EDA

August 17, 2022

1 Case

Imagine you've just been hired as a Data Analyst to join one of our Data Squads, responsible for leveraging the products team and operation team to make better decisions based on data and provide insights that improve the our stakeholders decisions.

This quarter, your main goal is to help the product team to increase the number of contracts signed, which has been in steady decline since early this year.

- 1. What seems to be happening? Try to understand main bottlenecks along the funnel and raise hypotheses that may have caused each behavior.
 - Os dados do de contratos assinados por visita sugerem que na maioria das vezes, as primeiras visitas não são convertidas em contratos assinados. Isto também é visto no funil de contrato de ofertas de visitas com o número de inquilinos que fizeram ofertas. Talvez se os inquilinos fizessem mais visitas, aumentasse a probabilidade de fazer ofertas e, por consequencia, assinar contrato. Uma outra possibilidade, é causar uma melhor impressão ou um melhor match nas primeiras visitas. Para isso, aumentar a verossimilhança dos anúncios pode ajudar: muitas reviews tem marcações da casa ser diferente do anúncio, não estar tão conservada e/ou ser maior nas fotos. As pessoas estão indo para a visita com expectavas diferentes do que vão encontrar. Ao mostrar anúncios que retraram melhor a condição da casa, pode resultar nas pessoas com as expextativas alinhadas na primeira visita, aumente a probabilidade de fazer uma oferta e assinar contrato.
 - Além disso, o **número de casas visitadas** parece ser um outro bottleneck, que limita todo o funil abaixo**: número de ofertas, de inquilinos que fizeram ofertas e, consequentemente, o número de contratos. Essa é a maior separação no gráfico de funil.
- 2. Is there any other information that would be relevant to your analysis? Why would it be relevant? What would be the outputs with the new data available?
 - In the reviews, would be helpful to know which visits resulted in contracts.
 - How many visits each tenant did up to that point. That is, would be helfull to have the
 other two tables at visit granularity. Com esses dados adicionais poderíamos conectar
 as informações das três tabelas fornecidas e fazer obter insights mais precisos e aprofundados.
- 3. Can you think of 3 other analysis that could help our stakeholder with our quarter goal? Please describe the analysis as well as the benefits of each one for our stakeholders. Try to explain how the idea could be implemented, connecting strategy with product team and how these analysis can make impact in our stakeholders decisions.

- Analisar a qualidade dos anúncios e o que fazer para melhor alinhar as expectativas dos clientes na primeira visita.
- Analisar como melhorar os filtros de anúncios/recomendação de anúncios, para que o cliente consiga um 'match' mais adequado e, assim, aumenta a probabilidade de assinar contrato nas primeiras visitas.
- Analisar/Entender o que faz o número de casas visitas ser menor. Talvez, novamente esteja relacionada com a qualidade dos anúncios, localidade, etc.

2 Loading data

```
[1]: %reload_ext autoreload
%autoreload 2
import pandas as pd
import psycopg2
from tqdm.cli import tqdm
import janitor

# tenant_reviews
TABLES_NAMES =□

□ □ ['tenant_reviews_clean','visits_contracts','visits_offers_contracts']
```

```
[2]: def load_tables_sql(tables_names = TABLES_NAMES):
         print("loading tables from sql")
         conn = psycopg2.connect(
             dbname="quintoandar",
             host="db.candidates.sandbox.data.quintoandar.com.br",
             port="5432",
             user="data-analytics",
             password="97PGf56yWCWa",
             client_encoding='UTF8'
         )
         query = "select * from {}"
         tables = {}
         try:
             for table in tqdm(tables_names):
                 df = pd.read_sql(query.format(table), con=conn)
                 tables[table] = df
                 print(table, df.shape)
         finally:
             conn.close()
         return tables
     def load_tables_parquet(tables = TABLES_NAMES):
         print("loading tables from parquet")
```

```
parquet_tables = {}
         for table in tables:
             df = pd.read_parquet(f'data/{table}.parquet')
             parquet_tables[table] = df
             print(table, df.shape)
             #print(f"{table}:shape of differences: {df.compare(df2).shape}")
         return parquet_tables
     load_sql = False
     if load sql:
         tables = load_tables_sql()
         # for table, df in tables.items():
              df.to_parquet(f'data/{table}.parquet')
     else:
         tables = load_tables_parquet()
    loading tables from parquet
    tenant_reviews_clean (13470, 18)
    visits_contracts (15, 3)
    visits_offers_contracts (8, 7)
[3]: #dftenant reviews = tables['tenant reviews']
     dftenant_reviews_raw = tables['tenant_reviews_clean']
     dfvisits_contracts = tables['visits_contracts']
     dfvisits_offers_contracts_raw = tables['visits_offers_contracts']
```

3 Tratamentos

```
[4]: # data treatments
dftenant_reviews = dftenant_reviews_raw.copy()

dftenant_reviews["visit_date"] = pd.to_datetime(dftenant_reviews["visit_date"])

dftenant_reviews['total'] = True

# order review categories
review_dtype = pd.CategoricalDtype(
        ["Visita não aconteceu.", "1", "2", "3", "4", "5", "VaiNegociar"],
        ordered=True
)

dftenant_reviews["review"] = dftenant_reviews["review"].replace(
        "EntradaNaoAutorizada", "Visita não aconteceu."
).astype(review_dtype)
```

```
# review dummies
dftenant_reviews = pd.concat(
    [dftenant_reviews, pd.get_dummies(dftenant_reviews["review"],
    prefix="review")],
    axis=1,
)
dftenant_reviews['visit_date_ym'] = dftenant_reviews['visit_date'].dt.
    strftime('%b')
#dftenant_reviews.columns
```

```
[5]: # grouping columns
     flags = [
         "building",
         "conservation",
         "size",
         "price",
         "location",
         "total",
     ]
     neg_flags = [
         "did_not_like_the_location",
         "expected_a_better_preserved_property",
         "bigger_in_photos",
         "different_from_the_listing",
         "total",
     ]
     others = [
        # "visit_date",
        # "day",
        # "time",
        # "house_id",
         # "neighbourhood",
        # "visitor",
         # "agent",
         # "review",
         # "others",
     dummies = list(dftenant_reviews.columns[dftenant_reviews.columns.str.
      ⇔startswith("review_")])
```

```
[6]: # dfvisits_offers_contracts
     months_cat = ['jan', 'fev', 'mar', 'abr', 'mai', 'jun', 'jul', 'ago']
     months_cat = pd.CategoricalDtype(months_cat, ordered=True)
     funnel_cat = [
         "visits",
         "tenant_that_visited",
         "visited_houses",
         "offers",
         "tenants_that_made_offers",
         "contracts",
     funnel_cat = pd.CategoricalDtype(reversed(funnel_cat), ordered=True)
     dfvisits_offers_contracts = pd.

-melt(dfvisits_offers_contracts_raw,id_vars=["month"])
     dfvisits_offers_contracts['month'] = dfvisits_offers_contracts['month'].
      ⇒astype(months_cat)
     dfvisits_offers_contracts['variable'] = dfvisits_offers_contracts['variable'].
      →astype(funnel_cat)
```

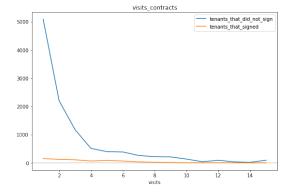
4 EDA

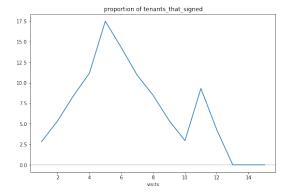
4.1 Contratos de visitas

Para começar, vamos explorar a tabela de contratos de visitas.

O gráfico abaixo mostra que o número inquilinos que não assinaram cai (quase que) exponencialmente com o número de visitas. Enquanto que o número de inquilinos que assinaram também decresce, mas num .

Se considerarmos a proporção de inquilinos que assinaram como uma estimativa de probabilidade, então a probabilidade do inquilino assinar parece crescer com o número de visitas até a quinta vista e depois decrescer.





4.2 Funil de contratos de ofertas de visitas

A figura mostra dois gráficos: no lado esquerdo, a tendências das variáveis ao longo do tempo e, no lado direito, as correlações (lineares) dessas variáveis entre si.

Como mencionado no case, o número de contratos apresenta uma tendência de queda.

O número de inquilinos que visitaram e o número de visitas mostram-se negativamente correlacionadas com o número de contratos. Enquanto que o número de inquilinos que fizeram uma oferta é positivamente correlacionado com o número de contratos.

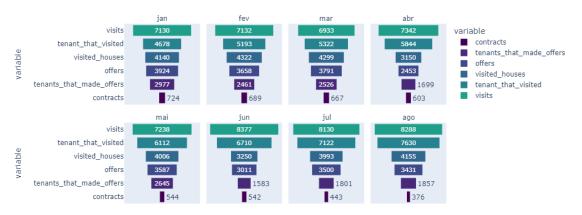
O bottleneck que parece ser o mais grave acontece no número de casas visitadas. O **número de** casas visitadas parece estar limitando o funil abaixo: número de ofertas, de inquilinos que fizeram ofertas e, consequentemente, o número de contratos.

Um outro bottleneck parece ser o número de inquilinos que fizeram ofertas.

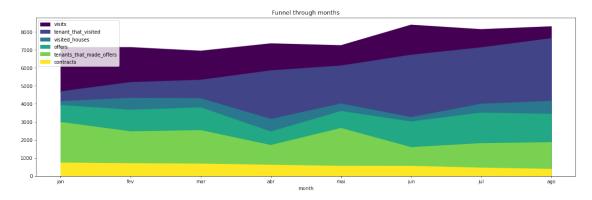
```
[38]: import plotly.express as px
      fig = px.funnel(
          dfvisits_offers_contracts,
          x="value",
          y="variable",
          color="variable",
          color_discrete_sequence=px.colors.sequential.Viridis,
          category_orders={"variable": funnel_cat.categories},
          facet_col="month",
          facet_col_wrap=4,
          facet_row_spacing=0.1,
          title='Funnel through months',
          width=1000, height=500
      fig.for_each_annotation(lambda a: a.update(text=a.text.split("=")[1]))
      \#fig.for\_each\_trace(lambda\ t:\ t.update(name=t.name.split("=")[1]))
      #fig.show()
      from IPython.display import Image
      Image(fig.to_image())
```

[38]:

Funnel through months



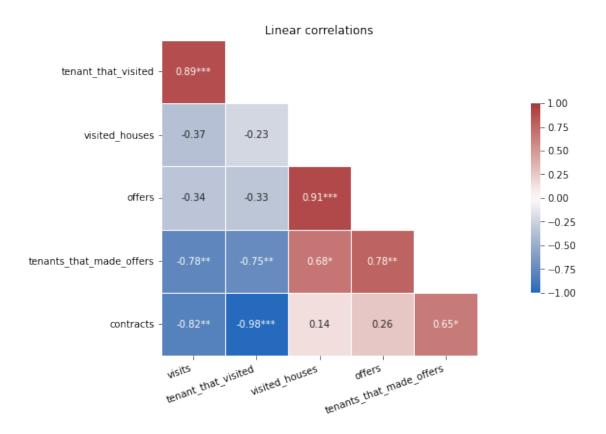
```
[9]: # dfvisits_offers_contracts
from plot_corr import plot_corr
fig, ax = plt.subplots(figsize=(20, 6))
# plot1
```



```
[10]: # dfvisits_offers_contracts
from plot_corr import plot_corr

fig, ax = plt.subplots(figsize=(20, 6))

plot_corr(dfvisits_offers_contracts_raw, ax=ax)
ax.set_title('Linear correlations')
plt.show()
```



4.3 Review dos inquilinos

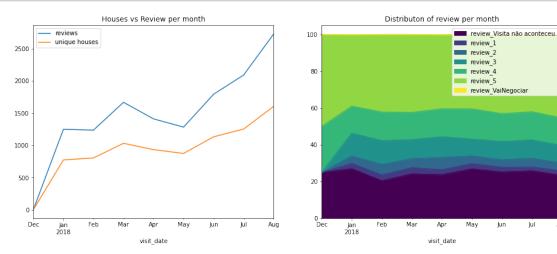
```
month = pd.Grouper(key='visit_date', freq='1M')

# dftenant_reviews
fig,axs=plt.subplots(ncols=2,figsize=(16,6))

#plot1
ax=axs[0]
dftenant_reviews.groupby(month)['total'].sum().plot(title="Houses vs Review perusimenth", label='reviews', ax=ax)
dftenant_reviews.groupby(month).house_id.nunique().plot(label='unique houses',usax=ax)
ax.legend()

#plot2
reviews_per_month = dftenant_reviews.groupby(month)[dummies+['total']]
(reviews_per_month.mean()* 100).iloc[:, :-1].plot.area(cmap='viridis',usatitle='Distributon of review per month', ax=axs[1])
```

```
\#display(reviews\_per\_month.agg(lambda\ x:\ (x.sum(),x.mean())).pipe(lambda\ x,\ fmt:
 \rightarrow x.set index(x.index.strftime(fmt)), "%Y-%m").T.style.format(fmtr))
# tabela numero de reviews por mes por tipo de review
#
#
      dftenant reviews.groupby(
           [pd.Grouper(key="visit_date", freq="1M"), "review"], dropna=True
#
#
      )["total"]
#
      .count()
#
      .unstack(1)
#
      .rename_axis(columns="reviews por mes")
      .pipe(lambda x: x.set_axis(x.index.strftime("%b"), axis=0))
# )
plt.show()
```



Jul

4.4 Tenant's opinions

```
[12]: # positive opinions
      flags_per_review = dftenant_reviews.groupby('review')[flags].agg(lambda x: (x.
       \rightarrowsum(),x.mean())).T
      display(flags_per_review.rename_axis(index='positive opinions (liked the ...)').
       ⇔style.format(fmtr))
      flags_per_review = dftenant_reviews.groupby('visit_date_ym', sort=False)[flags].
       \rightarrowagg(lambda x: (x.sum(),x.mean())).T
      display(flags_per_review.rename_axis(index='positive_opinions_per_month').style.
       →format(fmtr))
```

<pandas.io.formats.style.Styler at 0x27bbf82ce50>

<pandas.io.formats.style.Styler at 0x27bbf82ce50>

```
[13]: # negative opinions

#dftenant_reviews.groupby('review')[flags].apply(lambda x: x.sum().astype(str)_

+ " (" + (x.mean()*100).round(0).astype(str)+"%)").T

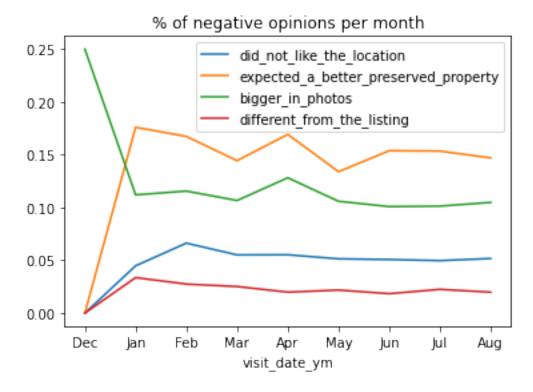
neg_flags_per_review = dftenant_reviews.groupby('review')[neg_flags].agg(lambda_

x: (x.sum(),x.mean())).T

display(neg_flags_per_review.rename_axis(index='Negative opinions').style.

format(fmtr))
```

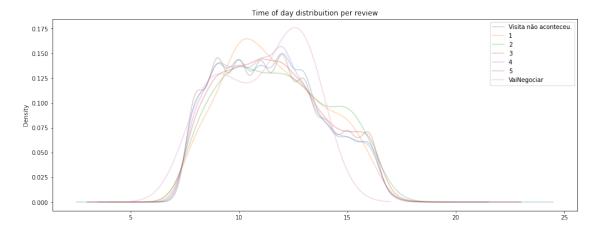
<pandas.io.formats.style.Styler at 0x27bc1c52f10>



```
[15]: # alguma review negativa?
dftenant_reviews['sum_neg_opn'] = dftenant_reviews[neg_flags[:-1]].sum(axis=1)
dftenant_reviews.groupby('review')[['sum_neg_opn']].agg(['mean','sum', 'count'])
```

```
[15]:
                              sum_neg_opn
                                     mean
                                             sum count
      review
      Visita não aconteceu.
                                 0.000303
                                                 3302
                                               1
                                 1.340659
                                                    364
                                             488
      2
                                 1.233071
                                             783
                                                    635
      3
                                 1.064266
                                            1507
                                                  1416
      4
                                 0.857073
                                            1757
                                                  2050
      5
                                 0.000000
                                                  5673
                                               0
      VaiNegociar
                                 0.566667
                                              17
                                                     30
```

4.5 Review changes with visit time?



```
[17]: def percentile(n):
          def percentile_(x):
              return np.percentile(x, n)
          percentile_.__name__ = "percentile_%s" % n
          return percentile_
      # numero médio de visitas por casa
      # em média, cada casa foi visitada n vezes por mes
      df visits per house = (
          dftenant_reviews.groupby([month, "house_id"])[["total"]]
          .sum()
          .reset_index()
          .groupby("visit_date")
          .total.agg(["min", "mean", "median", percentile(75), percentile(95),
       →percentile(99), "max",])# "count", "var", "std"])
          #.set_index('count', append=True)
      df_visits_per_house.plot(title='estatística do número de vistas por_
       ⇔casa',legend='reverse')
      plt.show()
      display(df_visits_per_house)
      # **3. ideias: analisar/entender o que faz as pessoas voltarem para outrasu
      # como aumentar o numero de casas e a qualidade das casas
```

```
Traceback (most recent call last)
NameError
c:\Users\rubens\PROJETOS\5oandar\EDA.ipynb Célula: 27 in <cell line: 12>()
      <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
 →ipynb#X36sZmlsZQ%3D%3D?line=5'>6</a>
                                            return percentile_
      <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
 →ipynb#X36sZmlsZQ%3D%3D?line=8'>9</a> # numero médio de visitas por casa
     <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</p>
 →ipynb#X36sZmlsZQ%3D%3D?line=9'>10</a> # em média, cada casa foi visitada n⊔
 ⇔vezes por mes
     <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
 ipynb#X36sZmlsZQ%3D%3D?line=10'>11</a> df_visits_per_house = (
---> <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.
 sipynb#X36sZmlsZQ%3D%3D?line=11'>12</a> dftenant_reviews.groupby([month,_
 →"house_id"])[["total"]]
```

```
<a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</p>
  \hookrightarrowipynb#X36sZmlsZQ%3D%3D?line=12'>13</a>
                                                                                    .sum()
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
  →ipynb#X36sZmlsZQ%3D%3D?line=13'>14</a>
                                                                                    .reset index()
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</p>
  →ipynb#X36sZmlsZQ%3D%3D?line=14'>15</a>
                                                                                    .groupby("visit date")
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.
  oipynb#X36sZmlsZQ%3D%3D?line=15'>16</a> .total.agg(["min", "mean", outpynb#X36sZmlsZQ%3D%3D?line=15'>16</a> .total.agg(["min", outpynb#X36sZmlsZQ%3D%3D?line=15</a> .total.agg(["min", outpynb#X36sZmlsZQ%3D%3D?line=15</a> .
  →ipynb#X36sZmlsZQ%3D%3D?line=15'>16</a>
  →"var", "std"])
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.
                                                                                   #.set index('count', append=True)
  →ipvnb#X36sZmlsZ0%3D%3D?line=16'>17</a>
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</p>
  →ipynb#X36sZmlsZQ%3D%3D?line=17'>18</a> )
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
  ⇒ipynb#X36sZmlsZQ%3D%3D?line=21'>22</a> df_visits_per_house.
  ⊶plot(title='estatística do número de vistas por casa',legend='reverse')
         <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
  →ipynb#X36sZmlsZQ%3D%3D?line=22'>23</a> plt.show()
File c:
  →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\generic.
  py:271, in SeriesGroupBy.aggregate(self, func, engine, engine_kwargs, *args,
  →**kwargs)
       267 elif isinstance(func, abc.Iterable):
       268
                      # Catch instances of lists / tuples
       269
                      # but not the class list / tuple itself.
                      func = maybe mangle lambdas(func)
       270
--> 271
                      ret = self. aggregate multiple funcs(func)
       272
                      if relabeling:
       273
                             # error: Incompatible types in assignment (expression has type
                             # "Optional[List[str]]", variable has type "Index")
       274
       275
                             ret.columns = columns # type: ignore[assignment]
File c:
  →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\gen_ric.
  →py:326, in SeriesGroupBy. aggregate multiple funcs(self, arg)
       323 for idx, (name, func) in enumerate(arg):
       325
                      key = base.OutputKey(label=name, position=idx)
--> 326
                      results[key] = self.aggregate(func)
       328 if any(isinstance(x, DataFrame) for x in results.values()):
       329
                      from pandas import concat
File c:
  →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\generic.
  py:287, in SeriesGroupBy.aggregate(self, func, engine, engine_kwargs, *args,
  →**kwargs)
       284
                      return self._python_agg_general(func, *args, **kwargs)
       286 try:
                      return self._python_agg_general(func, *args, **kwargs)
```

```
288 except KeyError:
            # TODO: KeyError is raised in _python_agg_general,
    290
            # see test_groupby.test_basic
    291
            result = self._aggregate_named(func, *args, **kwargs)
File c:
 →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\grc pby.
 opy:1490, in GroupBy. python agg general(self, func, *args, **kwargs)
   1486 name = obj.name
   1488 try:
   1489
            # if this function is invalid for this dtype, we will ignore it.
-> 1490
            result = self.grouper.agg_series(obj, f)
   1491 except TypeError:
   1492
            warn_dropping_nuisance_columns_deprecated(type(self), "agg")
File c:
 →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\ops
 apy:981, in BaseGrouper.agg_series(self, obj, func, preserve_dtype)
            preserve_dtype = True
    978
    980 else:
--> 981
            result = self._aggregate_series_pure_python(obj, func)
    983 npvalues = lib.maybe_convert_objects(result, try_float=False)
    984 if preserve_dtype:
File c:
 →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\ops
 apy:1005, in BaseGrouper._aggregate_series_pure_python(self, obj, func)
   1003 for i, group in enumerate(splitter):
            group = group.__finalize__(obj, method="groupby")
   1004
-> 1005
            res = func(group)
           res = libreduction.extract_result(res)
   1006
            if not initialized:
   1008
   1009
                # We only do this validation on the first iteration
File c:
 →\Users\rubens\PROJETOS\5oandar\venv\lib\site-packages\pandas\core\groupby\grc pby.
 →py:1476, in GroupBy._python_agg_general.<locals>.<lambda>(x)
   1473 @final
   1474 def _python_agg_general(self, func, *args, **kwargs):
            func = com.is_builtin_func(func)
   1475
-> 1476
            f = lambda x: func(x, *args, **kwargs)
            # iterate through "columns" ex exclusions to populate output dict
   1478
            output: dict[base.OutputKey, ArrayLike] = {}
   1479
c:\Users\rubens\PROJETOS\5oandar\EDA.ipynb Célula: 27 in percentile.<locals>.
 →percentile_(x)
      <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.</pre>
 →ipynb#X36sZmlsZQ%3D%3D?line=1'>2</a> def percentile_(x):
```

```
----> <a href='vscode-notebook-cell:/c%3A/Users/rubens/PROJETOS/5oandar/EDA.

sipynb#X36sZmlsZQ%3D%3D?line=2'>3</a> return np.percentile(x, n)

NameError: name 'np' is not defined
```

[]: