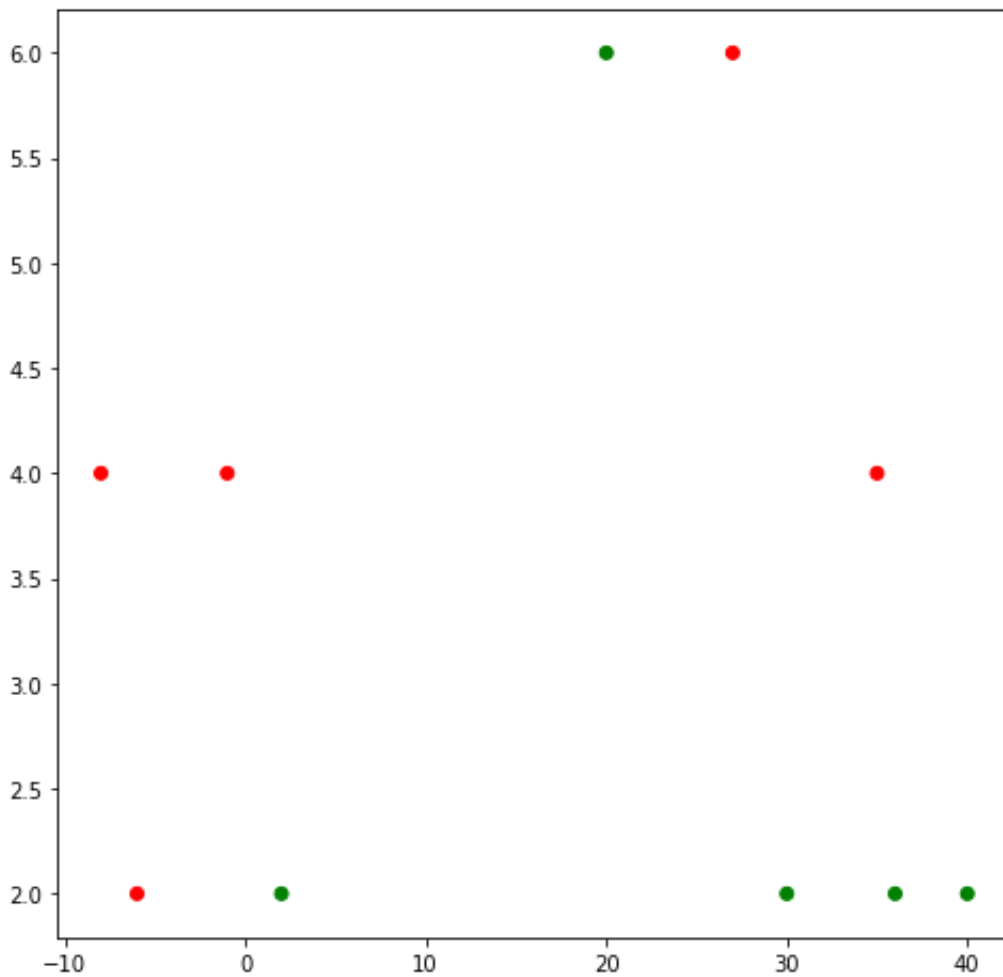


Q6

```
1 import matplotlib
2 import matplotlib.pyplot as plt
3 import numpy as np
4
5 ids = [1,2,3,4,5,6,7,8,9,10]
6 x = [27,-6,2,36,-8,40,35,30,20,-1]
7 y = [6,2,2,2,4,2,4,2,6,4]
8 label = [-1,-1,1,1,-1,1,-1,1,1,-1]
9 colors = ['red', 'green']
10
11 fig = plt.figure(figsize=(8,8))
12 plt.scatter(x, y, c=label, cmap=matplotlib.colors.ListedColormap(colors))
13
14 # cb = plt.colorbar()
15 # loc = np.arange(0,max(label),max(label)/float(len(colors)))
16 # cb.set_ticks(loc)
17 # cb.set_ticklabels(colors)
```

```
1 | <matplotlib.collections.PathCollection at 0x11b259128>
```



```
1 import math
```

```
1  
2 ids = [1,2,3,4,5,6,7,8,9,10]  
3 x = [27,-6,2,36,-8,40,35,30,20,-1]  
4 y = [6,2,2,2,4,2,4,2,6,4]  
5 label = [-1,-1,1,1,-1,1,-1,1,1,-1]  
6 dataset = list(zip(ids,x,y,label))
```

```
1 dataset
```

```

1  [(1, 27, 6, -1),
2   (2, -6, 2, -1),
3   (3, 2, 2, 1),
4   (4, 36, 2, 1),
5   (5, -8, 4, -1),
6   (6, 40, 2, 1),
7   (7, 35, 4, -1),
8   (8, 30, 2, 1),
9   (9, 20, 6, 1),
10  (10, -1, 4, -1)]

```

```

1  #Helper Method to find euclidian distance between two datapoints
2  def find_euclidian_distance(x,y):
3      sum = 0
4      sum += (x[1] - y[1]) ** 2
5      sum += (x[2] - y[2]) ** 2
6      return math.sqrt(sum)

```

```

1  find_euclidian_distance(dataset[0],dataset[1])

```

```

1  33.24154027718932

```

Q6(A)

3 Nearest datapoints for data point ID : 5

```

1  dist = []
2  for i in range(0,10):
3      dist.append((i+1,find_euclidian_distance(dataset[4],dataset[i])))
4  # dist
5  sorted(dist, key=lambda x: x[1])

```

```

1 [(5, 0.0),
2  (2, 2.8284271247461903),
3  (10, 7.0),
4  (3, 10.198039027185569),
5  (9, 28.071337695236398),
6  (1, 35.05709628591621),
7  (8, 38.05259518088089),
8  (7, 43.0),
9  (4, 44.04543109109048),
10 (6, 48.041648597857254)]

```

```

1 ### ANSWER: Therefore 3 neareset negihbors for datapoint ID: 5 are {2,10,3}

```

```

1 dist = []
2 for i in range(0,10):
3     dist.append((i+1,find_euclidian_distance(dataset[9],dataset[i])))
4
5 sorted(dist, key=lambda x: x[1])

```

```

1 [(10, 0.0),
2  (3, 3.605551275463989),
3  (2, 5.385164807134504),
4  (5, 7.0),
5  (9, 21.095023109728988),
6  (1, 28.071337695236398),
7  (8, 31.064449134018133),
8  (7, 36.0),
9  (4, 37.05401462729781),
10 (6, 41.048751503547585)]

```

Answer: Therefore 3 neareset negihbors for datapoint ID: 10 are {3,2,5}

Q6 (B)

```

1 # Experiment

```

```

2  from sklearn.neighbors import NearestNeighbors
3  import numpy as np
4  list_tuples = [(27, 6),
5                 (-6, 2),
6                 (2, 2),
7                 (36, 2),
8                 (-8, 4),
9                 (40, 2),
10                 (35, 4),
11                 (30, 2),
12                 (20, 6),
13                 (-1, 4)]
14  labels = np.array([-1,-1,1,1,-1,1,-1,1,1,-1])
15  X = np.array(list_tuples)
16  # X = np.array([[ -1, -1], [-2, -1], [-3, -2], [1, 1], [2, 1], [3, 2]])
17  # np.concatenate((X[:i],X[(i+1):]))

```

```

1  #Leave one out cross validation
2  total_errors = 0
3  for i in range(0,10):
4      nbrs = NearestNeighbors(n_neighbors=1, algorithm='auto',
5                             metric='euclidean').fit(np.concatenate((X[:i],X[(i+1):])))
6      temp_labels = np.concatenate((labels[:i],labels[(i+1):]))
7      distances, indices = nbrs.kneighbors([X[i]])
8      for nearest_index in indices:
9          # print(f'Nearest Index: {nearest_index[0]}')
10         print(f'Testing Point: {dataset[i]} -> Nearest Point:
11               {dataset[nearest_index[0]]}')
12         if temp_labels[nearest_index[0]] != dataset[i][3]:
13             total_errors += 1
14 print(f'Leave-one-out-cross-validation error: {total_errors/10}')

```

```

1 Testing Point: (1, 27, 6, -1) -> Nearest Point: (7, 35, 4, -1)
2 Testing Point: (2, -6, 2, -1) -> Nearest Point: (4, 36, 2, 1)
3 Testing Point: (3, 2, 2, 1) -> Nearest Point: (9, 20, 6, 1)
4 Testing Point: (4, 36, 2, 1) -> Nearest Point: (6, 40, 2, 1)
5 Testing Point: (5, -8, 4, -1) -> Nearest Point: (2, -6, 2, -1)
6 Testing Point: (6, 40, 2, 1) -> Nearest Point: (4, 36, 2, 1)
7 Testing Point: (7, 35, 4, -1) -> Nearest Point: (4, 36, 2, 1)
8 Testing Point: (8, 30, 2, 1) -> Nearest Point: (1, 27, 6, -1)
9 Testing Point: (9, 20, 6, 1) -> Nearest Point: (1, 27, 6, -1)
10 Testing Point: (10, -1, 4, -1) -> Nearest Point: (3, 2, 2, 1)
11 Leave-one-out-cross-validation error: 0.7

```

Leave-one-out-cross-validation error: 0.7

Q6 (C)

```

1 # Create Folds:
2 def createfolds(num_folds):
3     folds = {i: {'train': [], 'test': []} for i in range(1, num_folds+1)}
4     for i in range(1, num_folds+1):
5         for j in range(1, 11):
6             if j % 5 == i-1:
7                 folds[i]['test'].append(j)
8             else:
9                 folds[i]['train'].append(j)
10    return folds

```

```

1 folds = createfolds(5)
2 folds
3

```

```

1 {1: {'test': [5, 10], 'train': [1, 2, 3, 4, 6, 7, 8, 9]},
2  2: {'test': [1, 6], 'train': [2, 3, 4, 5, 7, 8, 9, 10]},
3  3: {'test': [2, 7], 'train': [1, 3, 4, 5, 6, 8, 9, 10]},
4  4: {'test': [3, 8], 'train': [1, 2, 4, 5, 6, 7, 9, 10]},
5  5: {'test': [4, 9], 'train': [1, 2, 3, 5, 6, 7, 8, 10]}}

```

```

1 total_errors = 0

```

```

2  for i in range(1,6):
3      test_data = []
4      test_labels = []
5      train_data = []
6      train_labels = []
7
8      for index in folds[i]['train']:
9          # print(f'Index: {index}')
10         train_data.append(list_tuples[index-1])
11         train_labels.append(labels[index-1])
12
13     train_data_narray = np.array(train_data)
14     print(f'Train Data: {train_data_narray}')
15     # print(f'{train_labels}')
16
17     for index in folds[i]['test']:
18         # print(f'Index: {index}')
19         test_data.append(list_tuples[index-1])
20         test_labels.append(labels[index-1])
21
22     test_data_narray = np.array(test_data)
23     # print(f'Test Data: {test_data_narray}')
24
25     nbrs = NearestNeighbors(n_neighbors=3, algorithm='auto',
metric='euclidean').fit(train_data_narray)
26
27     distances, indices = nbrs.kneighbors(test_data_narray)
28
29     # print(f'indices: {indices}\n')
30
31     k = 0
32     for nearest_indices in indices:
33         label_total = 0
34         for near_indice in nearest_indices:
35             # print(f'near_indice: {near_indice}')
36             label_total += train_labels[near_indice]
37
38         if label_total > 0:
39             predicted_label = 1
40         if label_total < 0:
41             predicted_label = -1
42
43         # print(f'predicted_label: {predicted_label}')
44         if predicted_label != test_labels[k]:
45             total_errors += 1
46

```

```
47         k += 1
48
49     print(f'5-fold-cross-validation error: {total_errors/10}')
```

```
1  Train Data: [[27  6]
2  [-6  2]
3  [ 2  2]
4  [36  2]
5  [40  2]
6  [35  4]
7  [30  2]
8  [20  6]]
9  indices: [[1 2 7]
10 [2 1 7]]
11
12 near_indice: 1
13 near_indice: 2
14 near_indice: 7
15 predicted_label: 1
16 near_indice: 2
17 near_indice: 1
18 near_indice: 7
19 predicted_label: 1
20 Train Data: [[-6  2]
21 [ 2  2]
22 [36  2]
23 [-8  4]
24 [35  4]
25 [30  2]
26 [20  6]
27 [-1  4]]
28 indices: [[5 6 4]
29 [2 4 5]]
30
31 near_indice: 5
32 near_indice: 6
33 near_indice: 4
34 predicted_label: 1
35 near_indice: 2
36 near_indice: 4
37 near_indice: 5
38 predicted_label: 1
39 Train Data: [[27  6]
40 [ 2  2]
41 [36  2]
```



```
42     [-8  4]
43     [40  2]
44     [30  2]
45     [20  6]
46     [-1  4]]
47 indices: [[3 7 1]
48           [2 5 4]]
49
50 near_indice: 3
51 near_indice: 7
52 near_indice: 1
53 predicted_label: -1
54 near_indice: 2
55 near_indice: 5
56 near_indice: 4
57 predicted_label: 1
58 Train Data: [[27  6]
59              [-6  2]
60              [36  2]
61              [-8  4]
62              [40  2]
63              [35  4]
64              [20  6]
65              [-1  4]]
66 indices: [[7 1 3]
67           [0 5 2]]
68
69 near_indice: 7
70 near_indice: 1
71 near_indice: 3
72 predicted_label: -1
73 near_indice: 0
74 near_indice: 5
75 near_indice: 2
76 predicted_label: -1
77 Train Data: [[27  6]
78              [-6  2]
79              [ 2  2]
80              [-8  4]
81              [40  2]
82              [35  4]
83              [30  2]
84              [-1  4]]
85 indices: [[5 4 6]
86           [0 6 5]]
87
```

```
88 near_indice: 5
89 near_indice: 4
90 near_indice: 6
91 predicted_label: 1
92 near_indice: 0
93 near_indice: 6
94 near_indice: 5
95 predicted_label: -1
96 5-fold-cross-validation error: 0.7
```

5-fold cross-validation error: 0.7

Q6 (D)

Based on the results of (B) and (C) we can not determine

1. Data points are very less.
2. For 5-fold cross-validation: testing set is size of 2 and training set size is of 8. Which is almost same as for leave-one-out (where ratio for testing is 1 and training 9).
3. So, We can not determine base