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TASK 5.1P - Store Data to Cloud

# SIT225: Data Capture Technologies

# Activity 5.1: Firebase Realtime database

The Firebase Realtime Database is a cloud-hosted NoSQL database that lets you store and sync data between your users in real-time. Data is stored as JSON and synchronized in real-time to every connected client. In this activity, you will set up and perform operations such as queries and updates on the database using Python programming language.

### Hardware Required

No hardware is required.

# Software Required

Firebase Realtime database Python 3

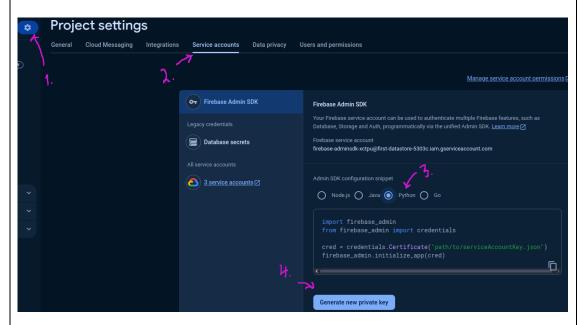
### Steps

Step	Action
1	Create an Account:
	First, you will need to create an account in the Firebase console, follow instructions in the official Firebase document
	(https://firebase.google.com/docs/database/rest/start).
2	Create a Database:
	Follow the above Firebase document to create a database. When you click on Create Database, you have to specify the location of the database and the security rules. Two rules are available – locked mode and test mode; since we will be using the database for reading, writing, and editing, we choose test mode.
3	Setup Python library for Firebase access:

We will be using Admin Database API, which is available in *firebase\_admin* library. Use the below command in the command line to install. You can follow a Firebase tutorial here (https://www.freecodecamp.org/news/how-to-get-started-with-firebase-using-python).

\$ pip install firebase\_admin

Firebase will allow access to Firebase server APIs from Google Service Accounts. To authenticate the Service Account, we require a private key in JSON format. To generate the key, go to project settings, click Generate new private key, download the file, and place it in your current folder where you will create your Python script.



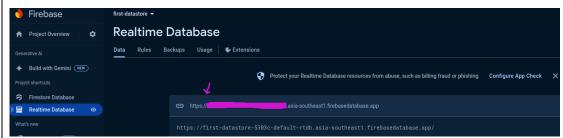
4 Connect to Firebase using Python version of Admin Database API:
A credential object needs to be created to initialise the Python library which can be done using the Python code below. Python notebook can be downloaded here (<a href="https://github.com/deakin-deep-dreamer/sit225/blob/main/week\_5/firebase\_explore.ipynb">https://github.com/deakin-deep-dreamer/sit225/blob/main/week\_5/firebase\_explore.ipynb</a>).

```
import firebase_admin

databaseURL = 'https://XXX.firebasedatabase.app/'
cred_obj = firebase_admin.credentials.Certificate(
    'first-datastore-5303c-firebase-adminsdk-xctpu-c9902044ac.json'

default_app = firebase_admin.initialize_app(cred_obj, {
    'databaseURL':databaseURL
})
```

The databaseURL is a web address to reach your Firebase database that you have created in step 2. This URL can be found in the Data tab of Realtime Database.



If you compile the code snippet above, it should do with no error.

#### Write to database Using the set() Function:

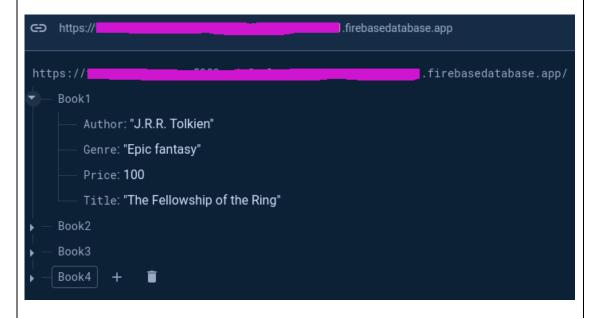
We set the reference to the root of the database (or we could also set it to a key value or child key value). Data needs to be in JSON format as below.

```
from firebase admin import db
     # before any operation is carried out on a database.
     ref = db.reference("/")
     data = { # Outer {} contains inner data structure
           "Book1":
                "Title": "The Fellowship of the Ring", "Author": "J.R.R. Tolkien",
                "Genre": "Epic fantasy",
                "Price": 100
           "Book2":
               "Title": "The Two Towers",
"Author": "J.R.R. Tolkien",
"Genre": "Epic fantasy",
"Price": 100
           "Book3":
                "Title": "The Return of the King",
                "Author": "J.R.R. Tolkien",
                "Genre": "Epic fantasy",
"Price": 100
           "Book4":
                "Title": "Brida",
"Author": "Paulo Coelho",
"Genre": "Fiction",
                "Price": 100
43 ref.set(data)
```

5

A reference point always needed to be set where the data read/write will take place. In the code above, the reference point is set at the root of the NoSQL Document, where consider the database is a JSON tree and / is the root node of the tree). The set() function writes (overwrites) data at the set reference point.

You can visualise the data in the Firebase console as below -



6 Read data using get() function:

Data can be read using get() function on the reference set beforehand, as shown below.

```
ref = db.reference("/") # set ref point
    4 books = ref.get()
     5 print(books)
    6 print(type(books))
    8 # print each item separately
    9 for key, value in books.items():
               print(f"{key}: {value}")
   13 # Query /Book1
   14 ref = db.reference("/Book1")
   15 books = ref.get()
   16 print(books)

√ 0.3s

 {'Bookl': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Titl
<class 'dict'>
Bookl: {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title':
Book1: { Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': Book3: {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': Book4: {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Price': 100, 'Title': 'Brida {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Fe
```

Consider the reference set in line 1 and the output compared to the reference set at line 14 and the bottom output line to understand the use of db.reference() and ref.get().

#### 7 Write to database Using the push() Function:

The push() function saves data under a *unique system generated key*. This is different than set() where you set the keys such as Book1, Book2, Book3 and Book4 under which the content (author, genre, price and title) appears. Let's try to push the same data in the root reference. Note that since we already has data under root / symbol, setting (or pushing) in the same reference point will eventually rewrite the original data.

The output will reset the previous data set in / node. The current data is shown below.

```
▼ — Books

▼ — Best_Sellers

▼ — -0-iqpiYlui92UKRmctM

— Author: "J.R.R. Tolkien"
— Genre: "Epic fantasy"
— Price: 100
— Title: "The Fellowship of the Ring"

▶ — -0-iqpnK8M8wjLiw2PTX
▶ — -0-iqptGIKG7WuxHdGsq
▶ — -0-iqpz_nsDjhwMzLmIw
```

As you can see, under /Books/Best\_Sellers there are 4 nodes where the node head (or node ID) is a randomly generated key which is due to the use of push() function. When data key does not matter, the use of push() function desirable.

Let's say the price of the books by J. R. R. Tolkien is reduced to 80 units to offer a discount. The first 3 books are written by this author, and we want to apply for a discount on all of them.

As you can see, the author name is compared and the new price is set in the best\_sellers dictionary and finally, an update() function is called on the ref, however, the current ref is a '/Books/Best\_Sellers/', so we need to locate the child under the ref node, so ref.child(key) is used in line 13. The output is shown below with a discounted price.



#### 9 **Delete data**:

Let's delete all bestseller books with J.R.R. Tolkien as the author. You can locate the node using db.reference() (line 4) and then locate specific record (for loop in line 6) and calling set() with empty data {} as a parameter, such as set({}). The particular child under the ref needs to be located first by using ref.child(key), otherwise, the ref node will be removed – BE CAREFUL.

This keeps only the other author data, as shown below.

```
▼ — Books

▼ — Best_Sellers

▼ — -0-iqpz_nsDjhwMzLmIw

— Author: "Paulo Coelho"

— Genre: "Fiction"

— Price: 100

— Title: "Brida"
```

If ref.child() not used, as shown the code below, all data will be removed.

```
1 ref = db.reference("/Books/Best_Sellers")
2 ref.set({})
```

Now in Firebase console you will see no data exists.

Question: Run all the cells in the Notebook you have downloaded in Step 4, fill in the student information at the top cell of the Notebook. Convert the Notebook to PDF and merge with this activity sheet PDF.

Answer: Convert the Notebook to PDF and merge with this activity sheet PDF.

Question: Create a sensor data structure for DHT22 sensor which contains attributes such as sensor\_name, timestamp, temperature and humidity. Remember there will be other sensors with different sensor variables such as DHT22 has 2 variables, accelerometer sensor has 3. For each such sensor, you will need to gather data over time. Discuss how you are going to handle multiple data values in JSON format? Justify your design.

Answer: <Your answer>

The top level objects will be the types of sensors such as DHT22, Accelerometer, Ultrasonic sensors, ... . And the inside key value will be the data and its value will be smaller keys with the **timestamp** with its corresponding datetime value and for example the DHT will be the **temperature** and **humidity** values. For Accelerometer will be the **x**, **y**, **z** as well as theirs correspondig values. Adn for Ultrasonic sensor the key will be **distance** be the key and the value will be the distance it has measured.

11 Question: Generate some random data for DHT22 sensor, insert data to database, query all data and screenshot the output here.

Answer: <Your answer>

```
ref = db.reference("/")

# Generate random data for the DHT22 sensor

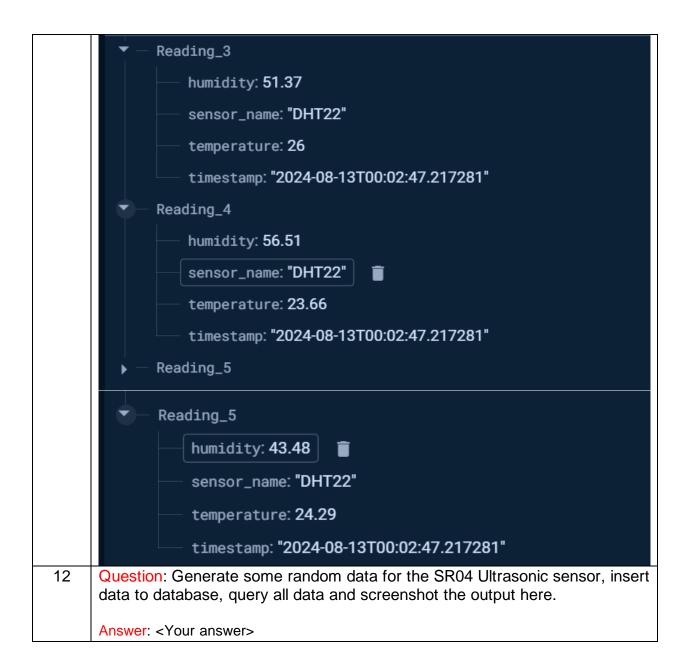
data = {
    f"Reading_{i}": {
        "sensor_name": "DHT22",
        "timestamp": datetime.now().isoformat(),
        "temperature": round(random.uniform(20.0, 30.0), 2),
        "humidity": round(random.uniform(30.0, 70.0), 2)
    } for i in range(1, 6) # Generate 5 random readings
}

# JSON format data is set (overwritten) to the reference
# point set at /, which is the root node.
#
ref.set(data)
```

```
https://truong-khang-thinh-nguyen-default-rtdb.firebaseio.com/

Reading_1
humidity: 58.07
sensor_name: "DHT22"
timestamp: "2024-08-13T00:02:47.217281"

Reading_2
humidity: 45.37
sensor_name: "DHT22"
temperature: 20.16
timestamp: "2024-08-13T00:02:47.217281"
```



```
"
ref = db.reference("/")

# Generate random data for the Ultrasonic sensor
data = {
    f"Reading_{i}": {
        "sensor_name": "Ultrasonic",
        "timestamp": datetime.now().isoformat(),
        "distance_cm": round(random.uniform(5.0, 400.0), 2)
    } for i in range(1, 3) # Generate 5 random readings
}

# Insert the data into the database
ref.set(data)
```

Question: Firebase Realtime database generates events on data operations. You can refer to section 'Handling Realtime Database events' in the document (<a href="https://firebase.google.com/docs/functions/database-events?gen=2nd">https://firebase.google.com/docs/functions/database-events?gen=2nd</a>). Discuss in the active learning session and summarise the idea of database events and how it is handled using Python SDK.

Note that these events are useful when your sensors (from Arduino script) store data directly to Firebase Realtime database and you would like to track data update actions from a central Python application such as a monitoring dashboard.

Answer: <Your answer>

Firebase Realtime Database events allow automatic triggering of actions based on data operations like creation, updates, or deletions. → Database Events.

The Python SDK primarily performs CRUD operations, and event handling is typically done using Firebase Cloud Functions. You can simulate real-time monitoring with Python by polling the database. → Python SDK Handling.

When sensors (e.g., Arduino) store data directly to Firebase, these events are useful for triggering actions, such as updating a monitoring dashboard. → Use Case.

⇒ We can actually use Firebase Cloud Functions to handle real-time events and then use Firebase Admin SDK in Python to retrieve and process data as needed. This combination leverages the strengths of both platforms, enabling efficient real-time monitoring and data handling.

# Activity 5.2: Data wrangling

Data wrangling is the process of converting raw data into a usable form. The process includes collecting, processing, analyzing, and tidying the raw data so that it can be easily read and analyzed. In this activity, you will use the common library in python, "pandas".

# Hardware Required

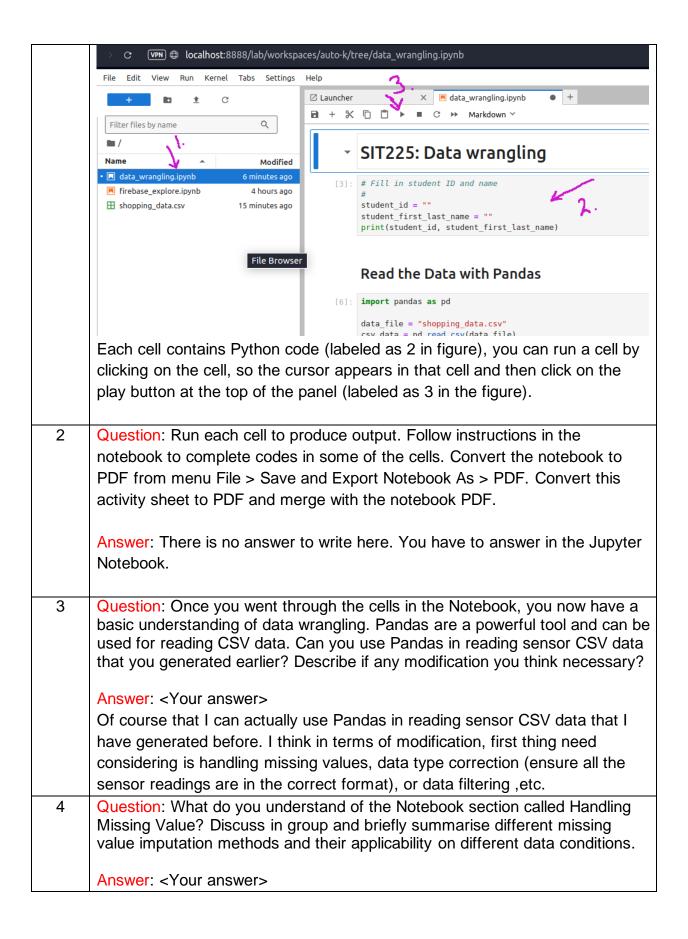
No hardware is required.

# Software Required

Python 3 Pandas Python library

# Steps

Step	Action
1	Install Pandas using the command below. Most likely you already have Pandas installed if you have installed Python using Anaconda disribution (https://www.anaconda.com/download).
	\$ pip install pandas
	A Python notebook is shared in the GitHub link (https://github.com/deakin-
	deep-dreamer/sit225/tree/main/week_5 ). There will be a
	data_wrangling.ipynb, shopping_data.csv and
	shopping_data_missingvalue.csv files among others. Download the week_5
	folder in your computer, open a command prompt in that folder, and write the command below in the command line:
	\$ jupyter lab
	This will open Python Jupyter Notebook where in the left panel you can see the files (labeled as 1 in figure).



- Removing missing data: applicable when the amount of missing data is minimal and when applying it does not affect significantly representativeness of the dataset. → Loss of valuable information, especially in small datasets.
- Mean / Median / Mode imputation: Mean and Median are appropriate when dealing with numerical data and Mode is for categorical data type
   Reduce variability and perhaps introduce bias if the missing data is not random.
- Forward and Backward fill: commonly used in time series data at which the values are expected to be continuous or similar in the course of short periods → it has to be in assumption that the missing data has a logical continuation from the adjacent values, and it might not be always the case.
- Interpolation: Continuous data where a trend or pattern between surrounding data points has been observed and assumed → Can be misleading if the data has sudden jumps or non-linear patterns.

## firebase-explore

#### August 12, 2024

```
[]: # Fill in student ID and name
      student_id = "223446545"
      student_first_last_name = "Truong Khang Thinh Nguyen"
      print(student_id, student_first_last_name)
 []: # Install libraris, if not yet.
      ! pip install firebase_admin pandas
      import firebase_admin
      databaseURL = 'https://truong-khang-thinh-nguyen-default-rtdb.firebaseio.com/'
      cred_obj = firebase_admin.credentials.Certificate(
          'truong-khang-thinh-nguyen-firebase-adminsdk-mne9g-2ae8218f25.json'
      default_app = firebase_admin.initialize_app(cred_obj, {
              'databaseURL':databaseURL
              })
[13]: from firebase_admin import db
      # A reference point is always needed to be set
      # before any operation is carried out on a database.
      ref = db.reference("/")
      # JSON format data (key/value pair)
      data = { # Outer {} contains inner data structure
              "Book1":
              {
                      "Title": "The Fellowship of the Ring",
                      "Author": "J.R.R. Tolkien",
                      "Genre": "Epic fantasy",
                      "Price": 100
              },
              "Book2":
              {
```

```
"Title": "The Two Towers",
                "Author": "J.R.R. Tolkien",
                "Genre": "Epic fantasy",
                "Price": 100
        },
        "Book3":
        {
                "Title": "The Return of the King",
                "Author": "J.R.R. Tolkien",
                "Genre": "Epic fantasy",
                "Price": 100
        },
        "Book4":
        {
                "Title": "Brida",
                "Author": "Paulo Coelho",
                "Genre": "Fiction",
                "Price": 100
        }
}
# JSON format data is set (overwritten) to the reference
# point set at /, which is the root node.
ref.set(data)
```

```
[14]: ref = db.reference("/")  # set ref point

# query all data under the ref
books = ref.get()
print(books)
print(type(books))

# print each item separately
for key, value in books.items():
    print(f"{key}: {value}")

# Query /Book1
ref = db.reference("/Book1")
books = ref.get()
print(books)
```

{'Book1': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Fellowship of the Ring'}, 'Book2': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Two Towers'}, 'Book3': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Return of the King'}, 'Book4': {'Author': 'Paulo Coelho', 'Genre':

```
'Fiction', 'Price': 100, 'Title': 'Brida'}}
     <class 'dict'>
     Book1: {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100,
     'Title': 'The Fellowship of the Ring'}
     Book2: {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100,
     'Title': 'The Two Towers'}
     Book3: {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100,
     'Title': 'The Return of the King'}
     Book4: {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Price': 100, 'Title':
     'Brida'}
     {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title':
     'The Fellowship of the Ring'}
[15]: print(data)
     {'Book1': {'Title': 'The Fellowship of the Ring', 'Author': 'J.R.R. Tolkien',
     'Genre': 'Epic fantasy', 'Price': 100}, 'Book2': {'Title': 'The Two Towers',
     'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100}, 'Book3':
     {'Title': 'The Return of the King', 'Author': 'J.R.R. Tolkien', 'Genre': 'Epic
     fantasy', 'Price': 100}, 'Book4': {'Title': 'Brida', 'Author': 'Paulo Coelho',
     'Genre': 'Fiction', 'Price': 100}}
[16]: # Write using push() function
      # Note that a set() is called on top of push()
      ref = db.reference("/")
      ref.set({
              "Books":
              ₹
                      "Best_Sellers": -1
              }
      })
      ref = db.reference("/Books/Best_Sellers")
      for key, value in data.items():
              ref.push().set(value)
[17]: # Update data
      # Requirement: The price of the books by
      # J. R. R. Tolkien is reduced to 80 units to
      # offer a discount.
      ref = db.reference("/Books/Best_Sellers/")
      best sellers = ref.get()
      print(best_sellers)
```

```
for key, value in best_sellers.items():
              if(value["Author"] == "J.R.R. Tolkien"):
                      value["Price"] = 90
                      ref.child(key).update({"Price":80})
     {'-045dqWXMwb_JudJ9X3c': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy',
     'Price': 100, 'Title': 'The Fellowship of the Ring'}, '-045dqcz5UwutpQYFqXS':
     {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Title':
     'The Two Towers'}, '-045dqkFGYxDTtamQkbm': {'Author': 'J.R.R. Tolkien', 'Genre':
     'Epic fantasy', 'Price': 100, 'Title': 'The Return of the King'},
     '-045dqrnhcMlTw_yYwjL': {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Price':
     100, 'Title': 'Brida'}}
[18]: # Let's delete all best seller books
      # with J.R.R. Tolkien as the author.
      ref = db.reference("/Books/Best Sellers")
      for key, value in best_sellers.items():
              if(value["Author"] == "J.R.R. Tolkien"):
                      ref.child(key).set({})
[19]: # Delete all best seller data.
      ref = db.reference("/Books/Best Sellers/")
      best_sellers = ref.get()
      print(best_sellers)
      print(type(best_sellers))
     {'-O45dqrnhcMlTw_yYwjL': {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Price':
     100, 'Title': 'Brida'}}
     <class 'dict'>
[20]: ref = db.reference("/Books/Best_Sellers")
      ref.set({})
```

# data-wrangling

August 12, 2024

### 1 SIT225: Data wrangling

Run each cell to generate output and finally convert this notebook to PDF.

```
[1]: # Fill in student ID and name
#
student_id = "223446545"
student_first_last_name = "Truong Khang Thinh Nguyen"
print(student_id, student_first_last_name)
```

223446545 Truong Khang Thinh Nguyen

#### 1.1 Read the Data with Pandas

Pandas has a dedicated function read\_csv() to read CSV files.

Just in case we have a large number of data, we can just show into only five rows with head function. It will show you 5 rows data automatically.

```
[2]: import pandas as pd

data_file = "shopping_data.csv"
    csv_data = pd.read_csv(data_file)

print(csv_data)

# show into only five rows with head function
    print(csv_data.head())
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40
	•••			•••	•••
195	196	Female	35	120	79
196	197	Female	45	126	28
197	198	Male	32	126	74

198	19	99 Mal	e 3	2		13	7	18
199	20	00 Mal	е 3	0		13	7	83
[20	0 rows x 5	columns]						
	CustomerID	Genre	Age	Annual	Income	(k\$)	Spending Score	(1-100)
0	1	Male	19			15		39
1	2	Male	21			15		81
2	3	Female	20			16		6
3	4	Female	23			16		77
4	5	Female	31			17		40

#### 1.2 Access the Column

Pandas has provided function .columns to access the column of the data source.

```
[3]: print(csv_data.columns)
     # if we want to access just one column, for example "Age"
     print("Age:")
     print(csv_data["Age"])
    Index(['CustomerID', 'Genre', 'Age', 'Annual Income (k$)',
            'Spending Score (1-100)'],
          dtype='object')
    Age:
    0
            19
    1
            21
    2
           20
    3
            23
    4
           31
            . .
    195
           35
    196
            45
    197
            32
            32
    198
    199
    Name: Age, Length: 200, dtype: int64
```

#### 1.3 Access the Row

In addition to accessing data through columns, using pandas can also access using rows. In contrast to access through columns, the function to display data from a row is the .iloc[i] function where [i] indicates the order of the rows to be displayed where the index starts from 0.

```
[4]: # we want to know what line 5 contains

print(csv_data.iloc[5])

print()
```

```
# We can combine both of those function to show row and column we want.
# For the example, we want to show the value in column "Age" at the first row
# (remember that the row starts at 0)
#
print(csv_data["Age"].iloc[1])
```

```
CustomerID 6
Genre Female
Age 22
Annual Income (k$) 17
Spending Score (1-100) 76
Name: 5, dtype: object
```

21

#### 1.4 Show Data Based on Range

After displaying a data set, what if you want to display data from rows 5 to 20 of a dataset? To anticipate this, pandas can also display data within a certain range, both ranges for rows only, only columns, and ranges for rows and columns

```
[9]: print("Shows data to 5th to less than 20th in a row:") print(csv_data.iloc[5:20])
```

Shows data to 5th to less than 20th in a row:

	${\tt CustomerID}$	Genre	Age	Annual	Income	(k\$)	Spending Score (1-100)
5	6	Female	22			17	76
6	7	Female	35			18	6
7	8	Female	23			18	94
8	9	Male	64			19	3
9	10	Female	30			19	72
10	11	Male	67			19	14
11	12	Female	35			19	99
12	13	Female	58			20	15
13	14	Female	24			20	77
14	15	Male	37			20	13
15	16	Male	22			20	79
16	17	Female	35			21	35
17	18	Male	20			21	66
18	19	Male	52			23	29
19	20	Female	35			23	98

#### 1.5 Using Numpy to Show the Statistic Information

The describe() function allows to quickly find statistical information from a dataset. Those information such as mean, median, modus, max min, even standard deviation. Don't forget to install Numpy before using describe function.

```
[6]: print(csv_data.describe(include="all"))
```

	${\tt CustomerID}$	Genre	Age	Annual Income (k\$)	\
count	200.000000	200	200.000000	200.000000	
unique	NaN	2	NaN	NaN	
top	NaN	Female	NaN	NaN	
freq	NaN	112	NaN	NaN	
mean	100.500000	NaN	38.850000	60.560000	
std	57.879185	NaN	13.969007	26.264721	
min	1.000000	NaN	18.000000	15.000000	
25%	50.750000	NaN	28.750000	41.500000	
50%	100.500000	NaN	36.000000	61.500000	
75%	150.250000	NaN	49.000000	78.000000	
max	200.000000	NaN	70.000000	137.000000	

Spending Score (1-100) 200.000000 count unique NaNtop NaNNaNfreq 50.200000 mean 25.823522 std 1.000000 min 25% 34.750000 50% 50.000000

#### 1.6 Handling Missing Value

73.000000

99.000000

```
[7]: # For the first step, we will figure out if there is missing value.

print(csv_data.isnull().values.any())
print()
```

False

75%

max

```
[8]: # We will use another data source with missing values to practice this part.
data_missing = pd.read_csv("shopping_data_missingvalue.csv")
print(data_missing.head())

print()

print("Missing? ", data_missing.isnull().values.any())
```

```
        CustomerID
        Genre
        Age
        Annual Income (k$)
        Spending Score (1-100)

        0
        1
        Male
        19.0
        15.0
        39.0

        1
        2
        Male
        NaN
        15.0
        81.0
```

```
2 3 Female 20.0 NaN 6.0
3 4 Female 23.0 16.0 77.0
4 5 Female 31.0 17.0 NaN
```

Missing? True

[]:

#### 1.6.1 Ways to deal with missing values.

Follow the tutorial (https://deepnote.com/app/rickyharyanto14-3390/Data-Wrangling-w-Python-e5d1a23e-33cf-416d-ad27-4c3f7f467442). It includes - 1. Delete data \* deleting rows \* pairwise deletion \* delete column 2. imputation \* time series problem - Data without trend with seasonality (mean, median, mode, random) - Data with trend and without seasonality (linear interpolation) \* general problem - Data categorical (Make NA as multiple imputation) - Data numerical or continuous (mean, median, mode, multiple imputation and linear regression)

#### 1.6.2 Filling with Mean Values

The mean is used for data that has a few outliers/noise/anomalies in the distribution of the data and its contents. This value will later fill in the empty value of the dataset that has a missing value case. To fill in an empty value use the fillna() function

```
[10]: print(data_missing.mean())

"""

Question: This code will generate error. Can you explain why and how it can be_\preceded \( \sigma \sigma \) solved?

Move on to the next cell to find one way it can be solved.

Answer: <your answer>

Because in the data_missing dataframe, we are currently having the non-numeric_\preceded \( \sigma \) column which is the Genre so it will terminate the program. Instead make sure to exclude that column.

"""
```

```
ValueError
                                                                                                    Traceback (most recent call last)
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:1683, in_
   ⇔_ensure_numeric(x)
       1682 try:
-> 1683
                            x = x.astype(np.float64)
       1684 except ValueError as err:
       1685
                             # GH#29941 we get here with object arrays containing strs
ValueError: could not convert string to float:
   →'MaleMaleFemaleFemaleFemaleFemaleFemaleFemaleMaleFemaleMaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemal
The above exception was the direct cause of the following exception:
TypeError
                                                                                                    Traceback (most recent call last)
Cell In[10], line 1
---> 1 print(data_missing.mean())
              3 """
              Δ
              5 Question: This code will generate error. Can you explain why and how it
   (...)
           10 """
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\generic.py:11556, in_
   →numeric_only, **kwargs)
    11539 @doc(
    11540
                             _num_doc,
                            desc="Return the mean of the values over the requested axis.",
    11541
       (...)
    11554
                            **kwargs,
    11555):
                            return NDFrame.mean(self, axis, skipna, numeric_only, **kwargs)
> 11556
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\generic.py:11201, in_
   →NDFrame.mean(self, axis, skipna, numeric_only, **kwargs)
    11194 def mean(
    11195
    11196
                            axis: Axis | None = 0,
       (...)
    11199
                            **kwargs,
    11200 ) -> Series | float:
> 11201
                            return self._stat_function(
                                      "mean", nanops.nanmean, axis, skipna, numeric_only, **kwargs
    11202
    11203
                            )
```

```
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\generic.py:11158, in_
 →NDFrame._stat_function(self, name, func, axis, skipna, numeric_only, **kwargs
            nv.validate_stat_func((), kwargs, fname=name)
  11156 validate_bool_kwarg(skipna, "skipna", none_allowed=False)
> 11158 return self. reduce(
            func, name=name, axis=axis, skipna=skipna, numeric_only=numeric_onl
  11159
  11160 )
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\frame.py:10519, in_
 DataFrame. reduce(self, op, name, axis, skipna, numeric_only, filter_type, u
 →**kwds)
  10515
            df = df.T
  10517 # After possibly _get_data and transposing, we are now in the
 10518 # simple case where we can use BlockManager.reduce
> 10519 res = df. mgr.reduce(blk func)
  10520 out = df._constructor(res).iloc[0]
  10521 if out_dtype is not None:
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\internals\managers.py:
 →1534, in BlockManager.reduce(self, func)
   1532 res_blocks: list[Block] = []
   1533 for blk in self.blocks:
-> 1534
           nbs = blk.reduce(func)
            res blocks.extend(nbs)
   1535
   1537 index = Index([None]) # placeholder
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\internals\blocks.py:339
 ⇔in Block.reduce(self, func)
    333 Ofinal
    334 def reduce(self, func) -> list[Block]:
            # We will apply the function and reshape the result into a single-r w
    335
            # Block with the same mgr locs; squeezing will be done at a higher
    336
 ⇔level
    337
            assert self.ndim == 2
--> 339
           result = func(self.values)
            if self.values.ndim == 1:
    341
                # TODO(EA2D): special case not needed with 2D EAs
    342
    343
                res_values = np.array([[result]])
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\frame.py:10482, in_
 ⇔DataFrame._reduce.<locals>.blk_func(values, axis)
  10480
           return values._reduce(name, skipna=skipna, **kwds)
  10481 else:
> 10482
            return op(values, axis=axis, skipna=skipna, **kwds)
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:96, in_
 ⇔disallow. call .<locals>. f(*args, **kwargs)
     94 try:
```

```
with np.errstate(invalid="ignore"):
---> 96
                                 return f(*args, **kwargs)
          97 except ValueError as e:
                         # we want to transform an object array
                         # ValueError message to the more typical TypeError
          99
                         # e.g. this is normally a disallowed function on
        100
                         # object arrays that contain strings
        101
                         if is_object_dtype(args[0]):
        102
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:158, in_
   sbottleneck switch.__call__.<locals>.f(values, axis, skipna, **kwds)
        156
                                 result = alt(values, axis=axis, skipna=skipna, **kwds)
        157 else:
                         result = alt(values, axis=axis, skipna=skipna, **kwds)
--> 158
         160 return result
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:421, in_
   → datetimelike compat.<locals>.new func(values, axis, skipna, mask, **kwargs)
        418 if datetimelike and mask is None:
                        mask = isna(values)
--> 421 result = func(values, axis=axis, skipna=skipna, mask=mask, **kwargs)
        423 if datetimelike:
                         result = _wrap_results(result, orig_values.dtype, fill_value=iNaT)
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:727, in_
   →nanmean(values, axis, skipna, mask)
                         dtype_count = dtype
        726 count = _get_counts(values.shape, mask, axis, dtype=dtype_count)
--> 727 the sum = _ensure_numeric(values.sum(axis, dtype=dtype sum))
        729 if axis is not None and getattr(the_sum, "ndim", False):
                         count = cast(np.ndarray, count)
        730
File ~\Documents\ANACONDA\Lib\site-packages\pandas\core\nanops.py:1686, in_
   →_ensure_numeric(x)
                                 x = x.astype(np.float64)
      1683
                         except ValueError as err:
      1684
                                 # GH#29941 we get here with object arrays containing strs
      1685
-> 1686
                                 raise TypeError(f"Could not convert {x} to numeric") from err
      1687 else:
                         if not np.any(np.imag(x)):
      1688
TypeError: Could not convert
   →['MaleMaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemaleFemale
   →to numeric
```

[11]: # Genre column contains string values and numerial operation mean fails.
# Lets drop Genre column since for numerial calculation.

```
data_missing_wo_genre = data_missing.drop(columns=['Genre'])
      print(data_missing_wo_genre.head())
        CustomerID
                           Annual Income (k$)
                                               Spending Score (1-100)
                      Age
     0
                 1 19.0
                                         15.0
                                                                  39.0
                                         15.0
                                                                  81.0
     1
                 2
                    NaN
                 3 20.0
                                                                   6.0
     2
                                          NaN
     3
                 4 23.0
                                         16.0
                                                                  77.0
                 5 31.0
                                         17.0
                                                                   NaN
[12]: print(data_missing_wo_genre.mean())
     CustomerID
                                100.500000
     Age
                                 38.939698
     Annual Income (k$)
                                 61.005051
     Spending Score (1-100)
                                 50.489899
     dtype: float64
[13]: print("Dataset with empty values! :")
      print(data_missing_wo_genre.head(10))
      data_filling=data_missing_wo_genre.fillna(data_missing_wo_genre.mean())
      print("Dataset that has been processed Handling Missing Values with Mean :")
      print(data_filling.head(10))
      # Observe the missing value imputation in corresponding rows.
     Dataset with empty values! :
                          Annual Income (k$)
        CustomerID
                     Age
                                               Spending Score (1-100)
                                         15.0
     0
                 1
                   19.0
                                                                  39.0
                 2
                    NaN
                                         15.0
                                                                  81.0
     1
                 3 20.0
     2
                                          \mathtt{NaN}
                                                                   6.0
     3
                 4 23.0
                                         16.0
                                                                  77.0
     4
                 5 31.0
                                         17.0
                                                                   NaN
     5
                 6 22.0
                                          NaN
                                                                  76.0
     6
                 7 35.0
                                         18.0
                                                                   6.0
     7
                 8 23.0
                                         18.0
                                                                  94.0
     8
                 9 64.0
                                         19.0
                                                                   NaN
                10 30.0
     9
                                         19.0
                                                                  72.0
     Dataset that has been processed Handling Missing Values with Mean :
                           Age Annual Income (k$)
        CustomerID
                                                    Spending Score (1-100)
     0
                 1 19.000000
                                         15.000000
                                                                  39.000000
     1
                 2 38.939698
                                         15.000000
                                                                  81.000000
     2
                 3 20.000000
                                         61.005051
                                                                   6.000000
     3
                 4 23.000000
                                         16.000000
                                                                  77.000000
                                                                  50.489899
                 5 31.000000
                                         17.000000
```

5	6	22.000000	61.005051	76.000000
6	7	35.000000	18.000000	6.000000
7	8	23.000000	18.00000	94.000000
8	9	64.000000	19.00000	50.489899
9	10	30.000000	19.000000	72.000000

#### 1.6.3 Filling with Median

The median is used when the data presented has a high outlier. The median was chosen because it is the middle value, which means it is not the result of calculations involving outlier data. In some cases, outlier data is considered disturbing and often considered noisy because it can affect class distribution and interfere with clustering analysis.

```
[14]: print(data_missing_wo_genre.median())
print("Dataset with empty values! :")
print(data_missing_wo_genre.head(10))

data_filling2=data_missing_wo_genre.fillna(data_missing_wo_genre.median())
print("Dataset that has been processed Handling Missing Values with Median :")
print(data_filling2.head(10))

# Observe the missing value imputation in corresponding rows.
#
```

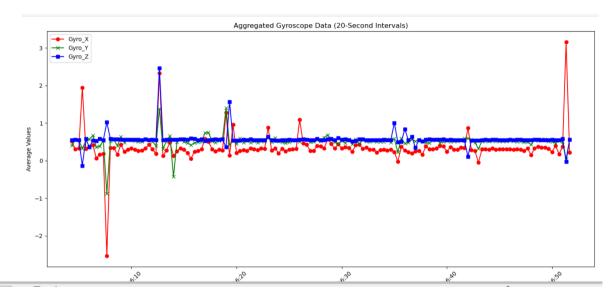
```
CustomerID
                           100.5
Age
                            36.0
Annual Income (k$)
                            62.0
Spending Score (1-100)
                            50.0
dtype: float64
Dataset with empty values! :
   CustomerID
                Age
                     Annual Income (k$)
                                           Spending Score (1-100)
0
               19.0
                                     15.0
                                                              39.0
            1
                NaN
                                     15.0
                                                              81.0
1
2
            3 20.0
                                      NaN
                                                               6.0
                                     16.0
3
               23.0
                                                              77.0
            4
4
            5
               31.0
                                     17.0
                                                               NaN
5
            6 22.0
                                      NaN
                                                              76.0
6
                                                               6.0
            7
               35.0
                                     18.0
7
            8 23.0
                                     18.0
                                                              94.0
8
            9
               64.0
                                     19.0
                                                               NaN
9
               30.0
                                     19.0
                                                              72.0
           10
Dataset that has been processed Handling Missing Values with Median :
                                           Spending Score (1-100)
   CustomerID
                Age
                      Annual Income (k$)
            1 19.0
0
                                     15.0
                                                              39.0
            2 36.0
                                     15.0
                                                              81.0
1
            3 20.0
2
                                     62.0
                                                               6.0
3
            4 23.0
                                     16.0
                                                              77.0
4
            5 31.0
                                     17.0
                                                              50.0
```

5	6	22.0	62.0	76.0
6	7	35.0	18.0	6.0
7	8	23.0	18.0	94.0
8	9	64.0	19.0	50.0
9	10	30.0	19.0	72.0

#### **TASK 5.1P - SIT225**

#### TRUONG KHANG THINH NGUYEN

**Hypothesis**: If the Arduino Nano 33 IoT with the gyroscope sensor is placed on a stable table for one hour, the x, y, and z readings from the gyroscope will exhibit minor fluctuations primarily due to sensor noise and drift. Despite these fluctuations, the overall values for the x, y and z axes will remain relatively stable, with any significant deviations likely attributed to sensor bias or external environmental factors.



As we can see from our generated graph from our recorded data, our assumption is true. We can see in some specific point of time, the x,y,z values did drop or increase significantly but then remained stable throughout the period of 1 hour.

Another point to know that , since during the time of implementing the work, I did put the Arduino align a little bit, which cause the x values remain smaller compared to other 2 y and z values. (Since y represents up and down, z represents forward and backward so they didn't change much is obvious since I kept the Arduino stayed the same on my table).

#### **Python Program:**

import firebase\_admin

import serial

from firebase\_admin import db

from datetime import datetime

import pandas as pd

import matplotlib.pyplot as plt

import json

import csv

```
databaseURL = 'https://truong-khang-thinh-nguyen-default-rtdb.firebaseio.com/'
cred_obj = firebase_admin.credentials.Certificate(
  'truong-khang-thinh-nguyen-firebase-adminsdk-mne9g-2ae8218f25.json'
)
default_app = firebase_admin.initialize_app(cred_obj, {
       'databaseURL':databaseURL
       })
ref = db.reference("/")
# Set the baud rate
baud rate = 9600
# Create a serial port communication
s = serial.Serial(port = "COM7", baudrate = baud_rate, timeout=20)
while True:
  gyro_data = s.readline().decode("utf-8").strip()
  # Timestamp
  timestamp = datetime.now().strftime("%Y-%m-%d %H:%M:%S")
  gyro_x , gyro_y , gyro_z = gyro_data.split(",")
  # Send data to Firebase using push
  ref.push({
  "Timestamp": timestamp,
  "Gyro_X": gyro_x,
  "Gyro_Y": gyro_y,
  "Gyro_Z": gyro_z
  })
# Load JSON data from a file
with open('Task 5.1P.json', 'r') as file:
  data = json.load(file)
```

```
# Convert the dictionary to a pandas DataFrame
df = pd.DataFrame.from_dict(data, orient='index')
# Save the DataFrame to a CSV file
df.to_csv('gyro_data.csv', index=False)
# Load converted CSV file from JSON file
df = pd.read_csv("gyro_data.csv")
df
# Check null valuesdf
df.info()
df['Timestamp'] = pd.to_datetime(df['Timestamp'])
# Set the Timestamp column as the index
df.set_index('Timestamp', inplace=True)
# Aggregate data to 20-second intervals, calculating the mean for each interval
df_aggregated = df.resample('20S').mean()
plt.figure(figsize=(14, 7))
plt.plot(df_aggregated.index, df_aggregated['Gyro_X'], label='Gyro_X', color='r', linestyle='-',
marker='o')
plt.plot(df_aggregated.index, df_aggregated['Gyro_Y'], label='Gyro_Y', color='g', linestyle='-',
marker='x')
plt.plot(df_aggregated.index, df_aggregated['Gyro_Z'], label='Gyro_Z', color='b', linestyle='-',
marker='s')
# Formatting the plot
plt.xlabel('Timestamp')
plt.ylabel('Average Values')
plt.title('Aggregated Gyroscope Data (20-Second Intervals)')
plt.legend()
```

#### plt.xticks(rotation=45)

# Load converted CSV file from JSON file
df = pd.read\_csv("gyro\_data.csv")

df

#### plt.tight\_layout()

```
import firebase_admin
import serial
\textbf{from} \  \, \texttt{firebase\_admin} \  \, \textbf{import} \  \, \texttt{db}
from datetime import datetime
import pandas as pd
import matplotlib.pyplot as plt
import ison
import csv
databaseURL = 'https://truong-khang-thinh-nguyen-default-rtdb.firebaseio.com/'
cred_obj = firebase_admin.credentials.Certificate(
    truong-khang-thinh-nguyen-firebase-adminsdk-mne9g-2ae8218f25.json'
default_app = firebase_admin.initialize_app(cred_obj, {
──*'databaseURL':databaseURL
ref = db.reference("/")
# Set the baud rate
baud rate = 9600
# Create a serial port communication
s = serial.Serial(port = "COM7", baudrate = baud_rate, timeout=20)
while True:
    gyro_data = s.readline().decode("utf-8").strip()
    # Timestamp
    timestamp = datetime.now().strftime("%Y-%m-%d %H:%M:%S")
    gyro_x , gyro_y , gyro_z = gyro_data.split(",")
    # Send data to Firebase using push
    ref.push({
    "Timestamp" : timestamp ,
    "Gyro_X": gyro_x ,
    "Gyro_Y" : gyro_y ,
    "Gyro_Z" : gyro_z
# Load JSON data from a file
with open('Task 5.1P.json', 'r') as file:
    data = json.load(file)
# Convert the dictionary to a pandas DataFrame
df = pd.DataFrame.from_dict(data, orient='index')
# Save the DataFrame to a CSV file
df.to_csv('gyro_data.csv', index=False)
```

[50]:		Gyro_X	Gyro_Y	Gyro_Z	Timestamp
	0	1.77	-0.37	0.55	2024-08-13 16:04:34
	1	0.49	0.92	0.55	2024-08-13 16:04:35
	2	0.24	0.55	0.55	2024-08-13 16:04:36
	3	0.24	0.43	0.55	2024-08-13 16:04:37
	4	0.12	0.37	0.55	2024-08-13 16:04:38
	2820	0.31	0.37	0.61	2024-08-13 16:51:40
	2821	0.37	0.37	0.55	2024-08-13 16:51:41
	2822	-0.18	0.92	0.55	2024-08-13 16:51:42
	2823	0.43	0.67	0.55	2024-08-13 16:51:43
	2824	0.18	0.49	0.55	2024-08-13 16:51:44

2825 rows × 4 columns

There are not any null values in the dataset. So we don't need to handle the missing values.

```
df['Timestamp'] = pd.to_datetime(df['Timestamp'])

# Set the Timestamp column as the index
df.set_index('Timestamp', inplace=True)

# Aggregate data to 20-second intervals, calculating the mean for each interval
df_aggregated = df.resample('205').mean()

plt.figure(figsize=(14, 7))

plt.plot(df_aggregated.index, df_aggregated['Gyro_X'], label='Gyro_X', color='r', linestyle='-', marker='o')
plt.plot(df_aggregated.index, df_aggregated['Gyro_Y'], label='Gyro_Y', color='g', linestyle='-', marker='x')
plt.plot(df_aggregated.index, df_aggregated['Gyro_Z'], label='Gyro_Z', color='b', linestyle='-', marker='s')

# Formatting the plot
plt.xlabel('Timestamp')
plt.ylabel('Average Values')
plt.title('Aggregated Gyroscope Data (20-Second Intervals)')
plt.legend()
plt.xticks(rotation=45)
plt.tight_layout()
```

Basically the Python script tried to open the connection between the Python script and the Firebase database. Then it connected with the Arduino IDE via the Serial Communication and then used the data recorded from the Gyroscope sensor from the Arduino Iot33

The next stage is that after it received the data from the Arduino it appended the row data to the firebase and just from that after 1 hour of executing. I stopped the program to receive any further data.

In terms of converting the CSV file, firstly I downloaded the JSON file from the Firebase and then wrote a program to convert it into CSV file. Then I plotted the x,y,z variables together with checking null values and interpreting the generated graph.

#### **Arduino Code:**

```
#include <Arduino_LSM6DS3.h>
float gyro_x, gyro_y, gyro_z;
void setup() {
 Serial.begin(9600); // set baud rate
 while (!Serial); // wait for port to init
 if (!IMU.begin()) {
  Serial.println("Failed to initialize IMU!");
  while (1);
 }
}
void loop() {
 // read accelero data
 if (IMU.gyroscopeAvailable()) {
  IMU.readGyroscope(gyro_x, gyro_y, gyro_z);
 }
 Serial.println(
  String(gyro_x) + ", " + String(gyro_y) + ", " + String(gyro_z));
 delay(1000);
```

}

For the Arduino IDE, after including the necessary packages for measuring the Gryroscope data and then it sent the recorded data to the Serial communication. And its sampling rate is 1, I think this one is the appropriate sampling rate since it provides a good balance between capturing detailed rotational data and maintaining manageable data volumes. This rate is suitable to detect meaningful changes in movement and avoids any redundant data that has already been recorded.

#### **Link Video Demonstrating**

https://deakin.au.panopto.com/Panopto/Pages/Viewer.aspx?id=5435cc4c-66ce-4202-ba8c-b1cb00992fa7