1 Praktikum 3

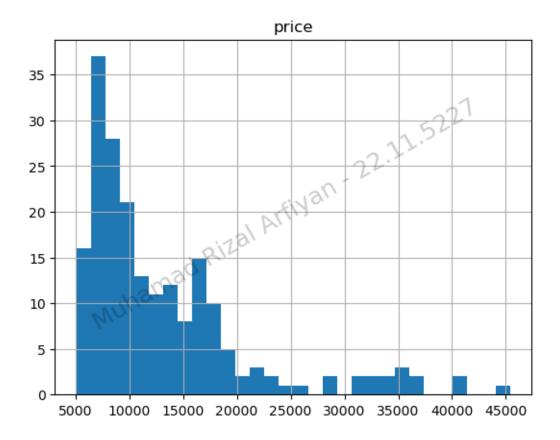
Muhamad Rizal Arfiyan - 22.11.5227 - IF11

```
https://github.com/rizalarfiyan/big-data
```

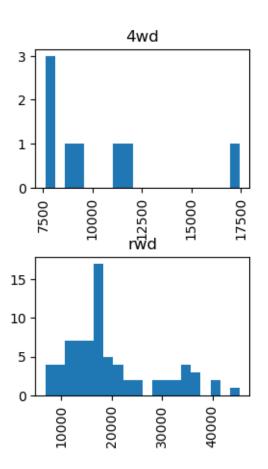
```
[1]: import matplotlib.pyplot as plt
     import numpy as np
     import pandas as pd
     import seaborn as sns
     path = "./automobileEDA.csv"
     df = pd.read_csv(path)
     df.head()
[1]:
        symboling normalized-losses
                                               make aspiration num-of-doors
                                       alfa-romero
                                  122
                                                           std
                                                                         t.wo
     1
                3
                                  122
                                       alfa-romero
                                                            std
                                                                         two
     2
                1
                                  122
                                       alfa-romero
                                                           std
                                                                         two
     3
                2
                                  164
                                               audi
                                                           std
                                                                        four
                2
     4
                                  164
                                               audi
                                                           std
                                                                        four
         body-style drive-wheels engine-location wheel-base
                                                                   length
       convertible
                              rwd
                                             front
                                                          88.6 0.811148
       convertible
                                                          88.6 0.811148
                              rwd
                                             front
     2
          hatchback
                              rwd
                                             front
                                                          94.5 0.822681
     3
              sedan
                              fwd
                                             front
                                                          99.8 0.848630
     4
                                                          99.4 0.848630
              sedan
                              4wd
                                             front
        compression-ratio
                           horsepower
                                        peak-rpm city-mpg highway-mpg
                                                                           price \
     0
                       9.0
                                 111.0
                                           5000.0
                                                        21
                                                                        13495.0
                       9.0
                                 111.0
                                           5000.0
                                                        21
                                                                     27 16500.0
     1
     2
                       9.0
                                 154.0
                                           5000.0
                                                        19
                                                                     26 16500.0
     3
                      10.0
                                 102.0
                                           5500.0
                                                        24
                                                                     30 13950.0
     4
                       8.0
                                 115.0
                                           5500.0
                                                        18
                                                                     22 17450.0
       city-L/100km horsepower-binned
                                         diesel
                                                  gas
     0
          11.190476
                                 Medium
                                               0
          11.190476
                                 Medium
                                               0
                                                    1
     1
     2
          12.368421
                                 Medium
                                               0
                                                    1
     3
           9.791667
                                 Medium
                                               0
                                                    1
                                                    1
     4
          13.055556
                                 Medium
                                               0
     [5 rows x 29 columns]
[2]: df.hist(column="price", bins=30)
     plt.text(
         0.5,
```

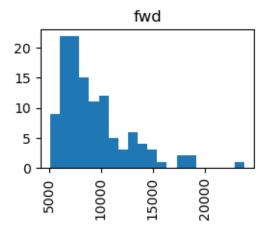
```
"Muhamad Rizal Arfiyan - 22.11.5227",
fontsize=18,
color="black",
ha="center",
va="center",
alpha=0.2,
transform=plt.gcf().transFigure,
rotation=30,
```

[2]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



```
[3]: df.hist(column="price", by="drive-wheels", bins=20);
```

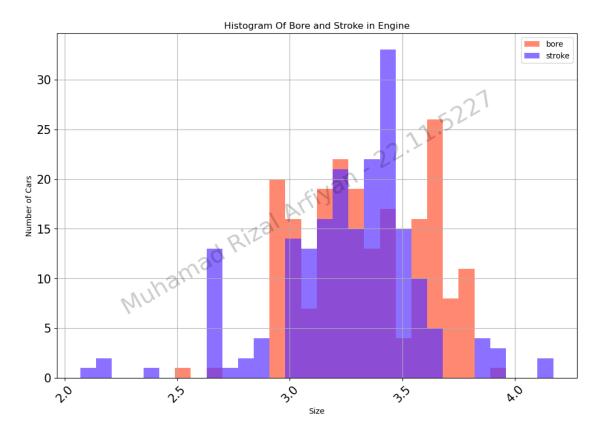




```
[4]: df[["bore", "stroke"]].plot(
         kind="hist",
         alpha=0.7,
         bins=30,
         title="Histogram Of Bore and Stroke in Engine",
         rot=45,
         grid=True,
         figsize=(12, 8),
         fontsize=15,
         color=["#FF5733", "#5C33FF"],
     )
     plt.xlabel("Size")
     plt.ylabel("Number of Cars")
     plt.text(
         0.5,
         "Muhamad Rizal Arfiyan - 22.11.5227",
         fontsize=30,
         color="black",
         ha="center",
```

```
va="center",
alpha=0.2,
transform=plt.gcf().transFigure,
rotation=30,
```

[4]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



```
[5]: from scipy import stats

pearson_coef, p_value = stats.pearsonr(df["wheel-base"], df["price"])
print(
    "The Pearson Correlation Coefficient is",
    pearson_coef,
    " with a P-value of P =",
    p_value,
)
```

The Pearson Correlation Coefficient is 0.5846418222655083 with a P-value of P = 8.076488270732552e-20

```
[6]: pearson_coef, p_value = stats.pearsonr(df["horsepower"], df["price"])
print(
     "The Pearson Correlation Coefficient is",
     pearson_coef,
     " with a P-value of P = ",
     p_value,
)
```

The Pearson Correlation Coefficient is 0.8095745670036555 with a P-value of P = 6.369057428260919e-48

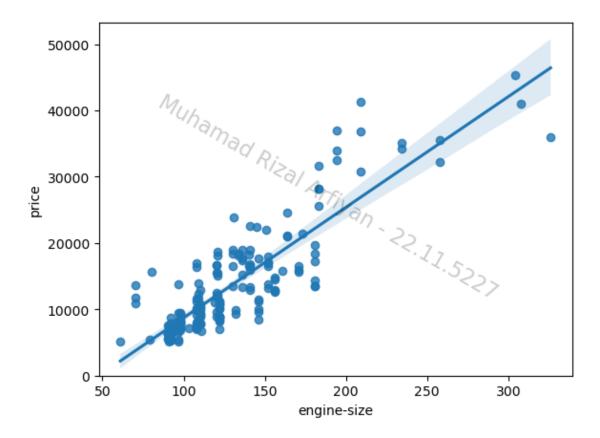
[7]: print(df.dtypes)

```
symboling
                        int64
normalized-losses
                        int64
make
                       object
aspiration
                       object
num-of-doors
                       object
body-style
                       object
drive-wheels
                       object
engine-location
                       object
wheel-base
                      float64
                      float64
length
width
                      float64
height
                      float64
                        int64
curb-weight
engine-type
                       object
num-of-cylinders
                       object
engine-size
                        int64
fuel-system
                       object
bore
                      float64
stroke
                      float64
compression-ratio
                      float64
                      float64
horsepower
peak-rpm
                      float64
                        int64
city-mpg
                        int64
highway-mpg
price
                      float64
city-L/100km
                      float64
horsepower-binned
                       object
                        int64
diesel
                        int64
gas
dtype: object
```

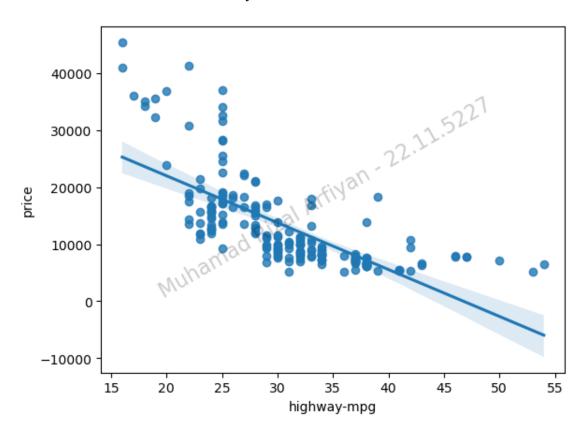
[8]: sns.regplot(x="engine-size", y="price", data=df)

```
plt.text(
    0.5,
    0.5,
    "Muhamad Rizal Arfiyan - 22.11.5227",
    fontsize=16,
    color="black",
    ha="center",
    va="center",
    alpha=0.2,
    transform=plt.gcf().transFigure,
    rotation=-30,
)
```

[8]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



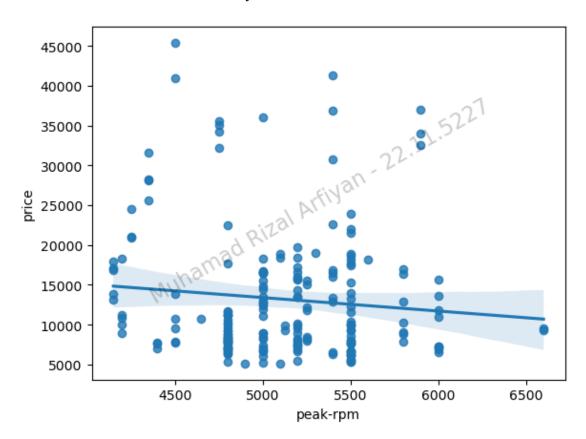
[10]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



```
[11]: df[["highway-mpg", "price"]].corr()

[11]: highway-mpg price
    highway-mpg 1.000000 -0.704692
    price -0.704692 1.000000
```

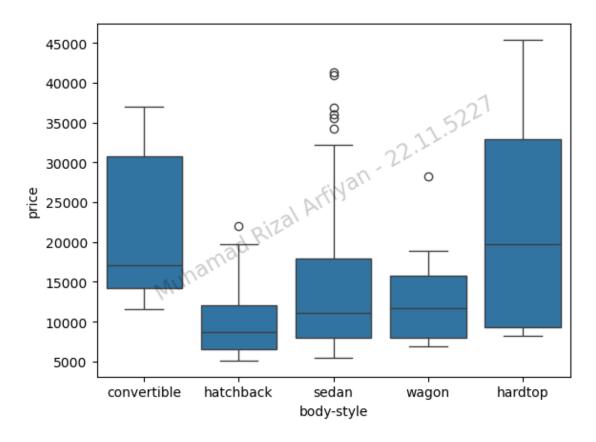
[12]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



```
[13]: df[["peak-rpm", "price"]].corr()
```

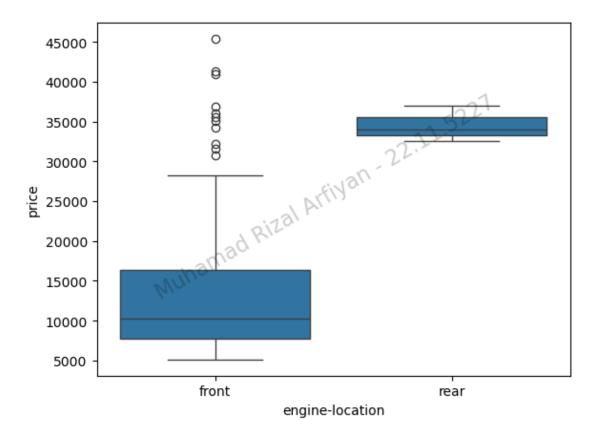
[13]: peak-rpm price peak-rpm 1.000000 -0.101616 price -0.101616 1.000000

[14]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



```
fontsize=16,
  color="black",
  ha="center",
  va="center",
  alpha=0.2,
  transform=plt.gcf().transFigure,
  rotation=30,
)
```

[15]: Text(0.5, 0.5, 'Muhamad Rizal Arfiyan - 22.11.5227')



[16]:	<pre>df.describe()</pre>						
[16]:		symboling	normalized-losses	wheel-base	length	width	\
	count	201.000000	201.00000	201.000000	201.000000	201.000000	
	mean	0.840796	122.00000	98.797015	0.837102	0.915126	
	std	1.254802	31.99625	6.066366	0.059213	0.029187	
	min	-2.000000	65.00000	86.600000	0.678039	0.837500	
	25%	0.000000	101.00000	94.500000	0.801538	0.890278	
	50%	1.000000	122.00000	97.000000	0.832292	0.909722	
	75%	2.000000	137.00000	102.400000	0.881788	0.925000	

```
3.000000
                                   256.00000
                                              120.900000
                                                              1.000000
                                                                           1.000000
      max
                  height
                           curb-weight
                                         engine-size
                                                             bore
                                                                        stroke
              201.000000
                            201.000000
                                          201.000000
                                                       201.000000
                                                                    197.000000
      count
               53.766667
                           2555.666667
                                          126.875622
                                                         3.330692
                                                                      3.256904
      mean
      std
                2.447822
                            517.296727
                                           41.546834
                                                         0.268072
                                                                      0.319256
                           1488.000000
      min
               47.800000
                                           61.000000
                                                         2.540000
                                                                      2.070000
      25%
               52.000000
                          2169.000000
                                           98.000000
                                                         3.150000
                                                                      3.110000
      50%
                           2414.000000
               54.100000
                                          120.000000
                                                         3.310000
                                                                      3.290000
      75%
               55.500000
                           2926.000000
                                          141.000000
                                                         3.580000
                                                                      3.410000
               59.800000
                           4066.000000
      max
                                          326.000000
                                                         3.940000
                                                                      4.170000
              compression-ratio
                                  horsepower
                                                  peak-rpm
                                                               city-mpg
                                                                          highway-mpg
                     201.000000
                                  201.000000
                                                201.000000
                                                             201.000000
                                                                           201.000000
      count
                      10.164279
                                  103.405534
                                               5117.665368
                                                              25.179104
                                                                            30.686567
      mean
      std
                       4.004965
                                   37.365700
                                                478.113805
                                                                6.423220
                                                                             6.815150
                       7.000000
                                   48.000000
                                               4150.000000
                                                               13.000000
                                                                             16.000000
      min
      25%
                       8.600000
                                   70.000000
                                               4800.000000
                                                               19.000000
                                                                            25.000000
      50%
                       9.000000
                                   95.000000
                                               5125.369458
                                                               24.000000
                                                                            30.000000
      75%
                       9.400000
                                  116.000000
                                               5500.000000
                                                               30.000000
                                                                            34.000000
                      23.000000
                                  262.000000
                                               6600.000000
                                                               49.000000
                                                                            54.000000
      max
                             city-L/100km
                                                diesel
                     price
                                                                 gas
                201.000000
                               201.000000
      count
                                            201.000000
                                                         201.000000
              13207.129353
                                              0.099502
                                                           0.900498
      mean
                                 9.944145
      std
               7947.066342
                                 2.534599
                                              0.300083
                                                           0.300083
                                 4.795918
      min
               5118.000000
                                              0.000000
                                                           0.000000
      25%
               7775.000000
                                 7.833333
                                              0.000000
                                                           1.000000
                                                           1.000000
      50%
              10295.000000
                                 9.791667
                                              0.000000
      75%
                                                           1.000000
              16500.000000
                                12.368421
                                              0.00000
              45400.000000
                                                           1.000000
      max
                                18.076923
                                              1.000000
      df.describe(include=["object"])
[17]:
                 make aspiration num-of-doors body-style drive-wheels
      count
                  201
                              201
                                            201
                                                        201
                                                                      201
                   22
                                2
                                              2
                                                          5
                                                                        3
      unique
      top
               toyota
                              std
                                           four
                                                      sedan
                                                                      fwd
      freq
                   32
                              165
                                            115
                                                         94
                                                                      118
              engine-location engine-type num-of-cylinders fuel-system
                           201
                                        201
      count
                                                          201
                                                                       201
                                                                         8
                                          6
                                                            7
      unique
      top
                        front
                                        ohc
                                                         four
                                                                      mpfi
                                                                        92
      freq
                           198
                                        145
                                                          157
```

horsepower-binned

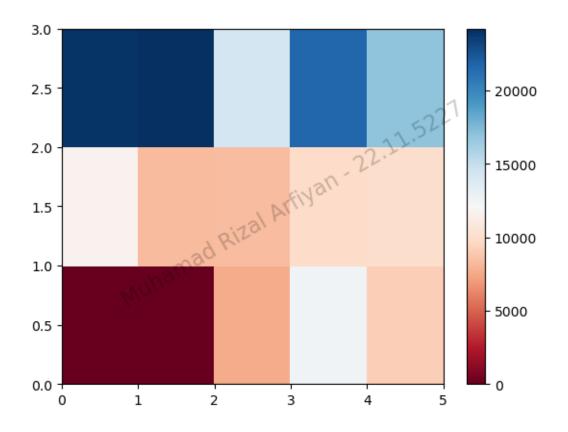
```
200
      count
                              3
      unique
      top
                           Low
      freq
                           115
[18]: df["drive-wheels"].value_counts()
[18]: drive-wheels
      fwd
             118
      rwd
              75
               8
      4wd
      Name: count, dtype: int64
[19]: df["drive-wheels"].value_counts().to_frame()
Γ197:
                    count
      drive-wheels
      fwd
                      118
      rwd
                       75
      4wd
                        8
[20]: drive_wheels_counts = df["drive-wheels"].value_counts().to_frame()
      drive_wheels_counts.rename(columns={"drivewheels": "value_counts"},__
       →inplace=True)
      drive_wheels_counts
[20]:
                    count
      drive-wheels
      fwd
                      118
      rwd
                       75
      4wd
                        8
[21]: drive_wheels_counts.index.name = "drive-wheels"
      drive_wheels_counts
[21]:
                    count
      drive-wheels
      fwd
                      118
      rwd
                       75
      4wd
                        8
[22]: # engine-location as variable
      engine_loc_counts = df["engine-location"].value_counts().to_frame()
      engine_loc_counts.rename(columns={"enginelocation": "value_counts"},__
       ⇔inplace=True)
      engine_loc_counts.index.name = "engine-location"
      engine_loc_counts.head(10)
```

```
[22]:
                       count
      engine-location
                         198
      front
      rear
                           3
[23]: df["drive-wheels"].unique()
[23]: array(['rwd', 'fwd', '4wd'], dtype=object)
[24]: df_group_one = df[["drive-wheels", "price"]]
      df_group_one
[24]:
          drive-wheels
                          price
      0
                   rwd
                        13495.0
                        16500.0
      1
                   rwd
      2
                   rwd
                        16500.0
      3
                   fwd
                        13950.0
      4
                        17450.0
                   4wd
      . .
                   rwd 16845.0
      196
      197
                        19045.0
                   rwd
      198
                        21485.0
                   rwd
      199
                        22470.0
                   rwd
      200
                   rwd
                        22625.0
      [201 rows x 2 columns]
[25]: df_group_one = df_group_one.groupby(["drive-wheels"], as_index=False).mean()
      df_group_one
[25]:
       drive-wheels
                             price
                 4wd 10241.000000
                       9244.779661
      1
                 fwd
                 rwd 19757.613333
[26]: # grouping results
      df_gptest = df[["drive-wheels", "body-style", "price"]]
      grouped_test1 = df_gptest.groupby(["drive-wheels", "body-style"],__
       ⇔as_index=False).mean()
      grouped_test1
[26]:
         drive-wheels
                        body-style
                                           price
                                     7603.000000
      0
                  4wd
                         hatchback
      1
                  4wd
                             sedan
                                   12647.333333
      2
                  4wd
                             wagon
                                     9095.750000
      3
                  fwd
                      convertible
                                   11595.000000
      4
                  fwd
                           hardtop
                                     8249.000000
```

```
5
                  fwd
                         hatchback
                                     8396.387755
      6
                                     9811.800000
                  fwd
                             sedan
      7
                  fwd
                             wagon
                                     9997.333333
      8
                       convertible 23949.600000
                  rwd
      9
                  rwd
                           hardtop
                                    24202.714286
                         hatchback
      10
                  rwd
                                    14337.777778
      11
                             sedan 21711.833333
                  rwd
      12
                                    16994.222222
                  rwd
                             wagon
[27]: grouped_pivot = grouped_test1.pivot(index="drive-wheels", columns="body-style")
      grouped_pivot
[27]:
                         price
                                      hardtop
      body-style
                   convertible
                                                  hatchback
                                                                     sedan
      drive-wheels
      4wd
                                                7603.000000 12647.333333
                           NaN
                                          {\tt NaN}
      fwd
                                 8249.000000
                       11595.0
                                                8396.387755
                                                              9811.800000
                       23949.6 24202.714286 14337.777778 21711.833333
      rwd
      body-style
                           wagon
      drive-wheels
      4wd
                     9095.750000
      fwd
                     9997.333333
                    16994.222222
      rwd
[28]: grouped_pivot = grouped_pivot.fillna(0)
      grouped_pivot
[28]:
                                                                            \
                         price
      body-style
                   convertible
                                     hardtop
                                                  hatchback
                                                                     sedan
      drive-wheels
      4wd
                           0.0
                                     0.000000
                                                7603.000000 12647.333333
      fwd
                       11595.0
                                 8249.000000
                                                8396.387755
                                                              9811.800000
                       23949.6 24202.714286 14337.777778 21711.833333
      rwd
      body-style
                           wagon
      drive-wheels
      4wd
                     9095.750000
                     9997.333333
      fwd
      rwd
                    16994.222222
[29]: # Write your code below and press Shift+Enter to execute
      df_gptest2 = df[["body-style", "price"]]
      grouped_test_bodystyle = df_gptest2.groupby(["body-style"], as_index=False).
       ⊶mean()
```

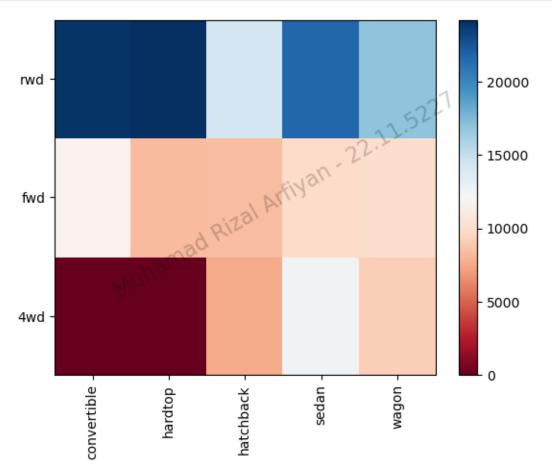
grouped_test_bodystyle

```
[29]:
         body-style
                             price
     0 convertible 21890.500000
      1
            hardtop 22208.500000
          hatchback 9957.441176
      2
      3
               sedan 14459.755319
      4
               wagon 12371.960000
[30]: import matplotlib.pyplot as plt
      # use the grouped results
      plt.pcolor(grouped_pivot, cmap="RdBu")
      plt.colorbar()
      plt.text(
         0.5,
          0.5,
          "Muhamad Rizal Arfiyan - 22.11.5227",
         fontsize=16,
          color="black",
          ha="center",
          va="center",
          alpha=0.2,
          transform=plt.gcf().transFigure,
          rotation=30,
     plt.show()
```



```
[31]: fig, ax = plt.subplots()
      im = ax.pcolor(grouped_pivot, cmap="RdBu")
      # label names
      row_labels = grouped_pivot.columns.levels[1]
      col_labels = grouped_pivot.index
      # move ticks and labels to the center
      ax.set_xticks(np.arange(grouped_pivot.shape[1]) + 0.5, minor=False)
      ax.set_yticks(np.arange(grouped_pivot.shape[0]) + 0.5, minor=False)
      # insert labels
      ax.set_xticklabels(row_labels, minor=False)
      ax.set_yticklabels(col_labels, minor=False)
      # rotate label if too long
      plt.xticks(rotation=90)
      fig.colorbar(im)
      plt.text(
          0.5,
          0.5,
          "Muhamad Rizal Arfiyan - 22.11.5227",
          fontsize=16,
          color="black",
          ha="center",
          va="center",
```

```
alpha=0.2,
  transform=plt.gcf().transFigure,
  rotation=30,
)
plt.show()
```



```
[32]: grouped_test2 = df_gptest[["drive-wheels", "price"]].groupby(["drive-wheels"]) grouped_test2.head(2)
```

```
[32]:
          drive-wheels
                           price
                         13495.0
      0
                    rwd
      1
                    rwd
                         16500.0
      3
                         13950.0
                    fwd
      4
                    4wd
                         17450.0
      5
                         15250.0
                    fwd
      136
                          7603.0
                    4wd
```

```
[33]: df_gptest
```

```
[33]:
         drive-wheels
                        body-style
                                      price
                       convertible 13495.0
     0
                  rwd
     1
                  rwd
                       convertible 16500.0
     2
                  rwd
                         hatchback 16500.0
     3
                             sedan 13950.0
                  fwd
     4
                             sedan 17450.0
                  4wd
     196
                  rwd
                             sedan 16845.0
     197
                             sedan 19045.0
                  rwd
     198
                  rwd
                             sedan 21485.0
     199
                             sedan 22470.0
                  rwd
     200
                             sedan 22625.0
                  rwd
     [201 rows x 3 columns]
[34]: grouped_test2.get_group("4wd")["price"]
[34]: 4
            17450.0
     136
             7603.0
     140
             9233.0
     141
            11259.0
     144
             8013.0
     145
            11694.0
     150
             7898.0
     151
             8778.0
     Name: price, dtype: float64
[35]: f_val, p_val = stats.f_oneway(
         grouped_test2.get_group("fwd")["price"],
         grouped_test2.get_group("rwd")["price"],
         grouped_test2.get_group("4wd")["price"],
     print("ANOVA results: F=", f_val, ", P =", p_val)
     ANOVA results: F = 67.95406500780399, P = 3.3945443577151245e-23
[36]: f_val, p_val = stats.f_oneway(
         grouped_test2.get_group("fwd")["price"], grouped_test2.
      print("ANOVA results: F=", f_val, ", P =", p_val)
     ANOVA results: F= 130.5533160959111 , P = 2.2355306355677845e-23
[39]: f_val, p_val = stats.f_oneway(
         grouped_test2.get_group("4wd")["price"], grouped_test2.

¬get_group("fwd")["price"]
```

```
print("ANOVA results: F=", f_val, ", P =", p_val)
```

ANOVA results: F= 0.665465750252303 , P= 0.41620116697845666