```
In [2]: #1-General view, picture,
                                          (Genel Resim)
In [2]:
         import pandas as pd
          import seaborn as sns
          import numpy as np
          import matplotlib.pyplot as plt
          pd.set option('display.max columns', None)
          pd.set_option('display.width',500)
          df=sns.load_dataset('titanic')
In [3]:
         df.head()
Out[3]:
              survived
                      pclass
                                 sex
                                      age
                                           sibsp
                                                  parch
                                                             fare
                                                                 embarked
                                                                            class
                                                                                     who
                                                                                           adult_male
                                                                                                      deck
                                                                                                            embark_
          0
                    0
                            3
                                male
                                      22.0
                                                          7.2500
                                                                         S
                                                                             Third
                                                                                     man
                                                                                                 True
                                                                                                       NaN
                                                                                                             Southan
          1
                    1
                            1
                              female
                                      38.0
                                               1
                                                      0
                                                         71.2833
                                                                         С
                                                                             First woman
                                                                                                False
                                                                                                         С
                                                                                                               Cherk
          2
                                                                             Third
                            3
                              female
                                      26.0
                                               0
                                                      0
                                                          7.9250
                                                                         S
                                                                                   woman
                                                                                                False
                                                                                                       NaN
                                                                                                             Southan
           3
                                      35.0
                                                         53.1000
                                                                         S
                                                                                                False
                                                                                                             Southan
                            1
                               female
                                               1
                                                      0
                                                                             First
                                                                                   woman
                                                                                                         С
                    0
                            3
                                male
                                      35.0
                                                          8.0500
                                                                         S
                                                                             Third
                                                                                                 True
                                                                                                       NaN
                                                                                                             Southan
                                                                                     man
                                                                                                                  >
In [4]:
         df.tail()
Out[4]:
                survived pclass
                                             sibsp
                                                    parch
                                                             fare
                                                                 embarked
                                                                              class
                                                                                       who
                                                                                            adult_male
                                                                                                        deck embarl
                                   sex
                                         age
           886
                      0
                              2
                                        27.0
                                                 0
                                                           13.00
                                                                         S
                                                                            Second
                                                                                                         NaN
                                  male
                                                        0
                                                                                                  True
                                                                                                               Southa
                                                                                       man
           887
                      1
                                        19.0
                                                 0
                                                           30.00
                                                                         S
                                                                                                  False
                                                                                                           В
                              1
                                 female
                                                        0
                                                                               First woman
                                                                                                               Southa
           888
                      0
                                 female
                                        NaN
                                                           23.45
                                                                         S
                                                                               Third
                                                                                    woman
                                                                                                  False
                                                                                                         NaN
                                                                                                               Southa
                                                                         С
                                                                                                           С
          889
                      1
                              1
                                  male
                                        26.0
                                                 0
                                                        0
                                                           30.00
                                                                               First
                                                                                       man
                                                                                                  True
                                                                                                                 Che
                      0
                              3
                                        32.0
                                                                         Q
                                                                                                  True
                                                                                                         NaN
          890
                                                 0
                                                        0
                                                            7.75
                                                                               Third
                                  male
                                                                                       man
                                                                                                                Quee
                                                                                                                  >
In [6]:
         df.shape
```

Out[6]: (891, 15)

```
In [7]: | df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 891 entries, 0 to 890
          Data columns (total 15 columns):
                             Non-Null Count Dtype
               Column
               -----
           0
               survived
                             891 non-null
                                               int64
           1
                             891 non-null
                                               int64
               pclass
           2
               sex
                             891 non-null
                                               object
           3
                             714 non-null
                                               float64
               age
           4
                             891 non-null
                                               int64
               sibsp
           5
                             891 non-null
                                               int64
               parch
                             891 non-null
           6
               fare
                                               float64
           7
               embarked
                             889 non-null
                                               object
           8
                             891 non-null
               class
                                               category
           9
               who
                             891 non-null
                                               object
           10
               adult_male
                             891 non-null
                                               bool
           11
               deck
                             203 non-null
                                               category
               embark_town
                             889 non-null
                                               object
           13
                             891 non-null
                                               object
               alive
           14 alone
                             891 non-null
                                               bool
          dtypes: bool(2), category(2), float64(2), int64(4), object(5)
          memory usage: 80.7+ KB
 In [8]: | df.columns
 Out[8]: Index(['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embarked', 'clas
          s', 'who', 'adult_male', 'deck', 'embark_town', 'alive', 'alone'], dtype='object')
 In [9]: |df.index
 Out[9]: RangeIndex(start=0, stop=891, step=1)
In [10]:
          df.describe().T
Out[10]:
                   count
                                                     25%
                                                            50% 75%
                             mean
                                         std min
                                                                          max
                   891.0
           survived
                          0.383838
                                    0.486592 0.00
                                                   0.0000
                                                           0.0000
                                                                   1.0
                                                                        1.0000
             pclass
                    891.0
                          2.308642
                                    0.836071 1.00
                                                   2.0000
                                                           3.0000
                                                                  3.0
                                                                        3.0000
                    714.0
                          29.699118 14.526497 0.42
                                                          28.0000
                                                                  38.0
                                                                        80.0000
               age
                                                  20.1250
                    891.0
                                                   0.0000
                                                           0.0000
                                                                   1.0
                                                                        8.0000
             sibsp
                          0.523008
                                    1.102743 0.00
             parch
                    891.0
                          0.381594
                                    0.806057 0.00
                                                   0.0000
                                                           0.0000
                                                                  0.0
                                                                        6.0000
                    891.0 32.204208 49.693429 0.00
                                                   7.9104 14.4542 31.0 512.3292
In [13]: df.isnull().values.any()
Out[13]: True
```

```
In [14]: | df.isnull().sum()
Out[14]: survived
                          0
         pclass
                          0
                          0
         sex
         age
                        177
                          0
         sibsp
                          0
         parch
         fare
                          0
         embarked
                          2
         class
                          0
         who
                          0
         adult_male
                          0
                        688
         deck
         embark_town
                          2
         alive
                          0
         alone
                          0
         dtype: int64
In [17]: def check_df (dataframe, head=5):
             print("#####Shape#####")
             print(dataframe.shape)
             print("#####Types#####")
             print(dataframe.dtypes)
             print("#####Head#####")
             print(dataframe.head(head))
             print("#####Tail#####")
             print(dataframe.tail(head))
             print("#####NA#####")
             print(dataframe.isnull().sum())
             print("#####Quantiles####")
             print(dataframe.describe([0,0.05,0.50,0.95,0.99,1]).T)
```

In [18]: check\_df(df)

```
#####$hape#####
(891, 15)
#####Types#####
survived
                   int64
pclass
                   int64
sex
                  object
                 float64
age
                   int64
sibsp
parch
                   int64
                 float64
fare
embarked
                  object
class
                category
who
                  object
adult_male
                    bool
deck
                category
                  object
embark_town
                  object
alive
alone
                    bool
dtype: object
#####Head#####
                                     sibsp
                                             parch
                                                        fare embarked
                                                                        class
                                                                                       adult m
   survived pclass
                         sex
                                age
                                                                                  who
ale deck
          embark_town alive
                               alone
                        male
                               22.0
                                          1
                                                      7.2500
                                                                     S
                                                                        Third
                                                                                  man
                                                                                              Т
rue
     NaN
          Southampton
                           no
                               False
1
                   1
                      female
                               38.0
                                          1
                                                     71.2833
                                                                     C
                                                                        First
                                                                                woman
                                                                                             Fa
            Cherbourg
                         yes
lse
       C
                               False
2
                               26.0
                                                                     S
          1
                   3
                      female
                                          0
                                                      7.9250
                                                                        Third
                                                                                woman
                                                                                             Fa
1se
     NaN
          Southampton
                         yes
                                True
3
                                                     53.1000
                                                                     S
          1
                   1
                      female
                               35.0
                                          1
                                                                        First
                                                                                woman
                                                                                             Fa
lse
          Southampton
                         yes
                               False
4
                   3
                         male
                               35.0
                                          0
                                                 0
                                                      8.0500
                                                                     S
                                                                        Third
                                                                                  man
                                                                                              Т
     NaN Southampton
                                True
rue
#####Tail#####
                                                        fare embarked
                                                                         class
     survived pclass
                            sex
                                  age
                                        sibsp
                                               parch
                                                                                   who
                                                                                        adult
male deck
           embark town alive alone
886
             0
                     2
                           male
                                 27.0
                                            0
                                                       13.00
                                                                     S
                                                                        Second
                                                    0
                                                                                   man
True
      NaN
           Southampton
                            no
                                 True
                                                                                              F
887
             1
                     1
                        female
                                 19.0
                                            a
                                                   0
                                                       30.00
                                                                     S
                                                                         First
                                                                                 woman
alse
           Southampton
                          yes
                                 True
                                                                                              F
888
                     3
                        female
                                  NaN
                                            1
                                                   2
                                                       23.45
                                                                     S
                                                                         Third
                                                                                 woman
           Southampton
                               False
alse
      NaN
                            no
889
                                            0
                                                       30.00
                                                                     C
                                                                         First
                     1
                           male
                                26.0
                                                                                   man
        C
                           yes
                                 True
True
              Cherbourg
890
                           male
                                 32.0
                                            0
                                                        7.75
                                                                         Third
                     3
                                                                                   man
True NaN
                                 True
             Queenstown
                            no
######NA######
survived
                  0
pclass
                  0
sex
                  0
                177
age
                  0
sibsp
                  0
parch
fare
                  0
embarked
                  2
class
                  0
who
                  0
adult_male
                  0
                688
deck
embark_town
                  2
alive
                  0
alone
                  0
dtype: int64
```

```
######Quantiles#####
                                                0%
                                                       5%
                                                                50%
                                                                           95%
                                                                                      99%
          count
                                   std
                                         min
                      mean
100%
           max
survived
          891.0
                  0.383838
                             0.486592
                                        0.00
                                              0.00
                                                    0.000
                                                            0.0000
                                                                       1.00000
                                                                                  1.00000
1.0000
          1.0000
pclass
          891.0
                  2.308642
                             0.836071
                                        1.00
                                              1.00
                                                    1.000
                                                            3.0000
                                                                       3.00000
                                                                                  3.00000
3.0000
          3.0000
age
          714.0
                 29.699118
                            14.526497
                                        0.42
                                              0.42
                                                    4.000
                                                           28.0000
                                                                      56.00000
                                                                                 65.87000
80.0000
          80.0000
          891.0
                  0.523008
                             1.102743
                                        0.00
                                              0.00
                                                    0.000
                                                            0.0000
                                                                       3.00000
                                                                                  5.00000
sibsp
8.0000
          8.0000
                                                            0.0000
          891.0
parch
                  0.381594
                             0.806057
                                        0.00
                                              0.00
                                                    0.000
                                                                       2.00000
                                                                                  4.00000
6.0000
          6.0000
          891.0 32.204208
                            49.693429
                                        0.00 0.00 7.225
fare
                                                           14.4542 112.07915
                                                                                249.00622
512.3292 512.3292
```

## In [ ]: #2-Analysis of Categorical Variables

```
In [4]: import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
pd.set_option('display.max_columns', None)
pd.set_option('display.width',500)
df=sns.load_dataset("titanic")
```

## In [20]: df.head()

Out[20]:		survived	nelace	cov	200	eihen	parch	faro	embarked	class	who	adult male	dock	ombark :
		Suiviveu	pciass	367	aye	sinsh	parcii	laie	eiiibaikeu	Ciass	WIIO	auuit_iiiaie	ueck	ellibaik_
	0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southan
	1	1	1	female	38.0	1	0	71.2833	С	First	woman	False	С	Chert
	2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southan
	3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	С	Southan
	4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southan
	<													>

```
In [28]: df['survived'].value_counts()
```

Out[28]: survived 0 549 1 342

Name: count, dtype: int64

```
In [31]: df['sex'].unique()
```

Out[31]: array(['male', 'female'], dtype=object)

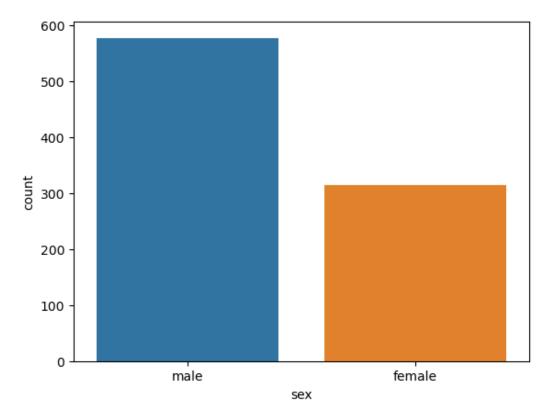
In [29]: check\_df(df)

```
#####$hape#####
(891, 15)
#####Types#####
survived
                   int64
pclass
                   int64
sex
                  object
                 float64
age
                   int64
sibsp
parch
                   int64
                 float64
fare
embarked
                  object
class
                category
who
                  object
adult_male
                    bool
deck
                category
                  object
embark_town
                  object
alive
alone
                    bool
dtype: object
#####Head#####
                                                        fare embarked
                                                                        class
                                                                                       adult m
   survived pclass
                          sex
                                age
                                     sibsp
                                             parch
                                                                                  who
ale deck
          embark_town alive
                               alone
                   3
                        male
                               22.0
                                          1
                                                      7.2500
                                                                     S
                                                                        Third
                                                                                  man
                                                                                              Т
rue
     NaN
          Southampton
                           no
                               False
1
                   1
                      female
                               38.0
                                          1
                                                     71.2833
                                                                     C
                                                                        First
                                                                                woman
                                                                                             Fa
            Cherbourg
                         yes
lse
       C
                               False
2
                               26.0
                                                                     S
          1
                   3
                      female
                                          0
                                                      7.9250
                                                                        Third
                                                                                woman
                                                                                             Fa
1se
     NaN
          Southampton
                          yes
                                True
3
                                                     53.1000
                                                                     S
          1
                   1
                      female
                               35.0
                                          1
                                                                        First
                                                                                woman
                                                                                             Fa
lse
          Southampton
                         yes
                               False
4
                   3
                         male
                               35.0
                                          0
                                                 0
                                                      8.0500
                                                                     S
                                                                        Third
                                                                                  man
                                                                                              Т
     NaN Southampton
                                True
rue
#####Tail#####
                                                        fare embarked
                                                                         class
     survived pclass
                            sex
                                  age
                                        sibsp
                                               parch
                                                                                   who
                                                                                         adult_
male deck
           embark town alive alone
886
             0
                                 27.0
                                            0
                                                                     S
                     2
                           male
                                                    0
                                                       13.00
                                                                        Second
                                                                                   man
True
      NaN
           Southampton
                            no
                                 True
                                                                                              F
887
             1
                     1
                        female
                                 19.0
                                            a
                                                    0
                                                       30.00
                                                                     S
                                                                         First
                                                                                 woman
alse
            Southampton
                           yes
                                 True
                                                                                              F
888
                     3
                        female
                                  NaN
                                            1
                                                    2
                                                       23.45
                                                                     S
                                                                         Third
                                                                                 woman
            Southampton
                               False
alse
      NaN
                            no
889
                                            0
                                                       30.00
                                                                     C
                                                                         First
                     1
                           male
                                26.0
                                                                                   man
        C
                           yes
                                 True
True
              Cherbourg
890
                           male
                                 32.0
                                            0
                                                        7.75
                                                                     0
                                                                         Third
                     3
                                                                                   man
True NaN
                                 True
             Queenstown
                            no
######NA######
survived
                  0
pclass
                  0
sex
                  0
                177
age
                  0
sibsp
                  0
parch
fare
                  0
embarked
                  2
class
                  0
who
                  0
adult_male
                  0
                688
deck
embark_town
                  2
alive
                  0
alone
                  0
dtype: int64
```

```
######Quantiles#####
                                                                                     95%
                                                                                                 99%
                    count
                                             std
                                                   min
                                                          0%
                                                                  5%
                                                                          50%
                                mean
          100%
                     max
                    891.0
                            0.383838
                                        0.486592
                                                  0.00
                                                        0.00
                                                               0.000
                                                                       0.0000
                                                                                 1.00000
                                                                                             1.00000
          survived
          1.0000
                    1.0000
          pclass
                    891.0
                            2.308642
                                        0.836071
                                                  1.00
                                                        1.00
                                                               1.000
                                                                       3.0000
                                                                                 3.00000
                                                                                             3.00000
          3.0000
                    3.0000
          age
                    714.0
                           29.699118
                                      14.526497
                                                  0.42
                                                        0.42
                                                              4.000
                                                                      28.0000
                                                                                56.00000
                                                                                            65.87000
          80.0000
                    80.0000
                    891.0
          sibsp
                            0.523008
                                        1.102743
                                                  0.00
                                                        0.00
                                                               0.000
                                                                       0.0000
                                                                                 3.00000
                                                                                             5.00000
                    8.0000
          8.0000
                    891.0
          parch
                            0.381594
                                        0.806057
                                                  0.00
                                                        0.00
                                                               0.000
                                                                       0.0000
                                                                                 2.00000
                                                                                             4.00000
          6.0000
                    6.0000
                    891.0 32.204208
          fare
                                      49.693429 0.00 0.00 7.225
                                                                      14.4542 112.07915
                                                                                           249.00622
          512.3292 512.3292
         cat cols = [col for col in df.columns if str(df[col].dtypes) in ["category", "object", "boo
In [32]:
          num but cat = [col for col in df.columns if df[col].nunique()<10 and df[col].dtypes in ["</pre>
          cat_but_car = [col for col in df.columns if df[col].nunique() > 20 and str(df[col].dtypes
          cat cols=cat cols+num but cat
In [33]: cat_cols
Out[33]: ['sex',
           embarked',
           'class',
           'who',
           'adult_male',
           'deck',
           'embark_town',
           'alive',
           'alone',
           'survived',
           'pclass',
           'sibsp',
           'parch']
 In [9]:
         def cat summmary (dataframe, col name):
              print(pd.DataFrame({col name:dataframe[col name].value counts(), "Ratio":100*dataframe
          cat summmary(df, "sex")
In [10]:
                           Ratio
                  sex
          sex
          male
                  577
                       64.758698
                       35.241302
          female
                  314
In [15]: | def cat_summary(dataframe, col_name, plot=False):
              print(pd.DataFrame({col_name: dataframe[col_name].value_counts(),
                                   "Ratio": 100*dataframe[col_name].value_counts()/len(dataframe)
                                 }))
              if plot:
                  sns.countplot(x=dataframe[col name],data=dataframe)
                  plt.show(block=True)
```

```
In [16]: cat_summary(df,"sex",plot=True)
```

sex Ratio sex male 577 64.758698 female 314 35.241302



In [17]: import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt
 pd.set\_option('display.max\_columns',None)
 pd.set\_option('display.width',500)
 df=sns.load\_dataset("titanic")
 df.head()

## Out[17]:

embark_	deck	adult_male	who	class	embarked	fare	parch	sibsp	age	sex	pclass	survived	
Southan	NaN	True	man	Third	S	7.2500	0	1	22.0	male	3	0	0
Chert	С	False	woman	First	С	71.2833	0	1	38.0	female	1	1	1
Southan	NaN	False	woman	Third	S	7.9250	0	0	26.0	female	3	1	2
Southan	С	False	woman	First	S	53.1000	0	1	35.0	female	1	1	3
Southan	NaN	True	man	Third	S	8.0500	0	0	35.0	male	3	0	4

In [18]: df[["age","fare"]].describe().T

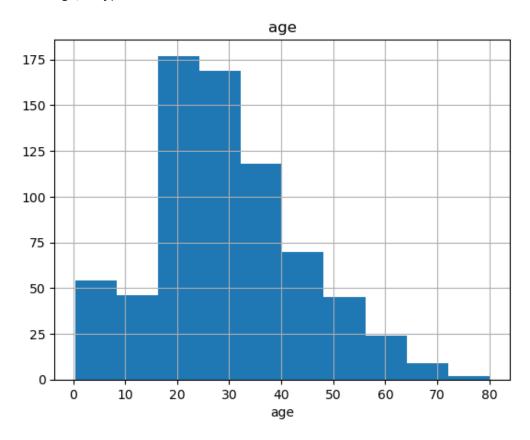
## Out[18]:

	count	mean	std	min	25%	50%	75%	max
age	714.0	29.699118	14.526497	0.42	20.1250	28.0000	38.0	80.0000
fare	891.0	32.204208	49.693429	0.00	7.9104	14.4542	31.0	512.3292

```
In [20]:
         num_cols=[col for col in df.columns if df[col].dtypes in ["int","float","int64"]]
In [21]: cat_cols = [col for col in df.columns if str(df[col].dtypes) in ["category", "object", "bool
         num_but_cat = [col for col in df.columns if df[col].nunique() < 10 and df[col].dtypes in</pre>
         cat but car = [col for col in df.columns if df[col].nunique() > 20 and str(df[col].dtypes
         cat_cols=cat_cols+num_but_cat
In [22]: cat cols= [col for col in cat cols if col not in cat but car]
In [23]: |num_cols=[col for col in df.columns if df[col].dtypes in ["int","float","int64"]]
         num_cols=[col for col in num_cols if col not in cat_cols]
In [24]: num_cols
Out[24]: ['age', 'fare']
In [32]: def num_summary (dataframe, num_cols):
             quantiles = [0.1, 0.3, 0.4, 0.75]
             print(dataframe[num_cols].describe(quantiles).T)
 In [ ]:
In [28]: num_summary(df, "age")
         count
                   714.000000
         mean
                    29.699118
         std
                    14.526497
         min
                    0.420000
         10%
                    14.000000
         30%
                    22.000000
         40%
                    25.000000
         50%
                    28.000000
         75%
                    38.000000
                    80.000000
         Name: age, dtype: float64
In [34]: def num summary(dataframe, numerical cols, plot=False):
             quantiles=[0.05,0.10,0.20,0.30,0.40,0.50,0.60,0.70,0.80,0.90,0.95,0.99]
             print(dataframe[numerical_cols].describe(quantiles).T)
             if plot:
                  dataframe[numerical cols].hist()
                  plt.xlabel(numerical_cols)
                  plt.title(numerical cols)
                  plt.show(block=True)
```

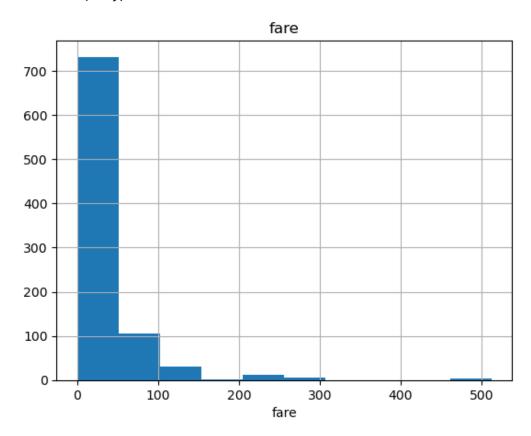
714.000000 count 29.699118 mean std 14.526497 min 0.420000 5% 4.000000 10% 14.000000 20% 19.000000 30% 22.000000 40% 25.000000 50% 28.000000 60% 31.800000 70% 36.000000 80% 41.000000 90% 50.000000 95% 56.000000 99% 65.870000 80.000000 max

Name: age, dtype: float64



count	891.000000
mean	32.204208
std	49.693429
min	0.000000
5%	7.225000
10%	7.550000
20%	7.854200
30%	8.050000
40%	10.500000
50%	14.454200
60%	21.679200
70%	27.000000
80%	39.687500
90%	77.958300
95%	112.079150
99%	249.006220
max	512.329200
_	

Name: fare, dtype: float64



```
In [36]: import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         pd.set_option('display.max_columns',None)
         pd.set option('display.width',500)
         df=sns.load dataset("titanic")
         df.head()
         df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 891 entries, 0 to 890
         Data columns (total 15 columns):
             Column
                           Non-Null Count Dtype
                           -----
         ---
             -----
          0
              survived
                           891 non-null
                                           int64
                                           int64
          1
                           891 non-null
              pclass
          2
              sex
                           891 non-null
                                           object
          3
                           714 non-null
                                           float64
              age
          4
                           891 non-null
                                           int64
              sibsp
          5
                           891 non-null
                                           int64
              parch
          6
                           891 non-null
                                           float64
              fare
          7
              embarked
                           889 non-null
                                           object
          8
              class
                           891 non-null
                                           category
          9
              who
                           891 non-null
                                           object
          10 adult_male 891 non-null
                                           bool
          11 deck
                           203 non-null
                                           category
          12 embark_town 889 non-null
                                           object
          13 alive
                           891 non-null
                                           object
          14 alone
                           891 non-null
                                           bool
         dtypes: bool(2), category(2), float64(2), int64(4), object(5)
         memory usage: 80.7+ KB
In [41]: def grab col names(dataframe, cat th=10, car th=20):
             cat_cols = [col for col in df.columns if str(df[col].dtypes) in ["category","object",
             num_but_cat = [col for col in df.columns if df[col].nunique() < 10 and df[col].dtypes</pre>
             cat_but_car = [col for col in df.columns if df[col].nunique() > 20 and str(df[col].dt
             cat cols=cat cols+num but cat
             cat cols= [col for col in cat cols if col not in cat but car]
             num_cols=[col for col in df.columns if df[col].dtypes in ["int","float","int64"]]
             num cols=[col for col in num cols if col not in cat cols]
             print(f"Observations: {dataframe.shape[0]}") #gözlemler
             print(f"Variables: {dataframe.shape[1]}") #değişkenler
             print(f"cat_cols: {len(cat_cols)}") #kategorik değişkenin boyutu
             print(f"cat cols: {len(num cols)}")
             print(f"cat_cols: {len(cat_but_car)}")
             print(f"cat cols: {len(num but cat)}")
             return cat_cols, num_cols,cat_but_car
                                                                                                 >
In [38]: help(grab_col_names)
         NameError
                                                   Traceback (most recent call last)
         Cell In[38], line 1
         ---> 1 help(grab col names)
         NameError: name 'grab_col_names' is not defined
```

```
In [39]:
         cat_cols = [col for col in df.columns if str(df[col].dtypes) in ["category","object","boo
         num_but_cat = [col for col in df.columns if df[col].nunique() < 10 and df[col].dtypes in</pre>
         cat_but_car = [col for col in df.columns if df[col].nunique() > 20 and str(df[col].dtypes
         cat_cols=cat_cols+num_but_cat
         cat cols= [col for col in cat cols if col not in cat but car]
                                                                                                   >
In [40]:
         num_cols=[col for col in df.columns if df[col].dtypes in ["int","float","int64"]]
         num_cols=[col for col in num_cols if col not in cat_cols]
In [42]: grab_col_names(df)
         Observations: 891
         Variables: 15
         cat cols: 13
         cat_cols: 2
         cat cols: 0
         cat_cols: 4
Out[42]: (['sex',
            'embarked',
            'class',
            'who',
            'adult male',
            'deck',
            'embark town',
            'alive',
            'alone',
            'survived',
            'pclass',
            'sibsp',
            'parch'],
           ['age', 'fare'],
           [])
In [44]: cat_cols,num_cols,cat_but_car= grab_col_names(df)
         Observations: 891
         Variables: 15
         cat cols: 13
         cat_cols: 2
         cat cols: 0
         cat_cols: 4
 In [ ]: #4 Analysis of Target Variable Hedef Değişken Analizi
```

```
In [1]: import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        pd.set_option('display.max_columns',None)
        pd.set option('display.width',500)
        df=sns.load_dataset("titanic")
        df.head()
        df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 891 entries, 0 to 890
        Data columns (total 15 columns):
                         Non-Null Count Dtype
            Column
        ---
            ----
                          -----
         0
             survived
                          891 non-null
                                         int64
         1
                          891 non-null
                                         int64
             pclass
         2
             sex
                         891 non-null
                                         object
         3
             age
                         714 non-null
                                         float64
         4
                          891 non-null
                                         int64
             sibsp
         5
             parch
                          891 non-null
                                         int64
         6
             fare
                         891 non-null
                                         float64
         7
             embarked
                         889 non-null
                                         object
         8
             class
                         891 non-null
                                         category
         9
                          891 non-null
             who
                                         object
         10 adult_male 891 non-null
                                         bool
         11 deck
                         203 non-null
                                         category
         12 embark_town 889 non-null
                                         object
         13 alive
                          891 non-null
                                         object
         14 alone
                          891 non-null
                                         bool
        dtypes: bool(2), category(2), float64(2), int64(4), object(5)
        memory usage: 80.7+ KB
In [5]: for col in df.columns:
            if df[col].dtypes =="bool":
                df[col]=df[col].astype(int)
In [6]: | def cat_summary(dataframe, col_name, plot=False):
            print(pd.DataFrame({col_name: dataframe[col_name].value_counts(),
                                "Ratio": 100*dataframe[col_name].value_counts()/len(dataframe)
                               }))
            if plot:
                sns.countplot(x=dataframe[col name],data=dataframe)
                plt.show(block=True)
```

```
In [8]: | def grab_col_names(dataframe, cat_th=10, car_th=20): #docstring
             Veri setindeki kategorik, numerik ve kategorik fakat kardinal değişkenlerin isimlerin
             #Parameters
             dataframe: dataframe
                 değişkenlerin isimleri alınmak istenen dataframe'dir.
             cat th: int,float
                 numerik fakat kategorik olan değişkenler için sınıf eşik değeri
             car_th: int, float
                  kategorik fakat kardinal değişkenler için sınıf eşik değeri
             cat cols = [col for col in df.columns if str(df[col].dtypes) in ["category", "object",
             num_but_cat = [col for col in df.columns if df[col].nunique() < 10 and df[col].dtypes</pre>
             cat_but_car = [col for col in df.columns if df[col].nunique() > 20 and str(df[col].dt)
             cat cols=cat cols+num but cat
             cat_cols= [col for col in cat_cols if col not in cat_but_car]
             num_cols=[col for col in df.columns if df[col].dtypes in ["int","float","int64"]]
             num_cols=[col for col in num_cols if col not in cat_cols]
             print(f"Observations: {dataframe.shape[0]}") #gözlemler
             print(f"Variables: {dataframe.shape[1]}") #değişkenler
             print(f"cat_cols: {len(cat_cols)}") #kategorik değişkenin boyutu
             print(f"num_cols: {len(num_cols)}")
             print(f"cat_but_car_cols: {len(cat_but_car)}")
             print(f"num_but_cat_cols: {len(num_but_cat)}")
             return cat_cols, num_cols,cat_but_car
                                                                                                  >
 In [9]: | cat_cols,num_cols,cat_but_car= grab_col_names(df)
         Observations: 891
         Variables: 15
         cat cols: 13
         num_cols: 2
         cat but car cols: 0
         num_but_cat_cols: 6
In [10]: df["survived"].value counts()
Out[10]: survived
         0
              549
               342
         1
         Name: count, dtype: int64
In [14]: | df.groupby("sex")["survived"].mean()
Out[14]: sex
         female
                    0.742038
                   0.188908
         male
         Name: survived, dtype: float64
In [18]: def target summary with cat(dataframe, target, categorical col):
             print(pd.DataFrame({"TARGET_MEAN":dataframe.groupby(categorical_col)[target].mean()})
```

```
In [20]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
pd.set_option('display.max_columns',None)
pd.set_option('display.width',500)
df=pd.read_csv("data.csv")
df.head()
df.info()
```

```
FileNotFoundError
                                          Traceback (most recent call last)
Cell In[20], line 6
      4 pd.set option('display.max columns', None)
      5 pd.set_option('display.width',500)
---> 6 df=pd.read csv("data.csv")
      7 df.head()
      8 df.info()
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\io\parsers\readers.py:912, in rea
d csv(filepath or buffer, sep, delimiter, header, names, index col, usecols, dtype, engi
ne, converters, true_values, false_values, skipinitialspace, skiprows, skipfooter, nrow
s, na values, keep default na, na filter, verbose, skip blank lines, parse dates, infer
datetime_format, keep_date_col, date_parser, date_format, dayfirst, cache_dates, iterato
r, chunksize, compression, thousands, decimal, lineterminator, quotechar, quoting, doubl
equote, escapechar, comment, encoding, encoding errors, dialect, on bad lines, delim whi
tespace, low_memory, memory_map, float_precision, storage_options, dtype_backend)
    899 kwds defaults = refine defaults read(
            dialect,
    900
    901
            delimiter,
   (\ldots)
    908
            dtype_backend=dtype_backend,
    909 )
    910 kwds.update(kwds_defaults)
--> 912 return read(filepath or buffer, kwds)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\io\parsers\readers.py:577, in re
ad(filepath_or_buffer, kwds)
    574 validate_names(kwds.get("names", None))
    576 # Create the parser.
--> 577 parser = TextFileReader(filepath_or_buffer, **kwds)
    579 if chunksize or iterator:
            return parser
    580
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\io\parsers\readers.py:1407, in Te
xtFileReader. init (self, f, engine, **kwds)
            self.options["has index names"] = kwds["has index names"]
   1406 self.handles: IOHandles | None = None
-> 1407 self._engine = self._make_engine(f, self.engine)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\io\parsers\readers.py:1661, in Te
xtFileReader. make engine(self, f, engine)
            if "b" not in mode:
   1659
                mode += "b"
   1660
-> 1661 self.handles = get handle(
   1662
            f,
   1663
            mode,
   1664
            encoding=self.options.get("encoding", None),
            compression=self.options.get("compression", None),
   1665
            memory map=self.options.get("memory map", False),
   1666
   1667
            is text=is text,
            errors=self.options.get("encoding errors", "strict"),
   1668
   1669
            storage options=self.options.get("storage options", None),
   1670 )
   1671 assert self.handles is not None
   1672 f = self.handles.handle
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\io\common.py:859, in get_handle(p
ath_or_buf, mode, encoding, compression, memory_map, is_text, errors, storage_options)
    854 elif isinstance(handle, str):
            # Check whether the filename is to be opened in binary mode.
    855
    856
            # Binary mode does not support 'encoding' and 'newline'.
```

```
if ioargs.encoding and "b" not in ioargs.mode:
              857
              858
                           # Encoding
                           handle = open(
          --> 859
              860
                               handle,
              861
                               ioargs.mode,
              862
                               encoding=ioargs.encoding,
              863
                               errors=errors,
                               newline="",
              864
              865
                           )
                       else:
              866
              867
                           # Binary mode
              868
                           handle = open(handle, ioargs.mode)
          FileNotFoundError: [Errno 2] No such file or directory: 'data.csv'
In [21]: |num_cols=[col for col in df.columns if df[col].dtype in [int,float]]
          num_cols
Out[21]: ['age', 'fare', 'adult_male', 'alone']
In [22]: |corr=df[num_cols].corr()
In [23]: sns.set(rc={'figure.figsize': (15,5)})
In [24]: sns.heatmap(corr, cmap="RdBu")
          plt.show()
                                                                                                      1.0
           age
                                                                                                     - 0.6
           fare
                                                                                                     - 0.4
           adult male
                                                                                                     - 0.2
                                                                                                     - 0.0
```

In [ ]:

adult\_male

alone

fare

age