeguridad)

# Quinto modelo
modelo5 <- update(modelo4,.~.-VariasLineas)

# Sexto modelo
modelo6 <- update(modelo5,.~.-Peliculas)

```

```

# Séptimo modelo
modelo7 <- update(modelo6, .~.-PersCargo)

# Octavo modelo
modelo8 <- update(modelo7, .~.-Telf_serv)

# Noveno modelo
modelo9 <- update(modelo8, .~.-Soporte_tecnico)
# Décimo modelo
modelo10 <- update(modelo9, .~.-TV_streaming)

# Undécimo modelo
modelo11 <- update(modelo10, .~.-MetodoPago)

# Duodécimo modelo
modelo12 <- update(modelo11, .~.-Seguridad)

# Decimotercer modelo
modelo13 <- update(modelo12, .~.-Fact_sinPapel)

```

V) Parámetros del modelo para su interpretación final

```

modFinal <- glm(formula(modeloManual), data = todo_obj, family = binomial)

coef(modFinal)

##              (Intercept)              Antig.fc.edad
##              -0.7029580              -0.0311027
##              Int_servFiber optic              Int_servNo
##              0.9340172              -1.0611609
##              Seguridad1              Soporte_tecnico1
##              -0.4592007              -0.4041569
##              TV_streaming1              ContratoOne year
##              0.4213377              -0.7793340
##              ContratoTwo year              Fact_sinPapel1
##              -1.5827798              0.4387844
## MetodoPagoNot Automatic Payments
##              0.3312746

logistic.display(modFinal)

##
## Logistic regression predicting varObj : 1 vs 0
##
##              crude OR(95%CI)
## Antig.fc.edad (cont. var.)              0.96 (0.96,0.96)
##
## Int_serv: ref.=DSL
##   Fiber optic              3.07 (2.69,3.5)
##   No              0.36 (0.28,0.45)
##
## Seguridad: 1 vs 0              0.36 (0.32,0.42)
##
## Soporte_tecnico: 1 vs 0              0.4 (0.34,0.46)
##
## TV_streaming: 1 vs 0              1.31 (1.17,1.47)
##
## Contrato: ref.=Month-to-month
##   One year              0.18 (0.15,0.21)
##   Two year              0.04 (0.03,0.05)
##
## Fact_sinPapel: 1 vs 0              2.53 (2.23,2.87)
##
## MetodoPago: Not Automatic Payments vs Automatic Payments 2.73 (2.41,3.1)
##
##              adj. OR(95%CI)

```

```

## Antig.fc.edad (cont. var.) 0.97 (0.97,0.97)
##
## Int_serv: ref.=DSL
##   Fiber optic 2.54 (2.18,2.98)
##   No 0.35 (0.27,0.45)
##
## Seguridad: 1 vs 0 0.63 (0.53,0.75)
##
## Soporte_tecnico: 1 vs 0 0.67 (0.56,0.79)
##
## TV_streaming: 1 vs 0 1.52 (1.31,1.78)
##
## Contrato: ref.=Month-to-month
##   One year 0.46 (0.37,0.57)
##   Two year 0.21 (0.14,0.3)
##
## Fact_sinPapel: 1 vs 0 1.55 (1.33,1.8)
##
## MetodoPago: Not Automatic Payments vs Automatic Payments 1.39 (1.2,1.62)
##
## P(Wald's test)
## Antig.fc.edad (cont. var.) < 0.001
##
## Int_serv: ref.=DSL
##   Fiber optic < 0.001
##   No < 0.001
##
## Seguridad: 1 vs 0 < 0.001
##
## Soporte_tecnico: 1 vs 0 < 0.001
##
## TV_streaming: 1 vs 0 < 0.001
##
## Contrato: ref.=Month-to-month
##   One year < 0.001
##   Two year < 0.001
##
## Fact_sinPapel: 1 vs 0 < 0.001
##
## MetodoPago: Not Automatic Payments vs Automatic Payments < 0.001
##
## P(LR-test)
## Antig.fc.edad (cont. var.) < 0.001
##
## Int_serv: ref.=DSL
##   Fiber optic < 0.001
##   No
##
## Seguridad: 1 vs 0 < 0.001
##
## Soporte_tecnico: 1 vs 0 < 0.001
##
## TV_streaming: 1 vs 0 < 0.001
##
## Contrato: ref.=Month-to-month
##   One year
##   Two year
##
## Fact_sinPapel: 1 vs 0 < 0.001
##
## MetodoPago: Not Automatic Payments vs Automatic Payments < 0.001
##
## Log-likelihood = -2674.2628
## No. of observations = 6353
## AIC value = 5370.5256

```