```
CS3-mid-p5
       5-1. (i)
       5-2. (i), (iii)
       5-3. (A) (i) (B) (iii) (C) (iv) (D) (ii)
       5-4.
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import statsmodels.api as sm
        C:\Users\shugo\Miniconda3\envs\cs3-2022\lib\site-packages\statsmodels\compat\panda
        s.py:61: FutureWarning: pandas.Int64Index is deprecated and will be removed from p
        andas in a future version. Use pandas. Index with the appropriate dtype instead.
         from pandas import Int64Index as NumericIndex
In [2]:
         csv_in = 'mid-p5.csv'
         df = pd.read_csv(csv_in, sep=',', skiprows=0, header=0)
         print(df.shape)
         print(df.info())
         display(df.head())
        (100, 2)
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 100 entries, 0 to 99
        Data columns (total 2 columns):
         # Column Non-Null Count Dtype
            -----
                    100 non-null float64
         0
           Χ
         1 y
                    100 non-null float64
        dtypes: float64(2)
        memory usage: 1.7 KB
        None
                  У
        0 3.89 15.94
        1 1.19 3.52
        2 4.12 20.55
        3 4.83 30.98
        4 4.86 21.34
In [3]:
         X = df[['x']]
         y = df['y']
         print('X:', X.shape)
         display(X.head())
         print('y:', y.shape)
         print(y.head())
        X: (100, 1)
             Х
        0 3.89
```

```
1 1.19
      2 4.12
      3 4.83
      y: (100,)
        15.94
      1
         3.52
      2 20.55
         30.98
      3
         21.34
      4
      Name: y, dtype: float64
In [4]:
      X c = sm.add constant(X)
       display(X.head())
       display(X_c.head())
          Х
      0 3.89
      1 1.19
      2 4.12
      3 4.83
      4 4.86
        const x
         1.0 3.89
         1.0 1.19
      2
         1.0 4.12
         1.0 4.83
         1.0 4.86
In [5]:
      model = sm.OLS(y, X_c)
       results = model.fit()
       print(results.summary())
                          OLS Regression Results
      ______
      Dep. Variable:
                                 y R-squared:
                                                       0.613
      Model:
                               OLS Adj. R-squared:
                                                           0.609
      Method:
                        Least Squares F-statistic:
                                                            155.4
                                   Prob (F-statistic):
      Date:
                     Fri, 03 Jun 2022
                                                         6.22e-22
      Time:
                            17:43:03
                                    Log-Likelihood:
                                                           -320.47
      No. Observations:
                               100
                                    AIC:
                                                             644.9
      Df Residuals:
                                98
                                   BIC:
                                                             650.2
      Df Model:
                                1
      Covariance Type: nonrobust
      ______
                  coef std err t P>|t| [0.025 0.975]
                1.8024 1.118 1.613 0.110 -0.416
      const
                                                           4.021
```

X

X	4.9406	0.396	12.	465	0.000	4.154	5.727
==========	========		====	=======			=====
Omnibus:	: 0.419		9	Durbin-Watson:			2.058
<pre>Prob(Omnibus):</pre>		0.811		Jarque-Bera (JB):			0.104
Skew:		0.008	3	Prob(JB):	:		0.949
Kurtosis:		3.157	7	Cond. No.			5.72

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [6]:
    print('R2:', results.rsquared)
    print('Adj R2:', results.rsquared_adj)
```

R2: 0.6132358565364777 Adj R2: 0.6092892836439928

```
In [7]: print(results.params)
```

const 1.802399 x 4.940600 dtype: float64

5-3. a0 1.802

5-3. a1 4.941

```
In [8]:
    a0 = results.params['const']
    a1 = results.params['x']
    x_min = X['x'].min()
    x_max = X['x'].max()
    x_min_max = np.array([x_min, x_max])
    y_min_max = a0 + a1 * x_min_max
```

```
In [9]:
    plt.title('y = ({:.3f}) + ({:.3f}) * x'.format(a0, a1))
    plt.scatter(X['x'], y)
    plt.plot(x_min_max, y_min_max, c='red')
    plt.xlabel('x')
    plt.ylabel('y')
    plt.show()
```

