

25. Project Falling Detection: Python + kNN + Colab



Process

1. Understand the project
 - Most mobile devices are equipped with different kind of [sensors](#).
 - We can use the data sent from [Gyroscope senso](#) and [Accelerometer sensor](#) to categorize any motion:
 - 3 numbers from [Accelerometer sensor](#)
 - 3 numbers from [Gyroscope sensor](#)
 - References
 - [A Review on Fall Prediction and Prevention System for Personal Devices: Evaluation and Experimental Results](#)
 - [Andorid FallArm](#) Project
 - [Sensors](#)
 - [Difference Between an Accelerometer and a Gyroscope](#)
2. Use this heuristic to decide the value of **K**
 - The choice of **K** equal to the **odd number** cloest to the **square root** of **the number of instances** is an empirical rule-of-thumb popularized by the ["Pattern Classification" book by Duda et al.](#)
 - References
 - [How can we find the optimum K in K-Nearest Neighbor?](#)
3. Using [KNN](#) to manually calculate the distance and predict the result.
 - The first 8 rows are one set of training data, including Accelerometer Data and Gyroscope Data. The last row is for prediction.

Accelerometer Data			Gyroscope Data			Fall (+), Not (-)
x	y	z	x	y	z	+/-
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	??

4. Use [Python](#) to implement the application of using kNN to predict fall.
 - Create the code by modifying [KNN from scratch](#)
 - Explain the [code](#)
 - Run the code on [Colab](#)
 - References
 - [Get Start with Colab](#)
 - a. References
 - [Python for KNN](#)
 - [- Santhi Sree Nagalla, 2021 Summer](#)
 - [Knn.ipynb](#) - Python code using Scikit-learn library
 - [Knn Python.ipynb](#) - Python Program to predict kNN Value
 - [Develop k-Nearest Neighbors in Python From Scratch](#)
 - [Quan Zhou](#) - Fall, 2019
 - [Wijian Xiong](#) - Fall, 2019
 - [Juilee Panse](#) - Fall, 2019
5. Comparing the result from the [Python program](#) and the result of [manual calculation](#).
6. [Adding the project to your portofolio](#)
 - [Please use Google Slides to document the project](#)
 - [Please link your presentation on GitHub](#) using this structure
 - - Machine Learning
 - - Supervised Learning
 - + Falling Prediction using KNN
 - Submit
 - The URLs of the Google Slides and GitHub web pages related to this project.
 - A PDF file of your Google Slides

Solution:

1. Use this heuristic to decide the value of **K**:

$$K = \sqrt{N}, \text{ where } n = \text{number of instances} = 9$$

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

2. Using [KNN](#) to manually calculate the distance and predict the result

$$D_a^2 = (d^{xta} - d^{xa})^2 + (d^{yta} - d^{ya})^2 + (d^{zta} - d^{za})^2$$

$$D_g^2 = (d^{xta} - d^{xg})^2 + (d^{yta} - d^{yg})^2 + (d^{zta} - d^{zg})^2$$

The distances which are calculated with excel sheet are given below.

Accelerometer Data			Gyroscope Data			Da	Dg	Fall (+), Not (-)
x	y	z	x	y	z			+/-
1	2	3	2	1	3	56	50	-
2	1	3	3	1	2	54	54	-
1	1	2	3	2	2	70	45	-
2	2	3	3	2	1	45	56	-
6	5	7	5	6	7	6	0	+
5	6	6	6	5	7	5	2	+
5	6	7	5	7	6	8	2	+
7	6	7	6	5	6	4	3	+
7	6	5	5	6	7			+

For $K = 3$, 3 closest instances are chosen, which are 4, 5 and 6 for Dist. Acce which represent Fall(+) and 0, 2 and 2 for Dist. Gyro which are again +/Falling. Hence, the the query instance is plus (+).

3. Use [Python](#) to implement the application of using kNN to predict fall

```
[35] import sklearn
      from sklearn.utils import shuffle
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn import linear_model, preprocessing
      import pandas as pd
      import numpy as np
```

```
[36] from google.colab import files
      uploaded = files.upload()
```

Choose Files knn_data_sample.csv
• knn_data_sample.csv(text/csv) - 152 bytes, last modified: 2/7/2023 - 100% done
Saving knn_data_sample.csv to knn_data_sample (6).csv

```
[37] data = pd.read_csv("knn_data_sample.csv")
      print(data.loc[:, : "FallOrNot"])
```

	x1	y1	z1	x2	y2	z2	FallOrNot
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	+

```
[38] x1 = list(data["x1"])
      y1 = list(data["y1"])
      z1 = list(data["z1"])
      x2 = list(data["x2"])
      y2 = list(data["y2"])
      z2 = list(data["z2"])
      fallOrNot = list(data["FallOrNot"])
```

```
[39] X = list(zip(x1, y1, z1, x2, y2, z2))
      Y = list(fallOrNot)
```

```
[40] x_train, x_test, y_train, y_test = sklearn.model_selection.train_test_split(X, Y, test_size=0.1)
```

```
[39] X = list(zip(x1, y1, z1, x2, y2, z2))
      Y = list(fallOrNot)
```

```
[40] x_train, x_test, y_train, y_test = sklearn.model_selection.train_test_split(X, Y, test_size=0.1)
```

```
[41] model = KNeighborsClassifier(n_neighbors=3)
```

```
[42] model.fit(x_train, y_train)

      KNeighborsClassifier(n_neighbors=3)
```

```
[43] model.score(x_test, y_test)

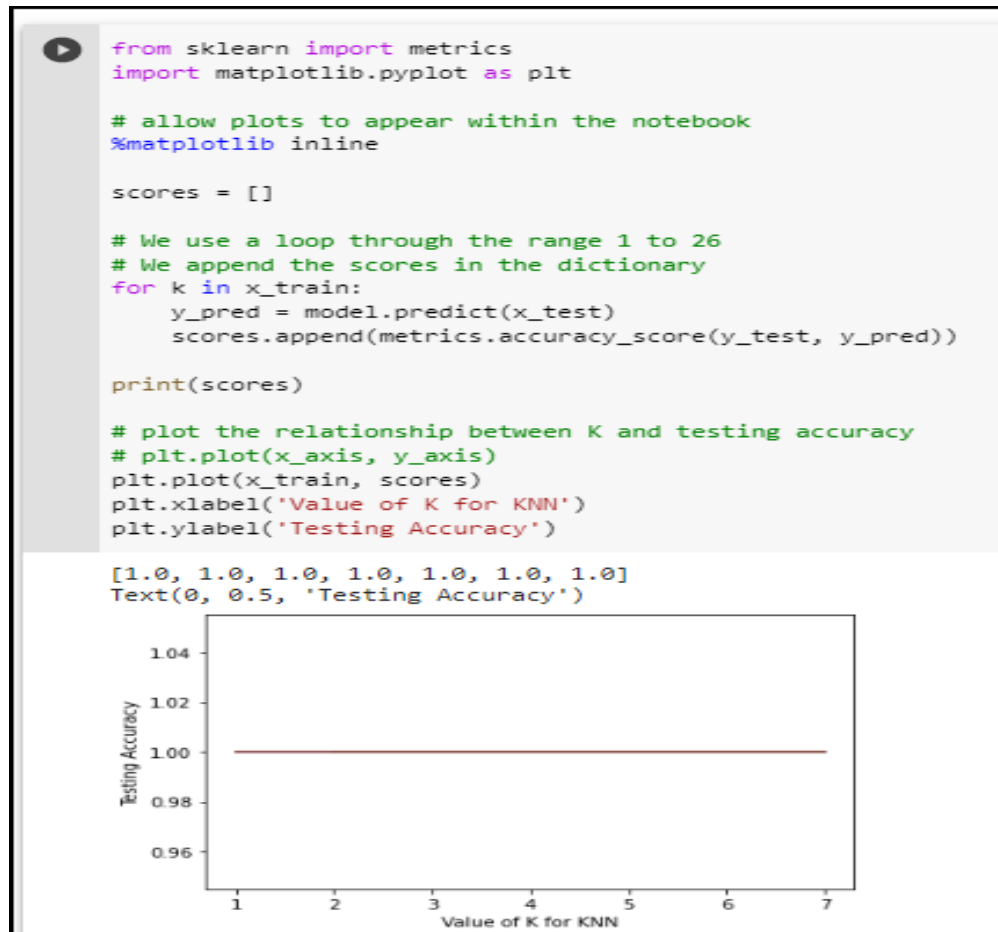
      1.0
```

```
[44] print(model.predict([(7, 6, 5, 5, 6, 7)]))

      ['+']
```

As its highlighted in yellow in the code above the predicted answer using the python code for KNN is the same as the one obtained manually, which is +

The screenshot below checks for the optimal K value, and as it can be seen all values of k from 1 to 7 generate the same accuracy of classification.



Google slides and GitHub Link

https://docs.google.com/presentation/d/1D0Eo_KUxQG6bOB5M4KofjIlsujM1RvEik_G9UPXOq6A/edit?usp=sharing

<https://github.com/momer22/Machine-Learning-Supervised-Learning-Falling-Prediction-using-KNN>