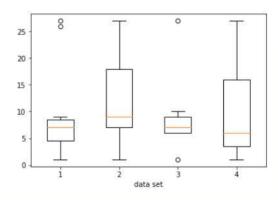
1. Show the execution result of the following code (1) \sim (4), (6) \sim (10) and the order of the list data (A,B,C,D) for the given boxplot graph (5). (40 points)

```
(1) import pandas as pd
    df = pd.DataFrame(np.ones((3,4))).astype(int)
    df.iloc[:2, 1] = 0
    df.iloc[:1, 2] = 0
    df.iloc[0,0] = 0
    print(df)
```

```
(5) import matplotlib.pyplot as plt
    A = [1,2,3,4,5,6,6,7,25,26,27]
    B = [1,6,6,6,7,7,8,9, 9,10,27]
    C = [1,5,7,7,8,9,13,15,21,25,27]
    D = [1,4,4,5,5,7,7,8,9,26,27]

    box_plot_data=[?,?,?,?] # order ?
    plt.boxplot(box_plot_data)
    plt.xlabel('data set')
    plt.show()
    print(box_plot_data)
```



```
(6) text = 'Write a Python code that accept some words and explore on this data'
    with open("words.txt", "w") asf:
        f.write(text)
    with open("words.txt", "r") asf:
        data = f.read()
    print(len(data.split(" ")))
(7) class myClass:
       def __init__(self, a: list):
           self.a = np.array(a)
       def get_result(self):
           return(np.argmax(self.a) * max(self.a))
   class_1 = myClass([1, 4, 2, 3])
   class_2 = myClass([5, 6, 9, 8])
   a = class_1.get_result()
   b = class_2.get_result()
   print(a + b)
(8) from numpy import nan as NA
   df = pd.DataFrame([[NA, 6.5, 3.],
   [NA, NA, NA],
   [NA, 6.5, 2.]])
   print(df)
              1 2
           0
       0 NaN 6.5 3.0
       1 NaN NaN NaN
       2 NaN 6.5 2.0
   print(df.fillna({0: 0.5, 1: -1, 2:df[2].mean()}))
(9) import pandas as pd
   df1 = pd.DataFrame([[1,2],[3,4]], index=['a','b'], columns=['A','B'])
   df2 = pd.DataFrame([[3,3],[7,7]], index=['a','b'], columns=['X','Y'])
   new_df = pd.concat([df1, df2], axis=1)
   print(new_df)
(10) import numpy as np
    N = 100000
    ev0, ev1 = 0, 0
    fori in range(N):
        if np.random.randint(0,2) == 0:
           ev0 +=1
        else:
            ev1 +=1
    print(round(ev0/N,1), round(ev1/N,1))
```

2. We have discussed the following Python code in a class studying Web scraping. Describe how you modified (or changed) the code to do Lab assignment #2. You don't have to provide the exact Python code, but you must clearly show the modified program flow using Python syntax or pseudo code. (20 points)

```
url = 'https://kr.indeed.com/jobs?q=data+science&l=%EC%84%9C%EC%9A%B8%ED%8A%B9%EB%B3%
    84%EC%8B%9C'
link = requests.get(url)
soup = BeautifulSoup(link.text, 'html.parser')
job_elems = soup.select('.resultContent') # class

for i in job_elems:
    title = i.find('h2')
    company = i.find('span', class_='companyName')
    location = i.find('div', class_='companyLocation')

    if None in(title, company, location):
        continue

    print(title.text.strip())
    print(company.text.strip())
    print(location.text.strip())
```

3. Answer the questions by referring to the information given below (Python code, graphs). (20 points)

```
def show plot(model, X, i):
    df = pd.DataFrame(X)
    coef = pd.Series(model.coef_, df.columns)
    plt.subplot(1,3,i)
    coef.plot(kind='bar', title='model coefficients')
X, y = make_regression(n_samples=100, n_features=10, noise=30, random_state=1)
X = StandardScaler().fit transform(X)
clf1, clf2, clf3 = (__
clfs = [clf1, clf2, clf3]
plt.figure(figsize=(12,4))
for i, clf in enumerate(clfs, 1):
    clf.fit(X, y)
    show_plot(clf, X, i)
           model coefficients
                                   model coefficients
                                                            model coefficients
     50
     40
                                                     30
     30
     20
     10
```

(1) Explain what a **regularizer** is and briefly describe Ridge (L2) and Lasso (L1) regularizers using their respective (regression) loss functions.

(2) The graphs above shows the weights (or coefficients) for features as a result of training three linear regression models: LinearRegression(), Ridge(alpha=100), and Lasso(alpha=30). Explain which one of them corresponds to each model along with your reasoning.

4. Explain concepts of the Gradient Descent algorithm using a graph of the loss function and a mathematical expression of the parameter updates. Use a model of a simple linear case with only one feature (x) and the target (y) as a training data set for both regression and classification. Assume the parameters of the model are w and b. (20 points)

	linear regression	logistic regression classifier
model diagram (with input, output, parameters)		
hypothesis: y_pred =		
loss function: Loss(w,b) =		
parameter update: w = b =		(hint) $\sigma(z) = 1/(1 + \exp(-z))$ $dLoss(w,b)/dz = \sigma(z) - y$