SEMINARIO ANALISE MULTIVARIADA

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ANÁLISE MULTIVARIADA NO BANCO DE DADOS SOBRE FRAUD NO CARTÃO DE CRÉDITO

BANCO DE DADOS COLETADO NO LINK:

HTTPS://WWW.KAGGLE.COM/CODE/SMNURUZZAMAN/FRAUD-DETECTION-WITH-SMOTE-AND-SHAP-XGB-99-99/INPUT

COM 6362620 OBSERVAÇÕES E 12 VARIÁVEIS, DAS QUAIS 5 SÃO NÚMERICAS FORAM SELECIONADAS DUAS DELAS

IMPORTANTANDO O BANCO DE DADOS

SUMARIO SOBRE AS VARIÁVEIS

> summary(dados)								
step	type	amount	nam	neOrig	oldbalance	eOrg	newbalance	eOrig
Min. : 1.0	Length: 6362620	Min. :	0 Lengt	:h:6362620	Min. :	0	Min. :	0
1st Qu.:156.0	Class :character	1st Qu.: 1339	00 Class	:character	1st Qu.:	0	1st Qu.:	0
Median :239.0	Mode :character	Median: 7487	72 Mode	:character	Median :	14208	Median :	0
Mean :243.4		Mean : 17986	52		Mean :	833883	Mean :	855114
3rd Qu.:335.0		3rd Qu.: 20872	21		3rd Qu.:	107315	3rd Qu.:	144258
Max. :743.0		Max. :9244551	.7		Max. :59	9585040	Max. :49	9585040
nameDest	oldbalanceDest	newbalanceD	est)	isFraud	isFla	aggedFraud	d	D2
Length: 6362620	Min. :	0 Min. :		Min. :0.000	000 Min.	:0.0e+0	90 Min.	: 0.0000
Class :character	1st Qu.:	0 1st Qu.:	0	1st Qu.:0.000	000 1st (Qu.:0.0e+0	90 1st Qı	u.: 0.7961
Mode :character	Median : 1327	06 Median:	214661	Median:0.000	000 Media	an :0.0e+0	90 Media	n : 1.5090
	Mean : 110070	92 Mean : 1	.224996	Mean :0.001	291 Mean	:2.5e-0	96 Mean	: 2.0000
	3rd Qu.: 94303	37 3rd Qu.: 1	111909	3rd Qu.:0.000	000 3rd (Qu.:0.0e+0	90 3rd Qu	u.: 2.6076
	Max. :3560158	39 Max. :356	179279	Max. :1.000	000 Max.	:1.0e+0	90 Max.	:41.6851

÷	step ^	type [‡]	amount [‡]	nameOrig [‡]	oldbalanceOrg [‡]	newbalanceOrig [‡]	nameDest [‡]	oldbalanceDest [‡]	newbalanceDest $^{\scriptsize \scriptsize igoplus}$	isFraud [‡]	isFlaggedFraud [‡]
1	1	PAYMENT	9839.64	C1231006815	170136.00	160296.36	M1979787155	0.00	0.00	0	0
2	1	PAYMENT	1864.28	C1666544295	21249.00	19384.72	M2044282225	0.00	0.00	0	0
3	1	TRANSFER	181.00	C1305486145	181.00	0.00	C553264065	0.00	0.00	1	0
4	1	CASH_OUT	181.00	C840083671	181.00	0.00	C38997010	21182.00	0.00	1	0
5	1	PAYMENT	11668.14	C2048537720	41554.00	29885.86	M1230701703	0.00	0.00	0	0
6	1	PAYMENT	7817.71	C90045638	53860.00	46042.29	M573487274	0.00	0.00	0	0
7	1	PAYMENT	7107.77	C154988899	183195.00	176087.23	M408069119	0.00	0.00	0	0
8	1	PAYMENT	7861.64	C1912850431	176087.23	168225.59	M633326333	0.00	0.00	0	0
9	1	PAYMENT	4024.36	C1265012928	2671.00	0.00	M1176932104	0.00	0.00	0	0
10	1	DEBIT	5337.77	C712410124	41720.00	36382.23	C195600860	41898.00	40348.79	0	0

SELECIONADO AS VARIÁVEIS PARA OS TESTES

FOI SELECIONADO AS VARIÁVEIS "AMOUNT" E "OLDBALANCEORG"

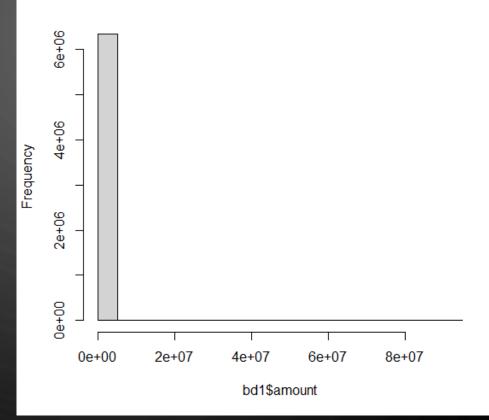
```
> bd = dados[c("amount", "oldbalanceOrg")]
```

> bd = na.omit(bd) #removendo os NA

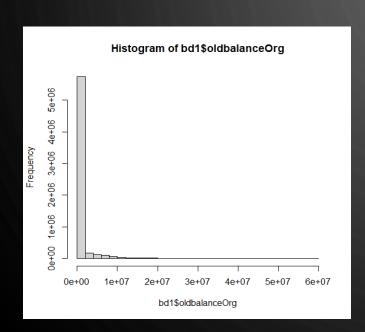
TESTES DE NORMALIDADE

PRIMEIRAMENTE FOI FEITO O HISTOGRAMAS DE AMBAS AS VARIÁVEIS

Histogram of bd1\$amount



HISTOGRAMA OLDBALANCEORG



Como se pode observar nenhuma das duas tem comportamento de seguir uma distribuição normal

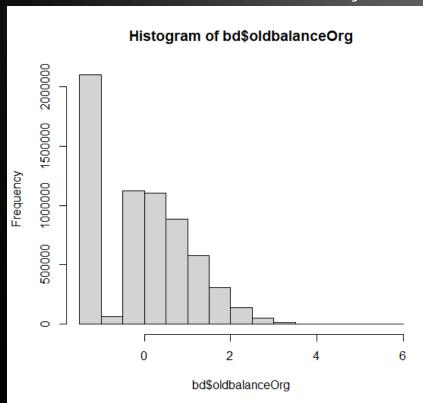
TRANSFORMAÇÃO PARA VARIÁVEL NORMAL

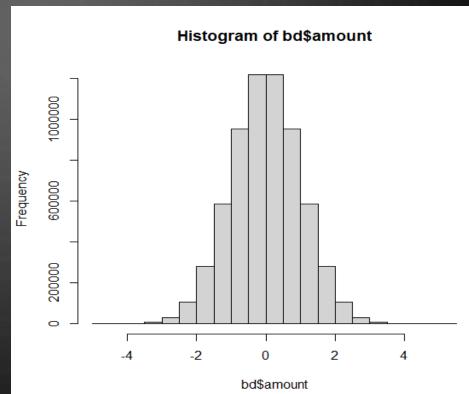
UTILIZANDO O PACOTE BESTNORMALIZE PARA FAZER AS TRANSFORMAÇÃO DAS VARIAVEIS AMOUNT E OLDBALANCEORG

BD\$AMOUNT = BESTNORMALIZE::BESTNORMALIZE(BD\$AMOUNT)\$X.T

BD\$OLDBALANCEORG = BESTNORMALIZE::BESTNORMALIZE(BD\$OLDBALANCEORG)\$X.T

APÓS AS TRANSFORMAÇÕES FORAM FEITOS NOVOS HISTOGRAMAS.

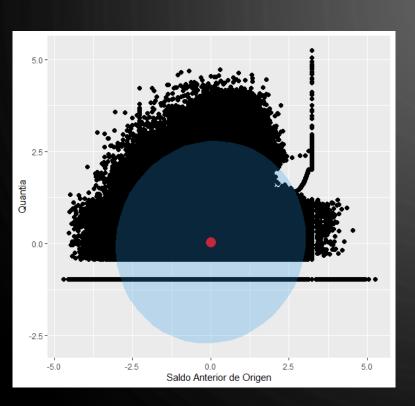




VETORES DE MÉDIA E MATRIZ DE CORRELAÇÃO E

```
> # Vetor de medias
> Vmean = colMeans(bd)
> Vmean
      amount oldbalanceOrg
-3.519306e-06 3.390802e-17
> # matriz de covariancia
> Mcov = cov(bd)
> Mcov
                 amount oldbalanceOrg
amount
            0.99997264
                           0.04671722
oldbalanceOrg 0.04671722 1.00000000
> ## Matriz de correlação
> cor(bd)
                 amount oldbalanceOrg
             1.00000000
                           0.04671786
amount
oldbalanceOrg 0.04671786
                           1.00000000
```

SCATTER PLOT



Como observado na Covariância é praticamente 0 assim não foi formado uma elipse e ser praticamente um círculo com um nível de confiança de 0.99

CALCULANDO A DISTANCIA DE MAHALANOBIS

CALCULAMOS A DISTANCIA DE MAHALANOBIS UTILIZANDO A FUNÇÃO MAHALANOBIS(BANCO DADOS, MÉDIA, COVARIÂNCIA) COMO SENDO AS VARIÁVEIS DA FUNÇÃO E FOI FEITO TAMBÉM O PONTO DE CORTE COMO A FUNÇÃO QCHISQ.

D2 <- MAHALANOBIS(BD, VMEAN, MCOV);D2

DADOSD2<-D2 PONTO_CORTE = QCHISQ(P = 0.01, DF = 2, LOWER.TAIL = F)

COMPARANDO AS MAIORES DISTANCIAS

				7 11 7 -	7 - 1		7 11 7	1 1	
step	type	amount	name0r1g	oldbalanceOrg	newbalanceOr1g	nameDest	oldbalanceDest	newbalanceDest	ısFraud
425	TRANSFER	1e+07	C40489106	59585040	49585040	C650095152	0	0	1
730	TRANSFER	1e+07	C726730575	57316255	47316255	C1364745638	0	0	1
646	TRANSFER	1e+07	C590657619	50399045	40399045	C1971187430	0	0	1
425	TRANSFER	1e+07	C1551381510	49585040	39585040	C1042012237	0	0	1
730	TRANSFER	1e+07	C507645439	47316255	37316255	C270374999	0	0	1
741	TRANSFER	1e+07	C780743034	45674548	35674548	C491519946	0	0	1
isFla	uggedFraud	1	D2						
	C	41.685	509						
	C	39.186	580						
	C	38.028	874						
	C	37.267	728						
	C	36.699	927						
	C	36.246	519						
	425 730 646 425 730 741	425 TRANSFER 730 TRANSFER 646 TRANSFER 425 TRANSFER 730 TRANSFER 741 TRANSFER isFlaggedFrauc	425 TRANSFER 1e+07 730 TRANSFER 1e+07 646 TRANSFER 1e+07 425 TRANSFER 1e+07 730 TRANSFER 1e+07 741 TRANSFER 1e+07 isFlaggedFraud 0 41.689 0 39.180 0 38.029 0 37.260 0 36.699	425 TRANSFER 1e+07 C40489106 730 TRANSFER 1e+07 C726730575 646 TRANSFER 1e+07 C590657619 425 TRANSFER 1e+07 C1551381510 730 TRANSFER 1e+07 C507645439 741 TRANSFER 1e+07 C780743034	425 TRANSFER 1e+07 C40489106 59585040 730 TRANSFER 1e+07 C726730575 57316255 646 TRANSFER 1e+07 C590657619 50399045 425 TRANSFER 1e+07 C1551381510 49585040 730 TRANSFER 1e+07 C507645439 47316255 741 TRANSFER 1e+07 C780743034 45674548 isFlaggedFraud D2 0 41.68509 0 39.18680 0 38.02874 0 37.26728 0 36.69927	425 TRANSFER 1e+07 C40489106 59585040 49585040 730 TRANSFER 1e+07 C726730575 57316255 47316255 646 TRANSFER 1e+07 C590657619 50399045 40399045 425 TRANSFER 1e+07 C1551381510 49585040 39585040 730 TRANSFER 1e+07 C507645439 47316255 37316255 741 TRANSFER 1e+07 C780743034 45674548 35674548 isFlaggedFraud D2 0 41.68509 0 39.18680 0 38.02874 0 37.26728 0 36.69927	425 TRANSFER 1e+07 C40489106 59585040 49585040 C650095152 730 TRANSFER 1e+07 C726730575 57316255 47316255 C1364745638 646 TRANSFER 1e+07 C590657619 50399045 40399045 C1971187430 425 TRANSFER 1e+07 C1551381510 49585040 39585040 C1042012237 730 TRANSFER 1e+07 C507645439 47316255 37316255 C270374999 741 TRANSFER 1e+07 C780743034 45674548 35674548 C491519946 isFlaggedFraud D2 0 41.68509 0 39.18680 0 38.02874 0 37.26728 0 36.69927	425 TRANSFER 1e+07 C40489106 59585040 49585040 C650095152 0 730 TRANSFER 1e+07 C726730575 57316255 47316255 C1364745638 0 646 TRANSFER 1e+07 C590657619 50399045 40399045 C1971187430 0 425 TRANSFER 1e+07 C1551381510 49585040 39585040 C1042012237 0 730 TRANSFER 1e+07 C507645439 47316255 37316255 C270374999 0 741 TRANSFER 1e+07 C780743034 45674548 35674548 C491519946 0 isFlaggedFraud D2 0 41.68509 0 39.18680 0 38.02874 0 37.26728 0 36.69927	425 TRANSFER 1e+07 C40489106 59585040 49585040 C650095152 0 0 730 TRANSFER 1e+07 C726730575 57316255 47316255 C1364745638 0 0 646 TRANSFER 1e+07 C590657619 50399045 40399045 C1971187430 0 0 425 TRANSFER 1e+07 C1551381510 49585040 39585040 C1042012237 0 0 730 TRANSFER 1e+07 C507645439 47316255 37316255 C270374999 0 0 741 TRANSFER 1e+07 C780743034 45674548 35674548 C491519946 0 0 isFlaggedFraud D2 0 41.68509 0 39.18680 0 38.02874 0 37.26728 0 36.69927

CONCLUISÕES

COMO OBSERVADO NA TABELA ANTERIOR OS VALORES COM AS DISTANCIAS GRANDES TEM VALORES GRANDE ASSOCIADOS TAMBÉM O QUE APARENTEMENTE TEM UMA CORRELAÇÃO COM AS FRAUDES.