

# RA27246\_aula6\_ex6-5\_cap7

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## 0.1 IA376I – Tópicos em Engenharia de Computação VII

### 0.1.1 Tópico: Análise de Dados Visual (Visual Analytics)

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Aula 06 - 19/04/2024

### 0.1.2 Exercícios 6.5

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**Exercício 2:** Reproduza os exemplos fornecidos nos Capítulos 6, 7 e 8 em [89] (Python) ou nos Capítulos 9 a 16 em [90] (R). Em ambas as referências, são abordadas diversas funções adicionais de manipulação dos dados, além das apresentadas neste capítulo, proporcionando uma visão mais abrangente das capacidades das respectivas linguagens de programação.

Reprodução dos passos:

#### Capítulo 7: limpeza e preparação de dados

##### Tratamento de dados ausentes

```
[ ]: import pandas as pd
import numpy as np

#Base da URL do repositório com os arquivos de exemplo utilizados no livro
base_url = "https://raw.githubusercontent.com/wesm/pydata-book/3rd-edition/"

float_data = pd.Series([1.2, -3.5, np.nan, 0])
float_data
```

```
[ ]: 0    1.2
     1   -3.5
     2    NaN
     3    0.0
     dtype: float64
```

```
[ ]: float_data.isna()
```

```
[ ]: 0    False
      1    False
      2     True
      3    False
      dtype: bool
```

```
[ ]: string_data = pd.Series(["aardvark", np.nan, None, "avocado"])
      string_data
```

```
[ ]: 0    aardvark
      1         NaN
      2        None
      3    avocado
      dtype: object
```

```
[ ]: string_data.isna()
```

```
[ ]: 0    False
      1     True
      2     True
      3    False
      dtype: bool
```

```
[ ]: float_data = pd.Series([1, 2, None], dtype='float64')
      float_data
```

```
[ ]: 0    1.0
      1    2.0
      2    NaN
      dtype: float64
```

```
[ ]: float_data.isna()
```

```
[ ]: 0    False
      1    False
      2     True
      dtype: bool
```

```
[ ]: data = pd.Series([1, np.nan, 3.5, np.nan, 7])
      data.dropna()
```

```
[ ]: 0    1.0
      2    3.5
      4    7.0
      dtype: float64
```

```
[ ]: data[data.notna()]
```

```
[ ]: 0    1.0
      2    3.5
      4    7.0
      dtype: float64
```

```
[ ]: data = pd.DataFrame([
      [1., 6.5, 3.],
      [1., np.nan, np.nan],
      [np.nan, np.nan, np.nan],
      [np.nan, 6.5, 3.]
    ])
      data
```

```
[ ]:      0    1    2
0  1.0  6.5  3.0
1  1.0  NaN  NaN
2  NaN  NaN  NaN
3  NaN  6.5  3.0
```

```
[ ]: data.dropna()
```

```
[ ]:      0    1    2
0  1.0  6.5  3.0
```

```
[ ]: data.dropna(how="all")
```

```
[ ]:      0    1    2
0  1.0  6.5  3.0
1  1.0  NaN  NaN
3  NaN  6.5  3.0
```

```
[ ]: data[4] = np.nan
      data
```

```
[ ]:      0    1    2    4
0  1.0  6.5  3.0  NaN
1  1.0  NaN  NaN  NaN
2  NaN  NaN  NaN  NaN
3  NaN  6.5  3.0  NaN
```

```
[ ]: data.dropna(axis="columns", how="all")
```

```
[ ]:      0    1    2
0  1.0  6.5  3.0
1  1.0  NaN  NaN
2  NaN  NaN  NaN
3  NaN  6.5  3.0
```

```
[ ]: df = pd.DataFrame(np.random.standard_normal((7, 3)))
df.iloc[:4, 1] = np.nan
df.iloc[:2, 2] = np.nan
df
```

```
[ ]:
      0      1      2
0 -1.000587    NaN    NaN
1  0.068673    NaN    NaN
2 -1.312555    NaN  1.303557
3 -0.243463    NaN -0.780911
4 -0.125387  0.708737 -0.966796
5  1.796565  2.950215 -0.947793
6 -0.360379 -1.212389  0.269852
```

```
[ ]: df.dropna()
```

```
[ ]:
      0      1      2
4 -0.125387  0.708737 -0.966796
5  1.796565  2.950215 -0.947793
6 -0.360379 -1.212389  0.269852
```

```
[ ]: df.dropna(thresh=2)
```

```
[ ]:
      0      1      2
2 -1.312555    NaN  1.303557
3 -0.243463    NaN -0.780911
4 -0.125387  0.708737 -0.966796
5  1.796565  2.950215 -0.947793
6 -0.360379 -1.212389  0.269852
```

```
[ ]: df.fillna(0)
```

```
[ ]:
      0      1      2
0 -1.000587  0.000000  0.000000
1  0.068673  0.000000  0.000000
2 -1.312555  0.000000  1.303557
3 -0.243463  0.000000 -0.780911
4 -0.125387  0.708737 -0.966796
5  1.796565  2.950215 -0.947793
6 -0.360379 -1.212389  0.269852
```

```
[ ]: df.fillna({1: 0.5, 2: 0})
```

```
[ ]:
      0      1      2
0 -1.000587  0.500000  0.000000
1  0.068673  0.500000  0.000000
2 -1.312555  0.500000  1.303557
```

```

3 -0.243463  0.500000 -0.780911
4 -0.125387  0.708737 -0.966796
5  1.796565  2.950215 -0.947793
6 -0.360379 -1.212389  0.269852

```

```

[ ]: df = pd.DataFrame(np.random.standard_normal((6, 3)))
      df.iloc[2:, 1] = np.nan
      df.iloc[4:, 2] = np.nan
      df

```

```

[ ]:
      0      1      2
0  0.354313  1.344383 -0.526733
1  1.090709 -1.020539 -0.914861
2 -0.738487      NaN  0.948925
3  0.453201      NaN -0.483352
4 -0.110790      NaN      NaN
5  0.029640      NaN      NaN

```

```

[ ]: df.ffill()

```

```

[ ]:
      0      1      2
0  0.354313  1.344383 -0.526733
1  1.090709 -1.020539 -0.914861
2 -0.738487 -1.020539  0.948925
3  0.453201 -1.020539 -0.483352
4 -0.110790 -1.020539 -0.483352
5  0.029640 -1.020539 -0.483352

```

```

[ ]: df.ffill(limit=2)

```

```

[ ]:
      0      1      2
0  0.354313  1.344383 -0.526733
1  1.090709 -1.020539 -0.914861
2 -0.738487 -1.020539  0.948925
3  0.453201 -1.020539 -0.483352
4 -0.110790      NaN -0.483352
5  0.029640      NaN -0.483352

```

```

[ ]: data = pd.Series([1., np.nan, 3.5, np.nan, 7])
      data.fillna(data.mean())

```

```

[ ]: 0    1.000000
      1    3.833333
      2    3.500000
      3    3.833333
      4    7.000000
      dtype: float64

```

## Transformação de dados

```
[ ]: data = pd.DataFrame({"k1": ["one", "two"] * 3 + ["two"], "k2": [1, 1, 2, 3, 3, 4, 4]})
data
```

```
[ ]:      k1  k2
0  one   1
1  two   1
2  one   2
3  two   3
4  one   3
5  two   4
6  two   4
```

```
[ ]: data.duplicated()
```

```
[ ]: 0    False
1    False
2    False
3    False
4    False
5    False
6     True
dtype: bool
```

```
[ ]: data.drop_duplicates()
```

```
[ ]:      k1  k2
0  one   1
1  two   1
2  one   2
3  two   3
4  one   3
5  two   4
```

```
[ ]: data["v1"] = range(7)
data
```

```
[ ]:      k1  k2  v1
0  one   1    0
1  two   1    1
2  one   2    2
3  two   3    3
4  one   3    4
5  two   4    5
6  two   4    6
```

```
[ ]: data.drop_duplicates(subset=["k1"])
```

```
[ ]:      k1  k2  v1
0  one   1   0
1  two   1   1
```

```
[ ]: data.drop_duplicates(["k1", "k2"], keep="last")
```

```
[ ]:      k1  k2  v1
0  one   1   0
1  two   1   1
2  one   2   2
3  two   3   3
4  one   3   4
6  two   4   6
```

```
[ ]: data = pd.DataFrame({
    "food": ["bacon", "pulled pork", "bacon", "pastrami", "corned beef",
↪ "bacon", "pastrami", "honey ham", "nova lox"],
    "ounces": [4, 3, 12, 6, 7.5, 8, 3, 5, 6]
})
data
```

```
[ ]:      food  ounces
0     bacon    4.0
1 pulled pork    3.0
2     bacon   12.0
3   pastrami    6.0
4 corned beef    7.5
5     bacon    8.0
6   pastrami    3.0
7  honey ham    5.0
8   nova lox    6.0
```

```
[ ]: meat_to_animal = {
    "bacon": "pig",
    "pulled pork": "pig",
    "pastrami": "cow",
    "corned beef": "cow",
    "honey ham": "pig",
    "nova lox": "salmon"
}

data["animal"] = data["food"].map(meat_to_animal)
data
```

```
[ ]:      food  ounces  animal
0      bacon    4.0    pig
1  pulled pork    3.0    pig
2      bacon   12.0    pig
3    pastrami    6.0    cow
4  corned beef    7.5    cow
5      bacon    8.0    pig
6    pastrami    3.0    cow
7  honey ham     5.0    pig
8    nova lox    6.0  salmon
```

```
[ ]: def get_animal(x):
      return meat_to_animal[x]

data["food"].map(get_animal)
```

```
[ ]: 0    pig
1    pig
2    pig
3    cow
4    cow
5    pig
6    cow
7    pig
8  salmon
Name: food, dtype: object
```

```
[ ]: data = pd.Series([1., -999., 2., -999., -1000., 3.])
data
```

```
[ ]: 0    1.0
1   -999.0
2    2.0
3   -999.0
4  -1000.0
5    3.0
dtype: float64
```

```
[ ]: data.replace(-999, np.nan)
```

```
[ ]: 0    1.0
1    NaN
2    2.0
3    NaN
4  -1000.0
5    3.0
dtype: float64
```



```
[ ]: data.replace([-999, -1000], np.nan)
```

```
[ ]: 0    1.0  
     1    NaN  
     2    2.0  
     3    NaN  
     4    NaN  
     5    3.0  
     dtype: float64
```

```
[ ]: data.replace([-999, -1000], [np.nan, 0])
```

```
[ ]: 0    1.0  
     1    NaN  
     2    2.0  
     3    NaN  
     4    0.0  
     5    3.0  
     dtype: float64
```

```
[ ]: data.replace({-999: np.nan, -1000: 0})
```

```
[ ]: 0    1.0  
     1    NaN  
     2    2.0  
     3    NaN  
     4    0.0  
     5    3.0  
     dtype: float64
```

```
[ ]: data = pd.DataFrame(  
     np.arange(12).reshape((3, 4)),  
     index=["Ohio", "Colorado", "New York"],  
     columns=["one", "two", "three", "four"]  
 )  
data
```

```
[ ]:      one  two  three  four  
Ohio      0   1     2     3  
Colorado  4   5     6     7  
New York  8   9    10    11
```

```
[ ]: def transform(x):  
     return x[:4].upper()  
  
data.index.map(transform)
```

```
[ ]: Index(['OHIO', 'COLO', 'NEW '], dtype='object')
```

```
[ ]: data.index = data.index.map(transform)
data
```

```
[ ]:
      one  two  three  four
OHIO    0   1     2    3
COLO    4   5     6    7
NEW     8   9    10   11
```

```
[ ]: data.rename(index=str.title, columns=str.upper)
```

```
[ ]:
      ONE  TWO  THREE  FOUR
Ohio    0   1     2    3
Colo    4   5     6    7
New     8   9    10   11
```

```
[ ]: data.rename(index={"OHIO": "INDIANA"}, columns={"three": "peekaboo"})
```

```
[ ]:
      one  two  peekaboo  four
INDIANA  0   1         2    3
COLO     4   5         6    7
NEW      8   9        10   11
```

```
[ ]: ages = [20, 22, 25, 27, 21, 23, 37, 31, 61, 45, 41, 32]
```

```
[ ]: bins = [18, 25, 35, 60, 100]
age_categories = pd.cut(ages, bins)
age_categories
```

```
[ ]: [(18, 25], (18, 25], (18, 25], (25, 35], (18, 25], ..., (25, 35], (60, 100],
(35, 60], (35, 60], (25, 35]]
Length: 12
Categories (4, interval[int64, right]): [(18, 25] < (25, 35] < (35, 60] < (60, 100]]
```

```
[ ]: age_categories.codes
```

```
[ ]: array([0, 0, 0, 1, 0, 0, 2, 1, 3, 2, 2, 1], dtype=int8)
```

```
[ ]: age_categories.categories
```

```
[ ]: IntervalIndex([(18, 25], (25, 35], (35, 60], (60, 100]], dtype='interval[int64, right]')
```

```
[ ]: age_categories.categories[0]
```

```
[ ]: Interval(18, 25, closed='right')
```

```
[ ]: pd.value_counts(age_categories)
```

```
/tmp/ipykernel_29990/3010498523.py:1: FutureWarning: pandas.value_counts is
deprecated and will be removed in a future version. Use
pd.Series(obj).value_counts() instead.
  pd.value_counts(age_categories)
```

```
[ ]: (18, 25]      5
     (25, 35]      3
     (35, 60]      3
     (60, 100]     1
     Name: count, dtype: int64
```

```
[ ]: pd.cut(ages, bins, right=False)
```

```
[ ]: [[18, 25), [18, 25), [25, 35), [25, 35), [18, 25), ..., [25, 35), [60, 100),
     [35, 60), [35, 60), [25, 35)]
     Length: 12
     Categories (4, interval[int64, left]): [[18, 25) < [25, 35) < [35, 60) < [60,
     100)]
```

```
[ ]: group_names = ["Youth", "YoungAdult", "MiddleAged", "Senior"]
     pd.cut(ages, bins, labels=group_names)
```

```
[ ]: ['Youth', 'Youth', 'Youth', 'YoungAdult', 'Youth', ..., 'YoungAdult', 'Senior',
     'MiddleAged', 'MiddleAged', 'YoungAdult']
     Length: 12
     Categories (4, object): ['Youth' < 'YoungAdult' < 'MiddleAged' < 'Senior']
```

```
[ ]: data = np.random.uniform(size=20)
     pd.cut(data, 4, precision=2)
```

```
[ ]: [(0.086, 0.31], (0.77, 1.0], (0.54, 0.77], (0.086, 0.31], (0.086, 0.31], ...,
     (0.77, 1.0], (0.086, 0.31], (0.086, 0.31], (0.31, 0.54], (0.31, 0.54)]
     Length: 20
     Categories (4, interval[float64, right]): [(0.086, 0.31] < (0.31, 0.54] < (0.54,
     0.77] < (0.77, 1.0)]
```

```
[ ]: data = np.random.standard_normal(1000)
     quartiles = pd.qcut(data, 4, precision=2)
     quartiles
```

```
[ ]: [(0.63, 3.93], (-2.96, -0.7], (-0.014, 0.63], (-0.014, 0.63], (0.63, 3.93], ...,
     (-0.014, 0.63], (-0.014, 0.63], (-0.7, -0.014], (0.63, 3.93], (-0.014, 0.63)]
     Length: 1000
```

```
Categories (4, interval[float64, right]): [(-2.96, -0.7] < (-0.7, -0.014] <
(-0.014, 0.63] < (0.63, 3.93]]
```

```
[ ]: pd.value_counts(quartiles)
```

```
/tmp/ipykernel_29990/3472704981.py:1: FutureWarning: pandas.value_counts is
deprecated and will be removed in a future version. Use
pd.Series(obj).value_counts() instead.
  pd.value_counts(quartiles)
```

```
[ ]: (-2.96, -0.7]      250
      (-0.7, -0.014]   250
      (-0.014, 0.63]   250
      (0.63, 3.93]     250
      Name: count, dtype: int64
```

```
[ ]: pd.qcut(data, [0, 0.1, 0.5, 0.9, 1.]).value_counts()
```

```
[ ]: (-2.979, -1.342]   100
      (-1.342, 0.0119]  400
      (0.0119, 1.338]   400
      (1.338, 2.832]    100
      Name: count, dtype: int64
```

```
[ ]: data = pd.DataFrame(np.random.standard_normal((1000, 4)))
      data.describe()
```

```
[ ]:
```

	0	1	2	3
count	1000.000000	1000.000000	1000.000000	1000.000000
mean	0.022434	-0.004741	-0.058187	0.044673
std	1.008762	0.995184	0.991329	0.997007
min	-3.184377	-3.745356	-3.428254	-3.645860
25%	-0.628122	-0.697084	-0.747478	-0.599807
50%	-0.013609	-0.029924	-0.091364	0.047101
75%	0.695298	0.694459	0.618965	0.740562
max	3.525865	2.735527	3.366626	2.653656

```
[ ]: col = data[2]
      col[col.abs() > 3]
```

```
[ ]: 270    -3.428254
      647     3.366626
      Name: 2, dtype: float64
```

```
[ ]: data[(data.abs() > 3).any(axis="columns")]
```

```
[ ]:      0      1      2      3
53 -0.025907 -3.399312 -0.974657 -0.685312
72  3.260383  0.963301  1.201206 -1.852001
148 -0.196713 -3.745356 -1.520113 -0.346839
247 -3.056990  1.918403 -0.578828  1.847446
270  0.326045  0.425384 -3.428254 -0.296336
334 -3.184377  1.369891 -1.074833 -0.089937
555  0.208011 -0.150923 -0.362528 -3.548824
647  0.193299  1.397822  3.366626 -2.372214
794  3.525865  0.283070  0.544635  0.462204
814 -0.450721 -0.080332  0.599947 -3.645860
```

```
[ ]: data[data.abs() > 3] = np.sign(data) * 3
data.describe()
```

```
[ ]:      0      1      2      3
count  1000.000000  1000.000000  1000.000000  1000.000000
mean      0.021889  -0.003596  -0.058126   0.045868
std       1.005520   0.991368   0.988761   0.992986
min      -3.000000  -3.000000  -3.000000  -3.000000
25%      -0.628122  -0.697084  -0.747478  -0.599807
50%      -0.013609  -0.029924  -0.091364   0.047101
75%       0.695298   0.694459   0.618965   0.740562
max       3.000000   2.735527   3.000000   2.653656
```

```
[ ]: np.sign(data).head()
```

```
[ ]:      0      1      2      3
0  1.0 -1.0 -1.0  1.0
1  1.0 -1.0  1.0  1.0
2 -1.0 -1.0 -1.0 -1.0
3  1.0  1.0  1.0 -1.0
4  1.0  1.0 -1.0 -1.0
```

```
[ ]: df = pd.DataFrame(np.arange(5 * 7).reshape((5, 7)))
df
```

```
[ ]:      0      1      2      3      4      5      6
0      0      1      2      3      4      5      6
1      7      8      9     10     11     12     13
2     14     15     16     17     18     19     20
3     21     22     23     24     25     26     27
4     28     29     30     31     32     33     34
```

```
[ ]: sampler = np.random.permutation(5)
sampler
```

```
[ ]: array([1, 3, 2, 4, 0])
```

```
[ ]: df.take(sampler)
```

```
[ ]:      0  1  2  3  4  5  6
1    7  8  9 10 11 12 13
3   21 22 23 24 25 26 27
2   14 15 16 17 18 19 20
4   28 29 30 31 32 33 34
0    0  1  2  3  4  5  6
```

```
[ ]: df.iloc[sampler]
```

```
[ ]:      0  1  2  3  4  5  6
1    7  8  9 10 11 12 13
3   21 22 23 24 25 26 27
2   14 15 16 17 18 19 20
4   28 29 30 31 32 33 34
0    0  1  2  3  4  5  6
```

```
[ ]: column_sampler = np.random.permutation(7)
      column_sampler
```

```
[ ]: array([6, 1, 2, 4, 5, 0, 3])
```

```
[ ]: df.take(column_sampler, axis="columns")
```

```
[ ]:      6  1  2  4  5  0  3
0    6  1  2  4  5  0  3
1   13  8  9 11 12  7 10
2   20 15 16 18 19 14 17
3   27 22 23 25 26 21 24
4   34 29 30 32 33 28 31
```

```
[ ]: df.sample(n=3)
```

```
[ ]:      0  1  2  3  4  5  6
4   28 29 30 31 32 33 34
0    0  1  2  3  4  5  6
1    7  8  9 10 11 12 13
```

```
[ ]: choices = pd.Series([5, 7, -1, 6, 4])
      choices.sample(n=10, replace=True)
```

```
[ ]: 3    6
     2   -1
     4    4
```

```

4    4
3    6
0    5
3    6
1    7
4    4
0    5
dtype: int64

```

```

[ ]: df = pd.DataFrame({
      "key": ["b", "b", "a", "c", "a", "b"],
      "data1": range(6)
    })
df

```

```

[ ]:   key  data1
0    b      0
1    b      1
2    a      2
3    c      3
4    a      4
5    b      5

```

```

[ ]: pd.get_dummies(df["key"], dtype=float)

```

```

[ ]:   a    b    c
0  0.0  1.0  0.0
1  0.0  1.0  0.0
2  1.0  0.0  0.0
3  0.0  0.0  1.0
4  1.0  0.0  0.0
5  0.0  1.0  0.0

```

```

[ ]: dummies = pd.get_dummies(df["key"], prefix="key", dtype=float)
df_with_dummy = df[["data1"]].join(dummies)
df_with_dummy

```

```

[ ]:   data1  key_a  key_b  key_c
0      0    0.0    1.0    0.0
1      1    0.0    1.0    0.0
2      2    1.0    0.0    0.0
3      3    0.0    0.0    1.0
4      4    1.0    0.0    0.0
5      5    0.0    1.0    0.0

```

```

[ ]: mnames = ["movie_id", "title", "genres"]
movies = pd.read_table(base_url + "datasets/movielens/movies.dat", sep="::",

```

```
header=None, names=mnames, engine="python")
movies[:10]
```

```
[ ]:  movie_id          title          genres
      0          1      Toy Story (1995)  Animation|Children's|Comedy
      1          2      Jumanji (1995)    Adventure|Children's|Fantasy
      2          3      Grumpier Old Men (1995)  Comedy|Romance
      3          4      Waiting to Exhale (1995)  Comedy|Drama
      4          5  Father of the Bride Part II (1995)  Comedy
      5          6          Heat (1995)      Action|Crime|Thriller
      6          7      Sabrina (1995)      Comedy|Romance
      7          8      Tom and Huck (1995)    Adventure|Children's
      8          9      Sudden Death (1995)    Action
      9         10      GoldenEye (1995)      Action|Adventure|Thriller
```

```
[ ]: dummies = movies["genres"].str.get_dummies("|")
      dummies.iloc[:10, :6]
```

```
[ ]:  Action  Adventure  Animation  Children's  Comedy  Crime
      0      0         0         1         1         1      0
      1      0         1         0         1         0      0
      2      0         0         0         0         1      0
      3      0         0         0         0         1      0
      4      0         0         0         0         1      0
      5      1         0         0         0         0      1
      6      0         0         0         0         1      0
      7      0         1         0         1         0      0
      8      1         0         0         0         0      0
      9      1         1         0         0         0      0
```

```
[ ]: movies_windic = movies.join(dummies.add_prefix("Genre_"))
      movies_windic.iloc[0]
```

```
[ ]: movie_id          1
      title          Toy Story (1995)
      genres      Animation|Children's|Comedy
      Genre_Action          0
      Genre_Adventure          0
      Genre_Animation          1
      Genre_Children's          1
      Genre_Comedy          1
      Genre_Crime          0
      Genre_Documentary          0
      Genre_Drama          0
      Genre_Fantasy          0
      Genre_Film-Noir          0
      Genre_Horror          0
```



```
Genre_Musical          0
Genre_Mystery          0
Genre_Romance          0
Genre_Sci-Fi          0
Genre_Thriller         0
Genre_War              0
Genre_Western          0
Name: 0, dtype: object
```

```
[ ]: np.random.seed(12345) # to make the example repeatable
values = np.random.uniform(size=10)
values
```

```
[ ]: array([0.92961609, 0.31637555, 0.18391881, 0.20456028, 0.56772503,
          0.5955447 , 0.96451452, 0.6531771 , 0.74890664, 0.65356987])
```

```
[ ]: bins = [0, 0.2, 0.4, 0.6, 0.8, 1]
pd.get_dummies(pd.cut(values, bins))
```

```
[ ]:      (0.0, 0.2]  (0.2, 0.4]  (0.4, 0.6]  (0.6, 0.8]  (0.8, 1.0]
0      False      False      False      False      True
1      False      True       False      False      False
2       True      False      False      False      False
3      False      True       False      False      False
4      False      False      True       False      False
5      False      False      True       False      False
6      False      False      False      False      True
7      False      False      False      True       False
8      False      False      False      True       False
9      False      False      False      True       False
```

### Tipos de dados de extensão

```
[ ]: s = pd.Series([1, 2, 3, None])
s
```

```
[ ]: 0    1.0
1    2.0
2    3.0
3    NaN
dtype: float64
```

```
[ ]: s.dtype
```

```
[ ]: dtype('float64')
```

```
[ ]: s = pd.Series([1, 2, 3, None], dtype=pd.Int64Dtype())
s
```

```
[ ]: 0      1
     1      2
     2      3
     3    <NA>
     dtype: Int64
```

```
[ ]: s.isna()
```

```
[ ]: 0    False
     1    False
     2    False
     3     True
     dtype: bool
```

```
[ ]: s.dtype
```

```
[ ]: Int64Dtype()
```

```
[ ]: s[3]
```

```
[ ]: <NA>
```

```
[ ]: s[3] is pd.NA
```

```
[ ]: True
```

```
[ ]: s = pd.Series([1, 2, 3, None], dtype="Int64")
```

```
[ ]: s = pd.Series(['one', 'two', None, 'three'], dtype=pd.StringDtype())
s
```

```
[ ]: 0      one
     1      two
     2    <NA>
     3     three
     dtype: string
```

```
[ ]: df = pd.DataFrame({"A": [1, 2, None, 4],
                        "B": ["one", "two", "three", None],
                        "C": [False, None, False, True]})
df
```

```
[ ]:      A      B      C
0  1.0    one  False
```

```

1  2.0    two  None
2  NaN   three False
3  4.0    None  True

```

```

[ ]: df["A"] = df["A"].astype("Int64")
     df["B"] = df["B"].astype("string")
     df["C"] = df["C"].astype("boolean")
     df

```

```

[ ]:      A      B      C
     0      1      one  False
     1      2      two  <NA>
     2  <NA>  three  False
     3      4  <NA>   True

```

### Manipulação de string

```

[ ]: val = "a,b, guido"
     val.split(",")

```

```

[ ]: ['a', 'b', ' guido']

```

```

[ ]: pieces = [x.strip() for x in val.split(",")]
     pieces

```

```

[ ]: ['a', 'b', 'guido']

```

```

[ ]: first, second, third = pieces
     first + "::" + second + "::" + third

```

```

[ ]: 'a::b::guido'

```

```

[ ]: "::".join(pieces)

```

```

[ ]: 'a::b::guido'

```

```

[ ]: "guido" in val

```

```

[ ]: True

```

```

[ ]: val.index(",")

```

```

[ ]: 1

```

```

[ ]: val.find("::")

```

```

[ ]: -1

```

```
[ ]: val.index(":")
```

```
-----  
ValueError                                Traceback (most recent call last)  
/tmp/ipykernel_29990/2601145560.py in <module>  
----> 1 val.index(":")  
  
ValueError: substring not found
```

```
[ ]: val.count(",")
```

```
[ ]: 2
```

```
[ ]: val.replace(",", " ::")
```

```
[ ]: 'a::b:: guido'
```

```
[ ]: val.replace(",", "")
```

```
[ ]: 'ab guido'
```

### Expressões regulares

```
[ ]: import re  
text = "foo    bar\t baz  \tqux"  
re.split(r"\s+", text)
```

```
[ ]: ['foo', 'bar', 'baz', 'qux']
```

```
[ ]: regex = re.compile(r"\s+")  
regex.split(text)
```

```
[ ]: ['foo', 'bar', 'baz', 'qux']
```

```
[ ]: regex.findall(text)
```

```
[ ]: [' ', '\t ', ' \t']
```

```
[ ]: text = """Dave dave@google.com  
Steve steve@gmail.com  
Rob rob@gmail.com  
Ryan ryan@yahoo.com"""  
pattern = r"[A-Z0-9._%+-]+@[A-Z0-9.-]+\.[A-Z]{2,4}"  
  
# re.IGNORECASE makes the regex case insensitive  
regex = re.compile(pattern, flags=re.IGNORECASE)
```

```
[ ]: regex.findall(text)

[ ]: ['dave@google.com', 'steve@gmail.com', 'rob@gmail.com', 'ryan@yahoo.com']
```

```
[ ]: m = regex.search(text)
      m
```

```
[ ]: <re.Match object; span=(5, 20), match='dave@google.com'>
```

```
[ ]: text[m.start():m.end()]
```

```
[ ]: 'dave@google.com'
```

```
[ ]: print(regex.match(text))
```

None

```
[ ]: print(regex.sub("REDACTED", text))
```

Dave REDACTED  
Steve REDACTED  
Rob REDACTED  
Ryan REDACTED

```
[ ]: pattern = r"([A-Z0-9._%+-]+)@([A-Z0-9.-]+)\.([A-Z]{2,4})"
      regex = re.compile(pattern, flags=re.IGNORECASE)
```

```
[ ]: m = regex.match("wesm@bright.net")
      m.groups()
```

```
[ ]: ('wesm', 'bright', 'net')
```

```
[ ]: regex.findall(text)
```

```
[ ]: [('dave', 'google', 'com'),
      ('steve', 'gmail', 'com'),
      ('rob', 'gmail', 'com'),
      ('ryan', 'yahoo', 'com')]
```

```
[ ]: print(regex.sub(r"Username: \1, Domain: \2, Suffix: \3", text))
```

Dave Username: dave, Domain: google, Suffix: com  
Steve Username: steve, Domain: gmail, Suffix: com  
Rob Username: rob, Domain: gmail, Suffix: com  
Ryan Username: ryan, Domain: yahoo, Suffix: com

**Funções de strings em pandas**

```
[ ]: data = {"Dave": "dave@google.com", "Steve": "steve@gmail.com",  
            "Rob": "rob@gmail.com", "Wes": np.nan}  
data = pd.Series(data)  
data
```

```
[ ]: Dave      dave@google.com  
     Steve    steve@gmail.com  
     Rob      rob@gmail.com  
     Wes              NaN  
dtype: object
```

```
[ ]: data.isna()
```

```
[ ]: Dave      False  
     Steve    False  
     Rob      False  
     Wes       True  
dtype: bool
```

```
[ ]: data.str.contains("gmail")
```

```
[ ]: Dave      False  
     Steve     True  
     Rob       True  
     Wes       NaN  
dtype: object
```

```
[ ]: data_as_string_ext = data.astype('string')  
data_as_string_ext
```

```
[ ]: Dave      dave@google.com  
     Steve    steve@gmail.com  
     Rob      rob@gmail.com  
     Wes              <NA>  
dtype: string
```

```
[ ]: data_as_string_ext.str.contains("gmail")
```

```
[ ]: Dave      False  
     Steve     True  
     Rob       True  
     Wes       <NA>  
dtype: boolean
```

```
[ ]: pattern = r"([A-Z0-9._%+-]+)@([A-Z0-9.-]+\.[A-Z]{2,4})"  
data.str.findall(pattern, flags=re.IGNORECASE)
```

```
[ ]: Dave      [(dave, google, com)]
      Steve    [(steve, gmail, com)]
      Rob      [(rob, gmail, com)]
      Wes      NaN
      dtype: object
```

```
[ ]: matches = data.str.findall(pattern, flags=re.IGNORECASE).str[0]
      matches
```

```
[ ]: Dave      dave@google.com
      Steve     steve@gmail.com
      Rob       rob@gmail.com
      Wes       NaN
      dtype: object
```

```
[ ]: matches.str.get(1)
```

```
[ ]: Dave      a
      Steve     t
      Rob       o
      Wes       NaN
      dtype: object
```

```
[ ]: data.str[:5]
```

```
[ ]: Dave      dave@
      Steve     steve
      Rob       rob@g
      Wes       NaN
      dtype: object
```

```
[ ]: data.str.extract(pattern, flags=re.IGNORECASE)
```

```
[ ]:           0      1      2
      Dave      dave  google  com
      Steve     steve   gmail  com
      Rob       rob    gmail  com
      Wes       NaN     NaN   NaN
```

### Dados categóricos

```
[ ]: values = pd.Series(['apple', 'orange', 'apple',
                        'apple'] * 2)
      values
```

```
[ ]: 0      apple
      1      orange
      2      apple
```

```
3    apple
4    apple
5    orange
6    apple
7    apple
dtype: object
```

```
[ ]: pd.unique(values)
```

```
[ ]: array(['apple', 'orange'], dtype=object)
```

```
[ ]: pd.value_counts(values)
```

```
/tmp/ipykernel_29990/3297668723.py:1: FutureWarning: pandas.value_counts is
deprecated and will be removed in a future version. Use
pd.Series(obj).value_counts() instead.
  pd.value_counts(values)
```

```
[ ]: apple    6
     orange    2
     Name: count, dtype: int64
```

```
[ ]: values = pd.Series([0, 1, 0, 0] * 2)
     dim = pd.Series(['apple', 'orange'])
     values
```

```
[ ]: 0    0
     1    1
     2    0
     3    0
     4    0
     5    1
     6    0
     7    0
     dtype: int64
```

```
[ ]: dim
```

```
[ ]: 0    apple
     1    orange
     dtype: object
```

```
[ ]: dim.take(values)
```

```
[ ]: 0    apple
     1    orange
     0    apple
```



```
0    apple
0    apple
1    orange
0    apple
0    apple
dtype: object
```

```
[ ]: fruits = ['apple', 'orange', 'apple', 'apple'] * 2
N = len(fruits)
rng = np.random.default_rng(seed=12345)
df = pd.DataFrame({'fruit': fruits,
                   'basket_id': np.arange(N),
                   'count': rng.integers(3, 15, size=N),
                   'weight': rng.uniform(0, 4, size=N)},
                  columns=['basket_id', 'fruit', 'count', 'weight'])
df
```

```
[ ]:  basket_id  fruit  count  weight
0         0   apple    11  1.564438
1         1  orange     5  1.331256
2         2   apple    12  2.393235
3         3   apple     6  0.746937
4         4   apple     5  2.691024
5         5  orange    12  3.767211
6         6   apple    10  0.992983
7         7   apple    11  3.795525
```

```
[ ]: fruit_cat = df['fruit'].astype('category')
fruit_cat
```

```
[ ]: 0    apple
1    orange
2    apple
3    apple
4    apple
5    orange
6    apple
7    apple
Name: fruit, dtype: category
Categories (2, object): ['apple', 'orange']
```

```
[ ]: c = fruit_cat.array
type(c)
```

```
[ ]: pandas.core.arrays.categorical.Categorical
```

```
[ ]: c.categories
```

```
[ ]: Index(['apple', 'orange'], dtype='object')
```

```
[ ]: c.codes
```

```
[ ]: array([0, 1, 0, 0, 0, 1, 0, 0], dtype=int8)
```

```
[ ]: dict(enumerate(c.categories))
```

```
[ ]: {0: 'apple', 1: 'orange'}
```

```
[ ]: df['fruit'] = df['fruit'].astype('category')
df["fruit"]
```

```
[ ]: 0    apple
    1    orange
    2    apple
    3    apple
    4    apple
    5    orange
    6    apple
    7    apple
    Name: fruit, dtype: category
    Categories (2, object): ['apple', 'orange']
```

```
[ ]: my_categories = pd.Categorical(['foo', 'bar', 'baz', 'foo', 'bar'])
my_categories
```

```
[ ]: ['foo', 'bar', 'baz', 'foo', 'bar']
Categories (3, object): ['bar', 'baz', 'foo']
```

```
[ ]: categories = ['foo', 'bar', 'baz']
    codes = [0, 1, 2, 0, 0, 1]
    my_cats_2 = pd.Categorical.from_codes(codes, categories)
    my_cats_2
```

```
[ ]: ['foo', 'bar', 'baz', 'foo', 'foo', 'bar']
Categories (3, object): ['foo', 'bar', 'baz']
```

```
[ ]: ordered_cat = pd.Categorical.from_codes(codes, categories, ordered=True)
ordered_cat
```

```
[ ]: ['foo', 'bar', 'baz', 'foo', 'foo', 'bar']
Categories (3, object): ['foo' < 'bar' < 'baz']
```

```
[ ]: my_cats_2.as_ordered()
```

```
[ ]: ['foo', 'bar', 'baz', 'foo', 'foo', 'bar']
Categories (3, object): ['foo' < 'bar' < 'baz']
```

```
[ ]: rng = np.random.default_rng(seed=12345)
draws = rng.standard_normal(1000)
draws[:5]
```

```
[ ]: array([-1.42382504,  1.26372846, -0.87066174, -0.25917323, -0.07534331])
```

```
[ ]: bins = pd.qcut(draws, 4)
bins
```

```
[ ]: [(-3.121, -0.675], (0.687, 3.211], (-3.121, -0.675], (-0.675, 0.0134], (-0.675,
0.0134], ..., (0.0134, 0.687], (0.0134, 0.687], (-0.675, 0.0134], (0.0134,
0.687], (-0.675, 0.0134]]
Length: 1000
Categories (4, interval[float64, right]): [(-3.121, -0.675] < (-0.675, 0.0134] <
(0.0134, 0.687] < (0.687, 3.211]]
```

```
[ ]: bins = pd.qcut(draws, 4, labels=['Q1', 'Q2', 'Q3', 'Q4'])
bins
```

```
[ ]: ['Q1', 'Q4', 'Q1', 'Q2', 'Q2', ..., 'Q3', 'Q3', 'Q2', 'Q3', 'Q2']
Length: 1000
Categories (4, object): ['Q1' < 'Q2' < 'Q3' < 'Q4']
```

```
[ ]: bins.codes[:10]
```

```
[ ]: array([0, 3, 0, 1, 1, 0, 0, 2, 2, 0], dtype=int8)
```

```
[ ]: bins = pd.Series(bins, name='quartile')
results = (pd.Series(draws)
           .groupby(bins)
           .agg(['count', 'min', 'max'])
           .reset_index())
results
```

/tmp/ipykernel\_29990/2483392743.py:3: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

```
.groupby(bins)
```

```
[ ]:  quartile  count      min      max
0      Q1     250 -3.119609 -0.678494
1      Q2     250 -0.673305  0.008009
2      Q3     250  0.018753  0.686183
```

```
3      Q4      250  0.688282  3.211418
```

```
[ ]: results['quartile']
```

```
[ ]: 0      Q1
      1      Q2
      2      Q3
      3      Q4
      Name: quartile, dtype: category
      Categories (4, object): ['Q1' < 'Q2' < 'Q3' < 'Q4']
```

```
[ ]: N = 10_000_000
      labels = pd.Series(['foo', 'bar', 'baz', 'qux'] * (N // 4))
      categories = labels.astype('category')
      labels.memory_usage(deep=True)
```

```
[ ]: 600000128
```

```
[ ]: categories.memory_usage(deep=True)
```

```
[ ]: 10000540
```

```
[ ]: %time _ = labels.astype('category')
```

```
CPU times: user 833 ms, sys: 364 ms, total: 1.2 s
Wall time: 1.26 s
```

```
[ ]: %timeit labels.value_counts()
```

```
808 ms ± 160 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
```

```
[ ]: %timeit categories.value_counts()
```

```
205 ms ± 39 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
```

```
[ ]: s = pd.Series(['a', 'b', 'c', 'd'] * 2)
      cat_s = s.astype('category')
      cat_s
```

```
[ ]: 0      a
      1      b
      2      c
      3      d
      4      a
      5      b
      6      c
      7      d
      dtype: category
```

```
Categories (4, object): ['a', 'b', 'c', 'd']
```

```
[ ]: cat_s.cat.codes
```

```
[ ]: 0    0
      1    1
      2    2
      3    3
      4    0
      5    1
      6    2
      7    3
      dtype: int8
```

```
[ ]: cat_s.cat.categories
```

```
[ ]: Index(['a', 'b', 'c', 'd'], dtype='object')
```

```
[ ]: actual_categories = ['a', 'b', 'c', 'd', 'e']
      cat_s2 = cat_s.cat.set_categories(actual_categories)
      cat_s2
```

```
[ ]: 0    a
      1    b
      2    c
      3    d
      4    a
      5    b
      6    c
      7    d
      dtype: category
      Categories (5, object): ['a', 'b', 'c', 'd', 'e']
```

```
[ ]: cat_s.value_counts()
```

```
[ ]: a    2
      b    2
      c    2
      d    2
      Name: count, dtype: int64
```

```
[ ]: cat_s2.value_counts()
```

```
[ ]: a    2
      b    2
      c    2
      d    2
```

```
e    0
Name: count, dtype: int64
```

```
[ ]: cat_s3 = cat_s[cat_s.isin(['a', 'b'])]
cat_s3
```

```
[ ]: 0    a
     1    b
     4    a
     5    b
     dtype: category
     Categories (4, object): ['a', 'b', 'c', 'd']
```

```
[ ]: cat_s3.cat.remove_unused_categories()
```

```
[ ]: 0    a
     1    b
     4    a
     5    b
     dtype: category
     Categories (2, object): ['a', 'b']
```

```
[ ]: cat_s = pd.Series(['a', 'b', 'c', 'd'] * 2, dtype='category')
cat_s
```

```
[ ]: 0    a
     1    b
     2    c
     3    d
     4    a
     5    b
     6    c
     7    d
     dtype: category
     Categories (4, object): ['a', 'b', 'c', 'd']
```

```
[ ]: pd.get_dummies(cat_s, dtype=float)
```

```
[ ]:      a    b    c    d
0  1.0  0.0  0.0  0.0
1  0.0  1.0  0.0  0.0
2  0.0  0.0  1.0  0.0
3  0.0  0.0  0.0  1.0
4  1.0  0.0  0.0  0.0
5  0.0  1.0  0.0  0.0
6  0.0  0.0  1.0  0.0
7  0.0  0.0  0.0  1.0
```