

CustomerChurn

June 21, 2019

0.1 Projeto com Feedback 4

0.2 Customer Churn em uma Operadora de Telecom

0.3 MARCIO DE LIMA

0.4

0.5 DICIONARIO DE DADOS - data/projeto4_telecom_treino.csv

Unnamed: 0 int64 state object account_length int64 area_code object international_plan object voice_mail_plan object number_vmail_messages int64 total_day_minutes float64 total_day_calls int64 total_day_charge float64 total_eve_minutes float64 total_eve_calls int64 total_eve_charge float64 total_night_minutes float64 total_night_calls int64 total_night_charge float64 total_intl_minutes float64 total_intl_calls int64 total_intl_charge float64 number_customer_service_calls int64 churn object => VARIÁVEL TARGET

0.6

```
In [1]: # Importando as bibliotecas
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
```

```
%matplotlib inline
warnings.filterwarnings("ignore")
```

```
In [2]: # Importando o arquivo
df = pd.read_csv('data/projeto4_telecom_treino.csv')
```

```
In [3]: df.head(10)
```

```
Out[3]:
```

	Unnamed: 0	state	account_length	area_code	international_plan	\
0	1	KS	128	area_code_415	no	
1	2	OH	107	area_code_415	no	
2	3	NJ	137	area_code_415	no	
3	4	OH	84	area_code_408	yes	
4	5	OK	75	area_code_415	yes	

5	6	AL	118	area_code_510	yes
6	7	MA	121	area_code_510	no
7	8	MO	147	area_code_415	yes
8	9	LA	117	area_code_408	no
9	10	WV	141	area_code_415	yes

	voice_mail_plan	number_vmail_messages	total_day_minutes	total_day_calls	\
0	yes	25	265.1	110	
1	yes	26	161.6	123	
2	no	0	243.4	114	
3	no	0	299.4	71	
4	no	0	166.7	113	
5	no	0	223.4	98	
6	yes	24	218.2	88	
7	no	0	157.0	79	
8	no	0	184.5	97	
9	yes	37	258.6	84	

	total_day_charge	...	total_eve_calls	total_eve_charge	\
0	45.07	...	99	16.78	
1	27.47	...	103	16.62	
2	41.38	...	110	10.30	
3	50.90	...	88	5.26	
4	28.34	...	122	12.61	
5	37.98	...	101	18.75	
6	37.09	...	108	29.62	
7	26.69	...	94	8.76	
8	31.37	...	80	29.89	
9	43.96	...	111	18.87	

	total_night_minutes	total_night_calls	total_night_charge	\
0	244.7	91	11.01	
1	254.4	103	11.45	
2	162.6	104	7.32	
3	196.9	89	8.86	
4	186.9	121	8.41	
5	203.9	118	9.18	
6	212.6	118	9.57	
7	211.8	96	9.53	
8	215.8	90	9.71	
9	326.4	97	14.69	

	total_intl_minutes	total_intl_calls	total_intl_charge	\
0	10.0	3	2.70	
1	13.7	3	3.70	
2	12.2	5	3.29	
3	6.6	7	1.78	
4	10.1	3	2.73	

5	6.3	6	1.70
6	7.5	7	2.03
7	7.1	6	1.92
8	8.7	4	2.35
9	11.2	5	3.02

	number_customer_service_calls	churn
0	1	no
1	1	no
2	0	no
3	2	no
4	3	no
5	0	no
6	3	no
7	0	no
8	1	no
9	0	no

[10 rows x 21 columns]

In [4]: df.dtypes

```
Out[4]: Unnamed: 0      int64
state      object
account_length      int64
area_code      object
international_plan  object
voice_mail_plan     object
number_vmail_messages      int64
total_day_minutes      float64
total_day_calls      int64
total_day_charge      float64
total_eve_minutes      float64
total_eve_calls      int64
total_eve_charge      float64
total_night_minutes      float64
total_night_calls      int64
total_night_charge      float64
total_intl_minutes      float64
total_intl_calls      int64
total_intl_charge      float64
number_customer_service_calls      int64
churn      object
dtype: object
```

In [5]: *#Dados estatísticos - analise descritiva*
df.describe()

```
Out[5]:      Unnamed: 0  account_length  number_vmail_messages  total_day_minutes  \
count  3333.00000  3333.000000      3333.000000      3333.000000
```

mean	1667.00000	101.064806	8.099010	179.775098
std	962.29855	39.822106	13.688365	54.467389
min	1.00000	1.000000	0.000000	0.000000
25%	834.00000	74.000000	0.000000	143.700000
50%	1667.00000	101.000000	0.000000	179.400000
75%	2500.00000	127.000000	20.000000	216.400000
max	3333.00000	243.000000	51.000000	350.800000

	total_day_calls	total_day_charge	total_eve_minutes	total_eve_calls	\
count	3333.000000	3333.000000	3333.000000	3333.000000	
mean	100.435644	30.562307	200.980348	100.114311	
std	20.069084	9.259435	50.713844	19.922625	
min	0.000000	0.000000	0.000000	0.000000	
25%	87.000000	24.430000	166.600000	87.000000	
50%	101.000000	30.500000	201.400000	100.000000	
75%	114.000000	36.790000	235.300000	114.000000	
max	165.000000	59.640000	363.700000	170.000000	

	total_eve_charge	total_night_minutes	total_night_calls	\
count	3333.000000	3333.000000	3333.000000	
mean	17.083540	200.872037	100.107711	
std	4.310668	50.573847	19.568609	
min	0.000000	23.200000	33.000000	
25%	14.160000	167.000000	87.000000	
50%	17.120000	201.200000	100.000000	
75%	20.000000	235.300000	113.000000	
max	30.910000	395.000000	175.000000	

	total_night_charge	total_intl_minutes	total_intl_calls	\
count	3333.000000	3333.000000	3333.000000	
mean	9.039325	10.237294	4.479448	
std	2.275873	2.791840	2.461214	
min	1.040000	0.000000	0.000000	
25%	7.520000	8.500000	3.000000	
50%	9.050000	10.300000	4.000000	
75%	10.590000	12.100000	6.000000	
max	17.770000	20.000000	20.000000	

	total_intl_charge	number_customer_service_calls
count	3333.000000	3333.000000
mean	2.764581	1.562856
std	0.753773	1.315491
min	0.000000	0.000000
25%	2.300000	1.000000
50%	2.780000	1.000000
75%	3.270000	2.000000
max	5.400000	9.000000

```
In [6]: #Tratamento dos dados NA - Missing
df.fillna(0)
```

```
Out[6]:
```

	Unnamed: 0	state	account_length	area_code	international_plan	\
0	1	KS	128	area_code_415	no	
1	2	OH	107	area_code_415	no	
2	3	NJ	137	area_code_415	no	
3	4	OH	84	area_code_408	yes	
4	5	OK	75	area_code_415	yes	
5	6	AL	118	area_code_510	yes	
6	7	MA	121	area_code_510	no	
7	8	MO	147	area_code_415	yes	
8	9	LA	117	area_code_408	no	
9	10	WV	141	area_code_415	yes	
10	11	IN	65	area_code_415	no	
11	12	RI	74	area_code_415	no	
12	13	IA	168	area_code_408	no	
13	14	MT	95	area_code_510	no	
14	15	IA	62	area_code_415	no	
15	16	NY	161	area_code_415	no	
16	17	ID	85	area_code_408	no	
17	18	VT	93	area_code_510	no	
18	19	VA	76	area_code_510	no	
19	20	TX	73	area_code_415	no	
20	21	FL	147	area_code_415	no	
21	22	CO	77	area_code_408	no	
22	23	AZ	130	area_code_415	no	
23	24	SC	111	area_code_415	no	
24	25	VA	132	area_code_510	no	
25	26	NE	174	area_code_415	no	
26	27	WY	57	area_code_408	no	
27	28	MT	54	area_code_408	no	
28	29	MO	20	area_code_415	no	
29	30	HI	49	area_code_510	no	
...
3303	3304	WI	114	area_code_415	no	
3304	3305	IL	71	area_code_510	yes	
3305	3306	IN	58	area_code_415	no	
3306	3307	AL	106	area_code_408	no	
3307	3308	OK	172	area_code_408	no	
3308	3309	IA	45	area_code_415	no	
3309	3310	VT	100	area_code_408	yes	
3310	3311	NY	94	area_code_415	no	
3311	3312	LA	128	area_code_415	no	
3312	3313	SC	181	area_code_408	no	
3313	3314	ID	127	area_code_408	no	
3314	3315	MO	89	area_code_415	no	
3315	3316	ME	149	area_code_415	no	

3316	3317	MS	103	area_code_510	no
3317	3318	SD	163	area_code_415	yes
3318	3319	OK	52	area_code_415	no
3319	3320	WY	89	area_code_415	no
3320	3321	GA	122	area_code_510	yes
3321	3322	VT	60	area_code_415	no
3322	3323	MD	62	area_code_408	no
3323	3324	IN	117	area_code_415	no
3324	3325	WV	159	area_code_415	no
3325	3326	OH	78	area_code_408	no
3326	3327	OH	96	area_code_415	no
3327	3328	SC	79	area_code_415	no
3328	3329	AZ	192	area_code_415	no
3329	3330	WV	68	area_code_415	no
3330	3331	RI	28	area_code_510	no
3331	3332	CT	184	area_code_510	yes
3332	3333	TN	74	area_code_415	no

	voice_mail_plan	number_vmail_messages	total_day_minutes \
0	yes	25	265.1
1	yes	26	161.6
2	no	0	243.4
3	no	0	299.4
4	no	0	166.7
5	no	0	223.4
6	yes	24	218.2
7	no	0	157.0
8	no	0	184.5
9	yes	37	258.6
10	no	0	129.1
11	no	0	187.7
12	no	0	128.8
13	no	0	156.6
14	no	0	120.7
15	no	0	332.9
16	yes	27	196.4
17	no	0	190.7
18	yes	33	189.7
19	no	0	224.4
20	no	0	155.1
21	no	0	62.4
22	no	0	183.0
23	no	0	110.4
24	no	0	81.1
25	no	0	124.3
26	yes	39	213.0
27	no	0	134.3
28	no	0	190.0

29	no	0	119.3
...
3303	yes	26	137.1
3304	no	0	186.1
3305	yes	22	224.1
3306	yes	29	83.6
3307	no	0	203.9
3308	no	0	211.3
3309	no	0	219.4
3310	no	0	190.4
3311	no	0	147.7
3312	no	0	229.9
3313	no	0	102.8
3314	no	0	178.7
3315	yes	18	148.5
3316	yes	29	164.1
3317	no	0	197.2
3318	no	0	124.9
3319	no	0	115.4
3320	no	0	140.0
3321	no	0	193.9
3322	no	0	321.1
3323	no	0	118.4
3324	no	0	169.8
3325	no	0	193.4
3326	no	0	106.6
3327	no	0	134.7
3328	yes	36	156.2
3329	no	0	231.1
3330	no	0	180.8
3331	no	0	213.8
3332	yes	25	234.4

	total_day_calls	total_day_charge	...	total_eve_calls	\
0	110	45.07	...	99	
1	123	27.47	...	103	
2	114	41.38	...	110	
3	71	50.90	...	88	
4	113	28.34	...	122	
5	98	37.98	...	101	
6	88	37.09	...	108	
7	79	26.69	...	94	
8	97	31.37	...	80	
9	84	43.96	...	111	
10	137	21.95	...	83	
11	127	31.91	...	148	
12	96	21.90	...	71	
13	88	26.62	...	75	

14	70	20.52	...	76
15	67	56.59	...	97
16	139	33.39	...	90
17	114	32.42	...	111
18	66	32.25	...	65
19	90	38.15	...	88
20	117	26.37	...	93
21	89	10.61	...	121
22	112	31.11	...	99
23	103	18.77	...	102
24	86	13.79	...	72
25	76	21.13	...	112
26	115	36.21	...	112
27	73	22.83	...	100
28	109	32.30	...	84
29	117	20.28	...	109
...
3303	88	23.31	...	125
3304	114	31.64	...	140
3305	127	38.10	...	85
3306	131	14.21	...	131
3307	109	34.66	...	123
3308	87	35.92	...	97
3309	112	37.30	...	102
3310	91	32.37	...	107
3311	94	25.11	...	83
3312	130	39.08	...	93
3313	128	17.48	...	95
3314	81	30.38	...	74
3315	106	25.25	...	106
3316	111	27.90	...	96
3317	90	33.52	...	113
3318	131	21.23	...	118
3319	99	19.62	...	115
3320	101	23.80	...	77
3321	118	32.96	...	110
3322	105	54.59	...	122
3323	126	20.13	...	97
3324	114	28.87	...	105
3325	99	32.88	...	88
3326	128	18.12	...	87
3327	98	22.90	...	68
3328	77	26.55	...	126
3329	57	39.29	...	55
3330	109	30.74	...	58
3331	105	36.35	...	84
3332	113	39.85	...	82

	total_eve_charge	total_night_minutes	total_night_calls \
0	16.78	244.7	91
1	16.62	254.4	103
2	10.30	162.6	104
3	5.26	196.9	89
4	12.61	186.9	121
5	18.75	203.9	118
6	29.62	212.6	118
7	8.76	211.8	96
8	29.89	215.8	90
9	18.87	326.4	97
10	19.42	208.8	111
11	13.89	196.0	94
12	8.92	141.1	128
13	21.05	192.3	115
14	26.11	203.0	99
15	27.01	160.6	128
16	23.88	89.3	75
17	18.55	129.6	121
18	18.09	165.7	108
19	13.56	192.8	74
20	20.37	208.8	133
21	14.44	209.6	64
22	6.20	181.8	78
23	11.67	189.6	105
24	20.84	237.0	115
25	23.55	250.7	115
26	16.24	182.7	115
27	13.22	102.1	68
28	21.95	181.5	102
29	18.28	178.7	90
...
3303	13.23	247.6	94
3304	16.88	206.5	80
3305	20.30	174.2	86
3306	17.33	229.5	73
3307	19.89	160.7	65
3308	14.08	265.9	72
3309	19.18	255.3	95
3310	7.82	224.8	108
3311	24.08	188.3	124
3312	12.27	262.4	110
3313	12.21	191.4	97
3314	19.86	131.9	120
3315	9.73	178.3	98
3316	18.62	220.3	108
3317	16.02	211.1	94
3318	25.54	192.5	106

3319	17.84	280.9	112
3320	16.69	120.1	133
3321	7.23	210.1	134
3322	22.57	180.5	72
3323	21.19	227.0	56
3324	16.80	193.7	82
3325	9.94	243.3	109
3326	24.21	178.9	92
3327	16.12	221.4	128
3328	18.32	279.1	83
3329	13.04	191.3	123
3330	24.55	191.9	91
3331	13.57	139.2	137
3332	22.60	241.4	77

	total_night_charge	total_intl_minutes	total_intl_calls	\
0	11.01	10.0	3	
1	11.45	13.7	3	
2	7.32	12.2	5	
3	8.86	6.6	7	
4	8.41	10.1	3	
5	9.18	6.3	6	
6	9.57	7.5	7	
7	9.53	7.1	6	
8	9.71	8.7	4	
9	14.69	11.2	5	
10	9.40	12.7	6	
11	8.82	9.1	5	
12	6.35	11.2	2	
13	8.65	12.3	5	
14	9.14	13.1	6	
15	7.23	5.4	9	
16	4.02	13.8	4	
17	5.83	8.1	3	
18	7.46	10.0	5	
19	8.68	13.0	2	
20	9.40	10.6	4	
21	9.43	5.7	6	
22	8.18	9.5	19	
23	8.53	7.7	6	
24	10.67	10.3	2	
25	11.28	15.5	5	
26	8.22	9.5	3	
27	4.59	14.7	4	
28	8.17	6.3	6	
29	8.04	11.1	1	
...	
3303	11.14	11.5	7	

3304	9.29	13.8	5
3305	7.84	11.5	7
3306	10.33	8.1	3
3307	7.23	17.8	4
3308	11.97	13.3	6
3309	11.49	12.0	4
3310	10.12	13.6	17
3311	8.47	6.9	5
3312	11.81	14.2	4
3313	8.61	10.0	5
3314	5.94	9.1	4
3315	8.02	6.5	4
3316	9.91	12.3	9
3317	9.50	7.8	8
3318	8.66	11.6	4
3319	12.64	15.9	6
3320	5.40	9.7	4
3321	9.45	13.2	8
3322	8.12	11.5	2
3323	10.22	13.6	3
3324	8.72	11.6	4
3325	10.95	9.3	4
3326	8.05	14.9	7
3327	9.96	11.8	5
3328	12.56	9.9	6
3329	8.61	9.6	4
3330	8.64	14.1	6
3331	6.26	5.0	10
3332	10.86	13.7	4

	total_intl_charge	number_customer_service_calls	churn
0	2.70	1	no
1	3.70	1	no
2	3.29	0	no
3	1.78	2	no
4	2.73	3	no
5	1.70	0	no
6	2.03	3	no
7	1.92	0	no
8	2.35	1	no
9	3.02	0	no
10	3.43	4	yes
11	2.46	0	no
12	3.02	1	no
13	3.32	3	no
14	3.54	4	no
15	1.46	4	yes
16	3.73	1	no

17	2.19	3	no
18	2.70	1	no
19	3.51	1	no
20	2.86	0	no
21	1.54	5	yes
22	2.57	0	no
23	2.08	2	no
24	2.78	0	no
25	4.19	3	no
26	2.57	0	no
27	3.97	3	no
28	1.70	0	no
29	3.00	1	no
...
3303	3.11	2	no
3304	3.73	4	yes
3305	3.11	2	no
3306	2.19	1	no
3307	4.81	4	no
3308	3.59	1	no
3309	3.24	4	no
3310	3.67	2	no
3311	1.86	2	no
3312	3.83	2	no
3313	2.70	1	no
3314	2.46	1	no
3315	1.76	0	no
3316	3.32	0	no
3317	2.11	1	no
3318	3.13	2	no
3319	4.29	3	no
3320	2.62	4	yes
3321	3.56	3	no
3322	3.11	4	yes
3323	3.67	5	yes
3324	3.13	1	no
3325	2.51	2	no
3326	4.02	1	no
3327	3.19	2	no
3328	2.67	2	no
3329	2.59	3	no
3330	3.81	2	no
3331	1.35	2	no
3332	3.70	0	no

[3333 rows x 21 columns]

In [7]: *#Tratamento e Transformação das colunas do DataFrame*

```

# remoção da coluna Unnamed: 0
df = df.drop('Unnamed: 0', axis=1)

#Colunas international_plan, voice_mail_plan, churn,
df['churn'] = df['churn'].map(lambda x: 1 if x == 'yes' else 0 )
df['voice_mail_plan'] = df['voice_mail_plan'].map(lambda x: 1 if x == 'yes' else 0 )
df['international_plan'] = df['international_plan'].map(lambda x: 1 if x == 'yes' else 0 )

# Coluna area_code
df['area_code'] = df['area_code'].map(lambda x: x.replace('area_code_', ''))

```

```

In [21]: #Transformando o tipo das colunas
df['area_code'] = df['area_code'].apply(pd.to_numeric, downcast='integer')
df['voice_mail_plan'] = df['voice_mail_plan'].apply(pd.to_numeric, downcast='integer')
df['international_plan'] = df['international_plan'].apply(pd.to_numeric, downcast='integer')
df['churn'] = df['churn'].apply(pd.to_numeric, downcast='integer')

```

```
df.dtypes
```

```

Out[21]: state                object
account_length              int64
area_code                  int64
international_plan          int64
voice_mail_plan             int64
number_vmail_messages       int64
total_day_minutes           float64
total_day_calls              int64
total_day_charge             float64
total_eve_minutes           float64
total_eve_calls              int64
total_eve_charge             float64
total_night_minutes         float64
total_night_calls           int64
total_night_charge           float64
total_intl_minutes          float64
total_intl_calls             int64
total_intl_charge           float64
number_customer_service_calls int64
churn                       int64
dtype: object

```

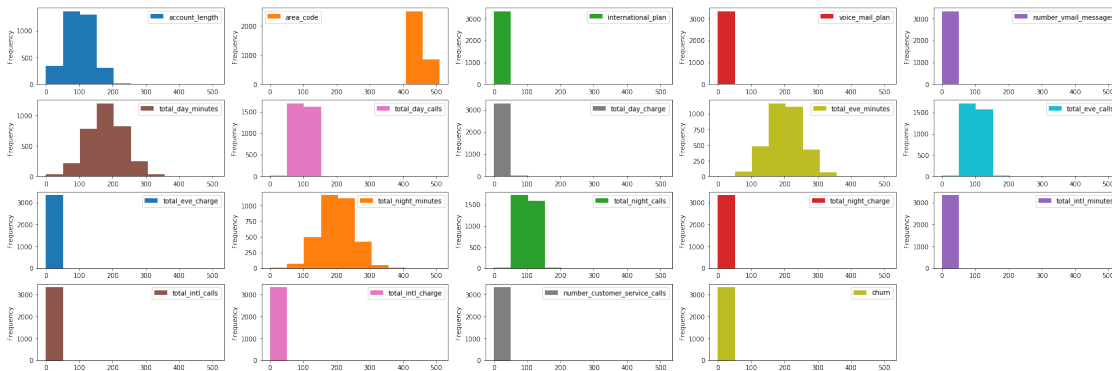
```
In [47]: df.shape
```

```
Out[47]: (3333, 20)
```

```
In [45]: #Gerando gráficos para análise das variaveis
```

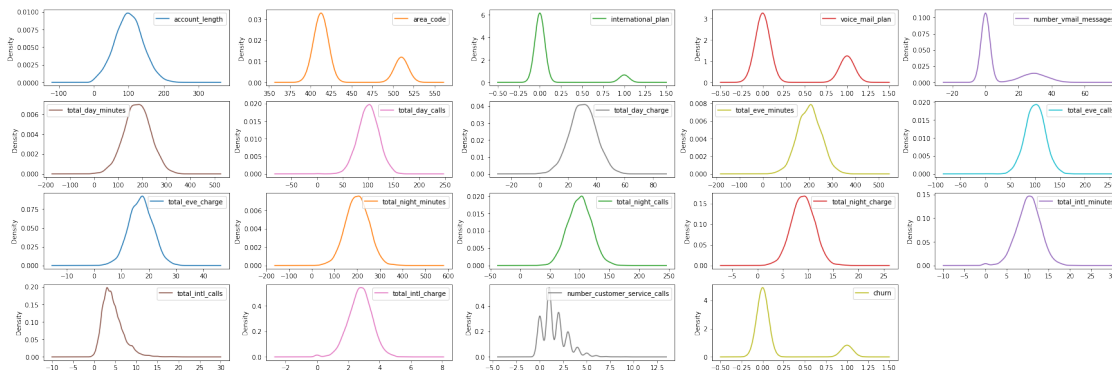
```
#Histogramas
```

```
df.plot(kind = 'hist', subplots = True, layout = (4,5), sharex = False, figsize=(30,10))
plt.show()
```



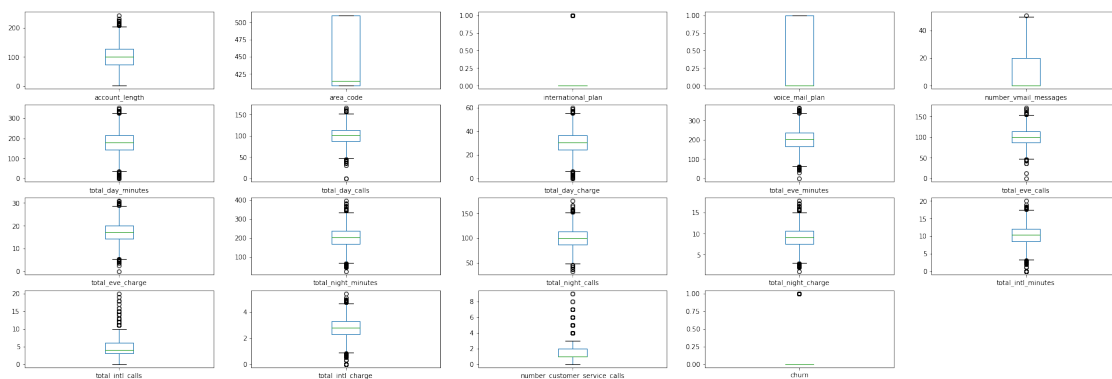
In [43]: #Gráfico de Densidade

```
df.plot(kind = 'density', subplots = True, layout = (4,5), sharex = False, figsize=(30,10))
plt.show()
```



In [42]: #Box-Plots

```
df.plot(kind = 'box', subplots = True, layout = (4,5), sharex = False, sharey = False)
plt.show()
```



```
In [46]: #Dados de Correlação entre as variáveis
df.corr()
```

```
Out [46]:
```

	account_length	area_code	international_plan \
account_length	1.000000	-0.012463	0.024735
area_code	-0.012463	1.000000	0.048551
international_plan	0.024735	0.048551	1.000000
voice_mail_plan	0.002918	-0.000747	0.006006
number_vmail_messages	-0.004628	-0.001994	0.008745
total_day_minutes	0.006216	-0.008264	0.049396
total_day_calls	0.038470	-0.009646	0.003755
total_day_charge	0.006214	-0.008264	0.049398
total_eve_minutes	-0.006757	0.003580	0.019100
total_eve_calls	0.019260	-0.011886	0.006114
total_eve_charge	-0.006745	0.003607	0.019106
total_night_minutes	-0.008955	-0.005825	-0.028905
total_night_calls	-0.013176	0.016522	0.012451
total_night_charge	-0.008960	-0.005845	-0.028913
total_intl_minutes	0.009514	-0.018288	0.045871
total_intl_calls	0.020661	-0.024179	0.017366
total_intl_charge	0.009546	-0.018395	0.045780
number_customer_service_calls	-0.003796	0.027572	-0.024522
churn	0.016541	0.006174	0.259852

	voice_mail_plan	number_vmail_messages \
account_length	0.002918	-0.004628
area_code	-0.000747	-0.001994
international_plan	0.006006	0.008745
voice_mail_plan	1.000000	0.956927
number_vmail_messages	0.956927	1.000000
total_day_minutes	-0.001684	0.000778
total_day_calls	-0.011086	-0.009548
total_day_charge	-0.001686	0.000776
total_eve_minutes	0.021545	0.017562
total_eve_calls	-0.006444	-0.005864
total_eve_charge	0.021559	0.017578
total_night_minutes	0.006079	0.007681
total_night_calls	0.015553	0.007123
total_night_charge	0.006064	0.007663
total_intl_minutes	-0.001318	0.002856
total_intl_calls	0.007618	0.013957
total_intl_charge	-0.001276	0.002884
number_customer_service_calls	-0.017824	-0.013263
churn	-0.102148	-0.089728

	total_day_minutes	total_day_calls \
account_length	0.006216	0.038470
area_code	-0.008264	-0.009646
international_plan	0.049396	0.003755
voice_mail_plan	-0.001684	-0.011086
number_vmail_messages	0.000778	-0.009548
total_day_minutes	1.000000	0.006750
total_day_calls	0.006750	1.000000
total_day_charge	1.000000	0.006753
total_eve_minutes	0.007043	-0.021451
total_eve_calls	0.015769	0.006462
total_eve_charge	0.007029	-0.021449
total_night_minutes	0.004323	0.022938
total_night_calls	0.022972	-0.019557
total_night_charge	0.004300	0.022927
total_intl_minutes	-0.010155	0.021565
total_intl_calls	0.008033	0.004574
total_intl_charge	-0.010092	0.021666
number_customer_service_calls	-0.013423	-0.018942
churn	0.205151	0.018459

	total_day_charge	total_eve_minutes \
account_length	0.006214	-0.006757
area_code	-0.008264	0.003580
international_plan	0.049398	0.019100
voice_mail_plan	-0.001686	0.021545
number_vmail_messages	0.000776	0.017562
total_day_minutes	1.000000	0.007043
total_day_calls	0.006753	-0.021451
total_day_charge	1.000000	0.007050
total_eve_minutes	0.007050	1.000000
total_eve_calls	0.015769	-0.011430
total_eve_charge	0.007036	1.000000
total_night_minutes	0.004324	-0.012584
total_night_calls	0.022972	0.007586
total_night_charge	0.004301	-0.012593
total_intl_minutes	-0.010157	-0.011035
total_intl_calls	0.008032	0.002541
total_intl_charge	-0.010094	-0.011067
number_customer_service_calls	-0.013427	-0.012985
churn	0.205151	0.092796

	total_eve_calls	total_eve_charge \
account_length	0.019260	-0.006745
area_code	-0.011886	0.003607
international_plan	0.006114	0.019106
voice_mail_plan	-0.006444	0.021559
number_vmail_messages	-0.005864	0.017578

total_day_minutes	0.015769	0.007029
total_day_calls	0.006462	-0.021449
total_day_charge	0.015769	0.007036
total_eve_minutes	-0.011430	1.000000
total_eve_calls	1.000000	-0.011423
total_eve_charge	-0.011423	1.000000
total_night_minutes	-0.002093	-0.012592
total_night_calls	0.007710	0.007596
total_night_charge	-0.002056	-0.012601
total_intl_minutes	0.008703	-0.011043
total_intl_calls	0.017434	0.002541
total_intl_charge	0.008674	-0.011074
number_customer_service_calls	0.002423	-0.012987
churn	0.009233	0.092786

	total_night_minutes	total_night_calls \
account_length	-0.008955	-0.013176
area_code	-0.005825	0.016522
international_plan	-0.028905	0.012451
voice_mail_plan	0.006079	0.015553
number_vmail_messages	0.007681	0.007123
total_day_minutes	0.004323	0.022972
total_day_calls	0.022938	-0.019557
total_day_charge	0.004324	0.022972
total_eve_minutes	-0.012584	0.007586
total_eve_calls	-0.002093	0.007710
total_eve_charge	-0.012592	0.007596
total_night_minutes	1.000000	0.011204
total_night_calls	0.011204	1.000000
total_night_charge	0.999999	0.011188
total_intl_minutes	-0.015207	-0.013605
total_intl_calls	-0.012353	0.000305
total_intl_charge	-0.015180	-0.013630
number_customer_service_calls	-0.009288	-0.012802
churn	0.035493	0.006141

	total_night_charge	total_intl_minutes \
account_length	-0.008960	0.009514
area_code	-0.005845	-0.018288
international_plan	-0.028913	0.045871
voice_mail_plan	0.006064	-0.001318
number_vmail_messages	0.007663	0.002856
total_day_minutes	0.004300	-0.010155
total_day_calls	0.022927	0.021565
total_day_charge	0.004301	-0.010157
total_eve_minutes	-0.012593	-0.011035
total_eve_calls	-0.002056	0.008703
total_eve_charge	-0.012601	-0.011043

total_night_minutes	0.999999	-0.015207
total_night_calls	0.011188	-0.013605
total_night_charge	1.000000	-0.015214
total_intl_minutes	-0.015214	1.000000
total_intl_calls	-0.012329	0.032304
total_intl_charge	-0.015186	0.999993
number_customer_service_calls	-0.009277	-0.009640
churn	0.035496	0.068239

	total_intl_calls	total_intl_charge \
account_length	0.020661	0.009546
area_code	-0.024179	-0.018395
international_plan	0.017366	0.045780
voice_mail_plan	0.007618	-0.001276
number_vmail_messages	0.013957	0.002884
total_day_minutes	0.008033	-0.010092
total_day_calls	0.004574	0.021666
total_day_charge	0.008032	-0.010094
total_eve_minutes	0.002541	-0.011067
total_eve_calls	0.017434	0.008674
total_eve_charge	0.002541	-0.011074
total_night_minutes	-0.012353	-0.015180
total_night_calls	0.000305	-0.013630
total_night_charge	-0.012329	-0.015186
total_intl_minutes	0.032304	0.999993
total_intl_calls	1.000000	0.032372
total_intl_charge	0.032372	1.000000
number_customer_service_calls	-0.017561	-0.009675
churn	-0.052844	0.068259

	number_customer_service_calls	churn
account_length	-0.003796	0.016541
area_code	0.027572	0.006174
international_plan	-0.024522	0.259852
voice_mail_plan	-0.017824	-0.102148
number_vmail_messages	-0.013263	-0.089728
total_day_minutes	-0.013423	0.205151
total_day_calls	-0.018942	0.018459
total_day_charge	-0.013427	0.205151
total_eve_minutes	-0.012985	0.092796
total_eve_calls	0.002423	0.009233
total_eve_charge	-0.012987	0.092786
total_night_minutes	-0.009288	0.035493
total_night_calls	-0.012802	0.006141
total_night_charge	-0.009277	0.035496
total_intl_minutes	-0.009640	0.068239
total_intl_calls	-0.017561	-0.052844
total_intl_charge	-0.009675	0.068259

number_customer_service_calls	1.000000	0.208750
churn	0.208750	1.000000

```
In [23]: # Distribuição da variável TARGET
df.groupby('churn').size()
```

```
Out[23]: churn
0      2850
1       483
dtype: int64
```

```
In [91]: # Separando os dados
array = df.values
colunas = df.columns[1:19]

# Separando o array em componentes de input (X) e output (Y)
X = array[:,1:19]
Y = array[:,19]
```

```
In [75]: # Normalizando e padronizando os dados do dataframe
```

```
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import StandardScaler
```

```
# Gerando a nova escala (normalizando os dados)
scaler = MinMaxScaler(feature_range = (0, 1))
rescaledX = scaler.fit_transform(X)
```

```
#Gerando padronização dos valores
scalerP = StandardScaler().fit(rescaledX)
standardX = scalerP.transform(rescaledX)
```

```
In [107]: #Montando novo DataFrame com State, Churn e as colunas normalizadas na escala e Padr
df_new = pd.DataFrame (data=standardX, columns=colunas)
df_new['state'] = df['state']
df_new['churn'] = df['churn']

df_new.head(10)

array1 = df_new.values
X1 = array1[:,0:18]
Y1 = array1[:,19]
Y1 = Y1.astype(int)
```

```
In [108]: #Gerando dados de Treino e de Teste para os modelos
from sklearn.model_selection import train_test_split

seed = 1313
X_treino, X_teste, y_treino, y_teste = train_test_split(X1, Y1, test_size = 0.30, ran
```

```

In [109]: y_treino

Out[109]: array([0, 0, 0, ..., 0, 0, 0])

In [110]: # Feature selection com Random Forest
from sklearn.ensemble import RandomForestClassifier
from sklearn.feature_selection import SelectFromModel

clf = RandomForestClassifier(random_state=seed)
selector = clf.fit(X_treino, y_treino)
fs = SelectFromModel(selector, prefit=True)

X_treino_new = fs.transform(X_treino)
X_teste_new = fs.transform(X_teste)

print(X_treino_new.shape, X_teste_new.shape)

(2333, 6) (1000, 6)

In [111]: #Criando o Modelo - Versao 1
from sklearn.linear_model import LogisticRegression

modelo = LogisticRegression()

In [112]: # Treinamento do modelo
modelo.fit(X_treino_new, y_treino)

Out[112]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
    intercept_scaling=1, max_iter=100, multi_class='warn',
    n_jobs=None, penalty='l2', random_state=None, solver='warn',
    tol=0.0001, verbose=0, warm_start=False)

In [113]: #Resultado do Modelo - Versao 1
resultado = modelo.score(X_teste_new, y_teste)
print("Acurácia nos Dados de Teste: %.3f%%" % (resultado * 100.0))

Acurácia nos Dados de Teste: 84.200%

In [119]: #Criando o modelo 2 - XGBClassifier
#!pip install xgboost
from sklearn.metrics import accuracy_score
from xgboost import XGBClassifier

modelo2 = XGBClassifier(n_estimators=110, nthread=-1, seed=seed)

# Treinando o modelo
modelo2.fit(X_treino_new, y_treino, eval_metric="auc", verbose = False)

```

```
Out[119]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
                        colsample_bynode=1, colsample_bytree=1, gamma=0, learning_rate=0.1,
                        max_delta_step=0, max_depth=3, min_child_weight=1, missing=None,
                        n_estimators=110, n_jobs=1, nthread=-1, objective='binary:logistic',
                        random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
                        seed=1313, silent=None, subsample=1, verbosity=1)
```

```
In [120]: # Fazendo previsões
```

```
        y_pred = modelo2.predict(X_teste_new)
        previsoes = [round(value) for value in y_pred]
```

```
In [121]: #Resultado do Modelo - Versao 2
```

```
        resultado2 = accuracy_score(y_teste, previsoes)
        print("Acuracia do Modelo 2: %.3f" % (resultado2 * 100.0))
```

Acuracia do Modelo 2: 90.000

0.7 FIM

0.8 OBRIGADO