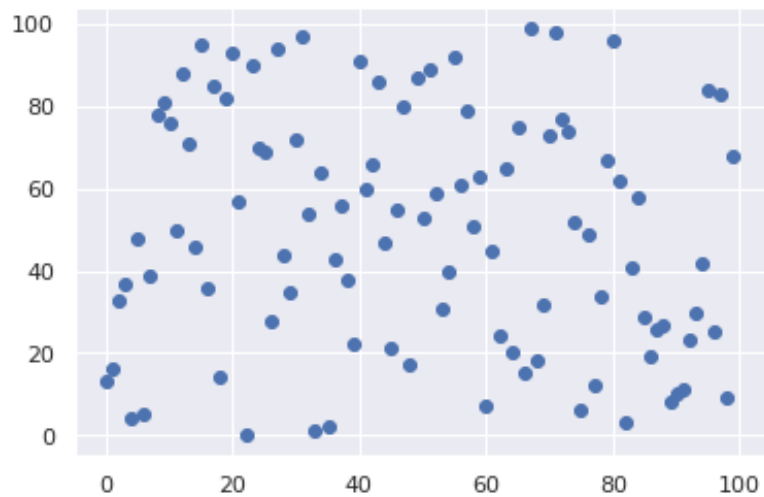


```
# Importing Library
import random
import matplotlib.pyplot as plt
import seaborn as sns
```

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random.seed(117)
sns.set()
```

```
# Creating a random dataset
x = random.sample(range(0, 100), 100)
y = random.sample(range(0, 100), 100)
dataset = [(i,j) for i,j in zip(x,y)]
```

```
plt.scatter(x,y)
plt.show()
```



```
# Class for KMeansClustering algorithm
class KMeansClustering():
    def __init__(self, dataset, k, innerIterations, outerIterations):
        self.k = k
        self.innerIterations = innerIterations
        self.outerIterations = outerIterations
        self.dataset = dataset
        self.clusters = []
        self.centroids = []

    def iteratingTheAlgorithm(self):
        for i in range(self.outerIterations):
            clusters, centroids = self.algorithm()
            self.clusters.append(clusters)
            self.centroids.append(centroids)

    def __repr__(self):
        representationString = f'Using k = {self.k} :-\n'
        for i in range(len(self.clusters)):
            representationString += f'\n\tTrial {i+1}:\n'
            for j in range(self.k):
                representationString += f'\t\tC{j+1}: {set(self.clusters[i][j])}\n'
            representationString += f'\nFinal Centroids: {self.centroids[len(self.centroids)-1]}\n'
        return representationString

    def findCentroids(self, clusters):
        centroids = []
        for i in clusters:
            xSum, ySum = 0, 0
            for j in i:
                xSum += j[0]
                ySum += j[1]
            centroids.append((xSum/len(i), ySum/len(i)))
        return centroids

    def algorithm(self):
        initialCentroids = random.sample(self.dataset, self.k)
        previousCentroids = initialCentroids
        previousClusters = [[] for i in range(self.k)]
        for i in range(self.innerIterations):
            if i == 0:
                for j in self.dataset:
```

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        distOfDataPointFromEachCentroid = [(idx, ((j[0] - k[0])**2 + (j[1] - k[1])**2)**(0.5)) for idx, k in enumerate(initialC
        centroidNearestToDataPoint = sorted(distOfDataPointFromEachCentroid, key=lambda x: x[1])[0][0]
        previousClusters[centroidNearestToDataPoint].append(j)
    tmp = []
    for j in previousClusters:
        tmp.append(frozenset(j))
    previousClusters = tmp
else:
    currentCentroids = self.findCentroids(previousClusters)
    if set(currentCentroids) == set(previousCentroids):
        return previousClusters, previousCentroids
    else:
        previousCentroids = currentCentroids
        currentClusters = [[] for i in range(self.k)]
        for j in self.dataset:
            distOfDataPointFromEachCentroid = [(idx, ((j[0] - k[0])**2 + (j[1] - k[1])**2)**(0.5)) for idx, k in enumerate(previo
            centroidNearestToDataPoint = sorted(distOfDataPointFromEachCentroid, key=lambda x: x[1])[0][0]
            currentClusters[centroidNearestToDataPoint].append(j)
        tmp = []
        for j in currentClusters:
            tmp.append(frozenset(j))
        currentClusters = tmp
        if set(currentClusters) == set(previousClusters):
            return previousClusters, previousCentroids
        else:
            previousClusters = currentClusters
return previousClusters, previousCentroids

```

```

# Input For K and Iteration and Plotting the graph
print(f'Dataset name: Random 2-D dataset\n', f'\tx = {x}', f'\ty = {y}', sep='\n')

while(True):

    k = int(input('\nEnter the value of k: '))
    print(f'The value of k is: {k}\n')

    innerIterations = int(input('Number of iterations in the algorithm: '))
    print(f'Number of iterations in the algorithm are: {innerIterations}\n')

    outerIterations = int(input('Number of times the algorithm should run: '))
    print(f'Number of times the algorithm should run is: {outerIterations}\n')

    if k > len(dataset):
        print(f'Error: 0 < Value of k <= number of points in dataset ({len(x)})\n')
    else:
        kMeansClustering = KMeansClustering(dataset, k, innerIterations, outerIterations)
        kMeansClustering.iteratingTheAlgorithm()
        print(kMeansClustering)

    if k == 2:
        clusters = kMeansClustering.clusters[len(kMeansClustering.centroids)-1]
        xCluster1 = []
        yCluster1 = []
        xCluster2 = []
        yCluster2 = []

        for i in range(len(clusters)):
            tmp = list(clusters[i])

            if i == 0:
                for j in clusters[i]:
                    xCluster1.append(j[0])
                    yCluster1.append(j[1])

            if i == 1:
                for j in clusters[i]:
                    xCluster2.append(j[0])
                    yCluster2.append(j[1])

        plt.scatter(xCluster1, yCluster1, color='deepskyblue')
        plt.scatter(xCluster2, yCluster2, color='limegreen')
        plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][0][0],
                    kMeansClustering.centroids[len(kMeansClustering.centroids)-1][0][1],
                    color='black')
        plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][1][0],
                    kMeansClustering.centroids[len(kMeansClustering.centroids)-1][1][1],
                    color='black')

```

```

plt.xlabel('x')
plt.ylabel('y')
plt.title(f'{k}- Means Clusters')
plt.show()

elif k == 3:
    clusters = kMeansClustering.clusters[len(kMeansClustering.centroids)-1]
    xCluster1 = []
    yCluster1 = []
    xCluster2 = []
    yCluster2 = []
    xCluster3 = []
    yCluster3 = []

    for i in range(len(clusters)):
        tmp = list(clusters[i])
        if i == 0:
            for j in clusters[i]:
                xCluster1.append(j[0])
                yCluster1.append(j[1])
        if i == 1:
            for j in clusters[i]:
                xCluster2.append(j[0])
                yCluster2.append(j[1])
        if i == 2:
            for j in clusters[i]:
                xCluster3.append(j[0])
                yCluster3.append(j[1])

    plt.scatter(xCluster1,yCluster1, color='deepskyblue')
    plt.scatter(xCluster2,yCluster2, color='limegreen')
    plt.scatter(xCluster3,yCluster3, color='coral')
    plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][0][0],
                kMeansClustering.centroids[len(kMeansClustering.centroids)-1][0][1],
                color='black')
    plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][1][0],
                kMeansClustering.centroids[len(kMeansClustering.centroids)-1][1][1],
                color='black')
    plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][2][0],
                kMeansClustering.centroids[len(kMeansClustering.centroids)-1][2][1],
                color='black')
    plt.xlabel('x')
    plt.ylabel('y')
    plt.title(f'{k}- Means Clusters')
    plt.show()

if k > 3 and k <= len(x):
    plt.scatter(x, y, color='deepskyblue')
    for i in range(k):
        plt.scatter(kMeansClustering.centroids[len(kMeansClustering.centroids)-1][i][0],
                    kMeansClustering.centroids[len(kMeansClustering.centroids)-1][i][1],
                    color='black')
    plt.xlabel('x')
    plt.ylabel('y')
    plt.title(f'{k}- Means Clusters')
    plt.show()

toContinue = True

while(True):
    choice = input('Do you want to continue? (Y/N): ')

    if choice == 'Y':
        toContinue = True
        break
    elif choice == 'N':
        toContinue = False
        break
    else:
        print('INVALID CHOICE!\n')

if toContinue:
    continue
else:
    break

```



x = [30, 23, 21, 26, 51, 20, 71, 52, 89, 15, 67, 72, 42, 44, 58, 83, 4, 16, 47, 55, 8, 94, 80, 49, 54, 53, 0, 78, 7]
y = [72, 90, 57, 28, 89, 93, 98, 59, 8, 95, 99, 77, 66, 47, 51, 41, 4, 36, 80, 92, 78, 42, 96, 87, 40, 31, 13, 34, 6]

Using $k = 2$:-

Trial 2:
C1: {(98, 9), (60, 7), (93, 30), (26, 28), (76, 49), (29, 35), (36, 43), (82, 3), (16, 36), (96, 25), (1, 16)}
C2: {(21, 57), (70, 73), (32, 54), (72, 77), (67, 99), (84, 58), (24, 70), (81, 62), (43, 86), (37, 56), (5, 1)}

Using $k = 3$:-

Trial 2:

| | |
|-----|--|
| C1: | {(98, 9), (60, 7), (93, 30), (82, 3), (96, 25), (85, 29), (90, 10), (66, 15), (45, 21), (75, 6), (48, 1) |
| C2: | {(84, 58), (70, 73), (63, 65), (72, 77), (67, 99), (55, 92), (79, 67), (81, 62), (43, 86), (73, 74), (5 |
| C3: | {(21, 57), (32, 54), (24, 70), (26, 28), (37, 56), (29, 35), (14, 46), (36, 43), (16, 36), (1, 16), (4, |

The scatter plot, titled "3- Means Clusters", displays three clusters of data points on a 2D coordinate system where both the X and Y axes range from 0 to 100. The clusters are represented by different colors: orange, green, and blue. Each cluster has a centroid marked by a black dot. The orange cluster is located in the upper-left region, the green cluster is in the upper-middle region, and the blue cluster is in the lower-right region. The centroids are positioned near the center of each cluster of points.

Do you want to continue? (Y/N): N

✓ 44s completed at 22:46

