	1- Análise exploratória dos dados utilizando estatística descritiva e inferencial, considerando uma, duas e/ou mais variáveis; Importação das libs
In [35]:	<pre>import pandas as pd import sweetviz as sv</pre>
	<pre>import numpy as np import sklearn import matplotlib.pyplot as plt</pre>
	Chamando o arquivo do excel
In [3]:	<pre>df = pd.read_excel('teste_smarkio_Lbs.xls','Análise_ML')</pre>
In [5]:	Verificando quantas linhas tem dado nulo df.isnull().sum()
Out[5]:	Pred_class 0 probabilidade 0
	status 0 True_class 462 dtype: int64
	Tratando os dados da coluna True_class(Todos que o true_class for NaN, considerar o valor da coluna Pred_class)
In [16]:	<pre>values = {'True_class': df['Pred_class']}</pre>
In [6]:	df.fillna(value=values,inplace = True)
In [83]:	Verificando quantas linhas tem dado nulo(depois de rodar o comando para tratar os dados) df.isnull().sum()
Out[83]:	Pred_class 0 probabilidade 0
	status 0 True_class 0 dtype: int64
In [8]:	Trazendo todas as informações do DataFrame
Out[8]:	Pred_class probabilidade status True_class
	0 2 0.079892 approved 0.0 1 2 0.379377 approved 74.0
	2 0.379377 approved 74.0 3 2 0.420930 approved 74.0
	4 2 0.607437 approved 2.0
	638 60 0.543772 revision 60.0 639 60 0.553846 revision 60.0
	640 77 0.606065 revision 77.0 641 84 0.561842 revision 84.0
	642 96 0.340740 revision 96.0 643 rows × 4 columns
	Agrupando a coluna status pela contagem dos valores da coluna probabilidade
In [119	<pre>status_group_probabi = df.groupby('status')['probabilidade'].value_counts()</pre>
In [120	status_group_probabi
Out[120	status probabilidade approved 1.000000 56 0.432421 14 0.426136 11
	0.057740 8 0.288653 5 revision 0.740292 1
	0.752448 1 0.784920 1 0.817525 1
In [96]:	0.909148 1 Name: probabilidade, Length: 492, dtype: int64
	status = df['status'] Agrupando a coluna status com as informações da coluna True_class
In [11]:	<pre>status_group_true =df.groupby('status')['True_class'].value_counts()</pre>
In [98]:	status_group_true
Out[98]:	approved 74.0 78 2.0 58
	3.0 57 0.0 54 77.0 26
	revision 81.0 1 84.0 1 86.0 1
	113.0 1
In [13]:	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64
In [13]: In [14]:	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status.
	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43
In [14]:	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600
<pre>In [14]: Out[14]: In [15]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() df.describe()
In [14]: Out[14]:	113.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() Pred_class probabilidade True_class count 643.000000 643.000000 643.000000
<pre>In [14]: Out[14]: In [15]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() df.describe() Pred_class probabilidade True_class count 643.000000 643.000000 643.000000 mean 52.712286 0.622436 48.251944 std 37.602068 0.266811 38.542269
<pre>In [14]: Out[14]: In [15]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() Pred_class probabilidade True_class count 643.000000 643.000000 mean 52.712286 0.622436 48.251944 std 37.602068 0.266811 38.542269 min 2.000000 0.043858 0.000000 25% 12.000000 0.408017 3.000000
<pre>In [14]: Out[14]: In [15]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() Pred_class probabilidade True_class count 643.000000 643.000000 643.000000 mean 52.712286 0.622436 48.251944 std 37.602068 0.266811 38.542269 min 2.000000 0.043858 0.000000 25% 12.000000 0.408017 3.000000 50% 59.000000 0.616809 55.000000 75% 81.000000 0.870083 77.000000
<pre>In [14]: Out[14]: In [15]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() df.describe() Pred_class probabilidade True_class count 643.000000 643.000000 mean 52.712286 0.622436 48.251944 std 37.602068 0.266811 38.542269 min 2.000000 0.043858 0.000000 25% 12.000000 0.408017 3.000000 50% 59.000000 0.616809 55.000000
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<pre>In [14]: Out[14]: In [15]: Out[15]: In [40]:</pre>	113.0 1 114.0 1 Name: True_class, Length: 91, dtype: int64 Fazendo a contagem do total da coluna status, agrupado por cada status. status_group_status = df.groupby('status')['status'].count() status_group_status status approved 600 revision 43 Name: status, dtype: int64 Utilizando o comando describe() Pred_class probabilidade True_class
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