

## Process

1. Understand the project
2. Use this heuristic to decide the value of K

$$n = 8$$

$$K = \sqrt{8} = 2.828427 = 3$$

3. Using KNN to manually calculate the distance and predict the result

Accelerometer Data			Gyroscope Data			Fall(+), Not(-)
x	y	z	x	y	z	(+) or (-)
1	2	3	2	1	3	-
2	1	3	3	1	2	-
1	1	2	3	2	2	-
2	2	3	3	2	1	-
6	5	7	5	6	7	+
5	6	6	6	5	7	+
5	6	7	5	7	6	+
7	6	7	6	5	6	+
7	6	5	5	6	7	??

Calculate the distance from each data set to the predicted data set:

$$(A_{pred} - A_x)^2 + (A_{pred} - A_y)^2 + (A_{pred} - A_z)^2 + (G_{pred} - G_x)^2 + (G_{pred} - G_y)^2 + (G_{pred} - G_z)^2$$

Accelerometer Data			Gyroscope Data			Distance to predict point	Fall(+), Not(-)
x	y	z	x	y	z		(+) or (-)
1	2	3	2	1	3	10.29563014	-
2	1	3	3	1	2	10.39230485	-
1	1	2	3	2	2	10.72380529	-
2	2	3	3	2	1	10.04987562	-
6	5	7	5	6	7	2.449489743	+
5	6	6	6	5	7	2.645751311	+
5	6	7	5	7	6	3.16227766	+
7	6	7	6	5	6	2.645751311	+
7	6	5	5	6	7		??

The highlighted data sets are ones selected for the prediction which are closest to the predict data set. All three samples selected are in the Fall(+) group .

**So with the predict dataset provided, it is predicted to fall(+).**

4. Use Python to implement the application of using KNN to predict fall.

➔ KNN with original code

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```
1 import math
2 # calculate the Euclidean distance between two vectors
3 #     Euclidean Distance = sqrt(sum i to N (x1_i - x2_i)^2)
4 def euclidean_distance(row1, row2):
5     distance = 0.0
6     for i in range(len(row1)-1):
7         distance += (row1[i] - row2[i])**2
8     return math.sqrt(distance)
```

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```
[25] 1 # Locate the most similar neighbors
2 def get_neighbors(train, test_row, k_neighbors):
3     distances = list()
4     for train_row in train:
5         dist = euclidean_distance(test_row, train_row)
6         distances.append((train_row, dist))
7     distances.sort(key=lambda tup: tup[1])
8     print("List of calculated distance from predict data:")
9     for i in range(len(distances)):
10         print(distances[i])
11     neighbors = list()
12     for i in range(k_neighbors):
13         neighbors.append(distances[i][0])
14     return neighbors
```

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```
[26] 1 # Make a classification prediction with neighbors
2 # - test_row is row 0
3 # - k_neighbors is 3
4 def predict_classification(train, test_row, num_neighbors):
5     neighbors = get_neighbors(train, test_row, num_neighbors)
6     print("\nNeighbors selected are: ")
7     for i in range(len(neighbors)):
8         print(neighbors[i])
9
10    output_values = [row[-1] for row in neighbors]
11    prediction = max(set(output_values), key=output_values.count)
12    return prediction
```

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```
1 # Test distance function
2 dataset = [[7,6,5,5,6,7,0],
3   [1,2,3,2,1,3,0],
4   [2,1,3,3,1,2,0],
5   [1,1,2,3,2,2,0],
6   [2,2,3,3,2,1,0],
7   [6,5,7,5,6,7,1],
8   [5,6,6,6,5,7,1],
9   [5,6,7,5,7,6,1],
10  [7,6,7,6,5,6,1]]
11
12 # row 0 (i.e., dataset[0]) is the one to be predicted
13 prediction = predict_classification(dataset[1:], dataset[0], 3)
14
15 # - Display predict result
16 print('\n')
17 'Prediction result is Fall(+)' if prediction==1 else '\nPrediction result is Not Fall(-)'
18
```



List of calculated distance from predict data:

```
([6, 5, 7, 5, 6, 7, 1], 2.449489742783178)
([5, 6, 6, 6, 5, 7, 1], 2.6457513110645907)
([7, 6, 7, 6, 5, 6, 1], 2.6457513110645907)
([5, 6, 7, 5, 7, 6, 1], 3.1622776601683795)
([2, 2, 3, 3, 2, 1, 0], 10.049875621120889)
([1, 2, 3, 2, 1, 3, 0], 10.295630140987)
([2, 1, 3, 3, 1, 2, 0], 10.392304845413264)
([1, 1, 2, 3, 2, 2, 0], 10.723805294763608)
```

Neighbors selected are:

```
[6, 5, 7, 5, 6, 7, 1]
[5, 6, 6, 6, 5, 7, 1]
[7, 6, 7, 6, 5, 6, 1]
```

'Prediction result is Fall(+)'

## ➔ KNN with sklearn

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```
[15] 1 import sklearn
      2 from sklearn.utils import shuffle
      3 from sklearn.neighbors import KNeighborsClassifier
      4 from sklearn import linear_model, preprocessing
      5 import pandas as pd
      6 import numpy as np
      7
```

✓  
9s

```
[16] 1
      2 from google.colab import files
      3 uploaded = files.upload()
```

Choose Files knn\_data.csv

- **knn\_data.csv**(text/csv) - 151 bytes, last modified: 2/6/2023 - 100% done
- Saving knn\_data.csv to knn\_data.csv

```
✓ [18] 1 data = pd.read_csv("knn_data.csv")
      2 data
```

	Ax	Ay	Az	Gx	Gy	Gz	Fall or Not
0	1	2	3	2	1	3	-
1	2	1	3	3	1	2	-
2	1	1	2	3	2	2	-
3	2	2	3	3	2	1	-
4	6	5	7	5	6	7	+
5	5	6	6	6	5	7	+
6	5	6	7	5	7	6	+
7	7	6	7	6	5	6	+

```
✓ [28] 1 Ax= list(data["Ax"])
      2 Ay = list(data["Ay"])
      3 Az = list(data["Az"])
      4 Gx = list(data["Gx"])
      5 Gy = list(data["Gy"])
      6 Gz = list(data["Gz"])
      7 fallOrNot = list(data["Fall or Not"])
      8
      9 X = list(zip(Ax, Ay,Az, Gx, Gy, Gz))
     10 Y = list(fallOrNot)
```

```
✓ [29] 1 x_train, x_test, y_train, y_test = sklearn.model_selection.train_test_split(X,Y, test_size=0.1)
      2
      3 model = KNeighborsClassifier(n_neighbors=3)
      4 model.fit(x_train, y_train)
```

KNeighborsClassifier(n\_neighbors=3)

```
✓ [30] 1 model.score(x_test, y_test)
```

1.0

```
✓ [ ] 1 print("Predicted result is ", model.predict([(7, 6, 5, 5, 6, 7)]))
```

Predicted result is ['+']

5. Comparing the result from the Python program and the result of manual calculation.

All methods conclude with the prediction of Fall(+)