Hypothesis

The gyroscope readings (X, Y, Z axes) should display consistent patterns, particularly when the device undergoes smooth, repetitive motions. If there are rapid spikes or dips, they might indicate abrupt movements or me getting tired moving the chair.

Data Collection

I collected gyroscope data (X, Y, Z axes) to analyze the rotational movement of the device while spinning myself and moving sitting in a desk chair. This can be used to detect orientation changes, motion patterns, or even sudden movements like shakes or twists.

I Collected the data continuously over a period of around 10 minutes. Sampling at 1-second intervals should provide enough granularity to identify meaningful changes in motion.

Observation and Analysis

Pattern Observation:

Repeating Patterns: I moved the device in a cyclic motion so you'll see repeating patterns but i made sure to change directing and move around to get various data. You may observe downward trends in the axes if the device is slowly tilted in a single direction. Relative Changes: By plotting all three axes together, you can see how the X, Y, and Z gyroscope readings change relative to one another.

If the hypothesis is that there will be a repeating pattern during a circular or repetitive motion, and the data supports this, the hypothesis holds. If the hypothesis doesn't hold, potential reasons could be inconsistencies in the motion, errors in the sensor readings, or environmental factors.

```
In [5]: import firebase_admin
    from firebase_admin import credentials, db
    import pandas as pd
    import csv
    import time

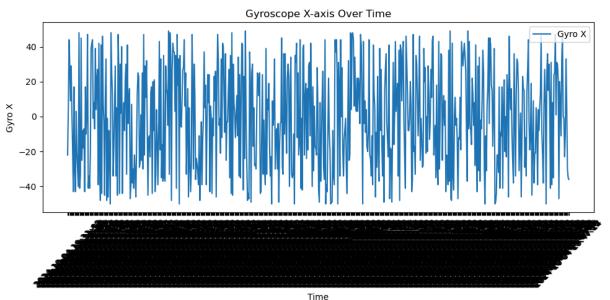
# Initialize Firebase
    cred_obj = firebase_admin.credentials.Certificate(
        r'C:\Users\ramma\Downloads\sit225-5-firebase-adminsdk-o982n-f4f8bea959.j
)
    firebase_admin.initialize_app(cred_obj, {
```

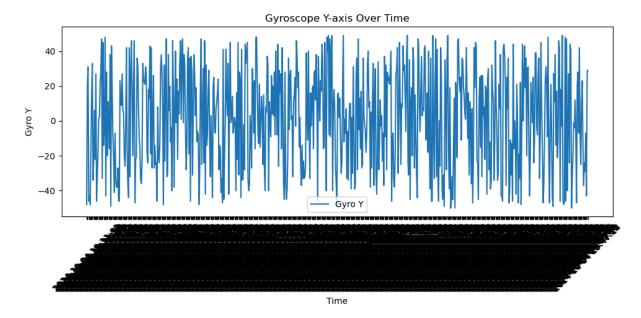
```
'databaseURL': 'https://sit225-5-default-rtdb.firebaseio.com/'
})
# Reference to the Firebase data
ref = db.reference('/mpu6050 data')
# Get all data
all data = ref.get()
# Convert data to CSV format and save as a CSV file
with open('gyro data.csv', mode='w', newline='') as file:
   writer = csv.writer(file)
   # Write the header
   writer.writerow(["timestamp", "gyro_x", "gyro y", "gyro z"])
   # Write the data
    for key, value in all data.items():
        writer.writerow([value["timestamp"], value["gyro x"], value["gyro y"
# Load the CSV into a Pandas DataFrame for cleaning and analysis
df = pd.read csv('gyro data.csv')
# Cleaning the data (removing non-number and empty fields)
df.dropna(inplace=True) # Removes rows with empty fields
# Save the cleaned data back to CSV
df.to csv('cleaned gyro data.csv', index=False)
print("Cleaned data saved to 'cleaned gyro data.csv'")
```

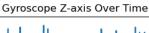
Cleaned data saved to 'cleaned gyro data.csv'

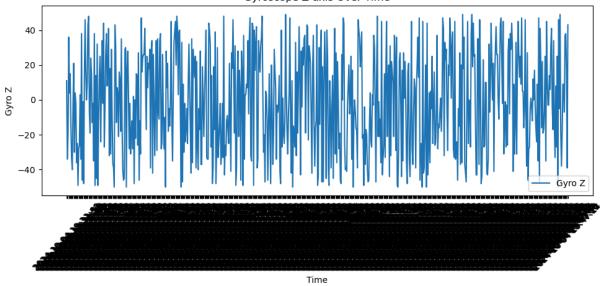
```
In [8]: import matplotlib.pyplot as plt
        # Load the cleaned data
        df = pd.read csv('cleaned gyro data.csv')
        # Plot X-axis Gyroscope data
        plt.figure(figsize=(10,5))
        plt.plot(df['timestamp'], df['gyro x'], label='Gyro X')
        plt.xlabel('Time')
        plt.ylabel('Gyro X')
        plt.title('Gyroscope X-axis Over Time')
        plt.xticks(rotation=45)
        plt.tight layout()
        plt.legend()
        plt.show()
        # Plot Y-axis Gyroscope data
        plt.figure(figsize=(10,5))
        plt.plot(df['timestamp'], df['gyro_y'], label='Gyro Y')
        plt.xlabel('Time')
        plt.ylabel('Gyro Y')
        plt.title('Gyroscope Y-axis Over Time')
        plt.xticks(rotation=45)
        plt.tight layout()
```

```
plt.legend()
plt.show()
# Plot Z-axis Gyroscope data
plt.figure(figsize=(10,5))
plt.plot(df['timestamp'], df['gyro z'], label='Gyro Z')
plt.xlabel('Time')
plt.ylabel('Gyro Z')
plt.title('Gyroscope Z-axis Over Time')
plt.xticks(rotation=45)
plt.tight layout()
plt.legend()
plt.show()
# Plot all three axes in one graph for comparison
plt.figure(figsize=(10,5))
plt.plot(df['timestamp'], df['gyro_x'], label='Gyro X', color='red')
plt.plot(df['timestamp'], df['gyro_y'], label='Gyro Y', color='green')
plt.plot(df['timestamp'], df['gyro_z'], label='Gyro Z', color='blue')
plt.xlabel('Time')
plt.ylabel('Gyroscope Values')
plt.title('Gyroscope X, Y, Z Over Time')
plt.xticks(rotation=45)
plt.tight layout()
plt.legend()
plt.show()
```

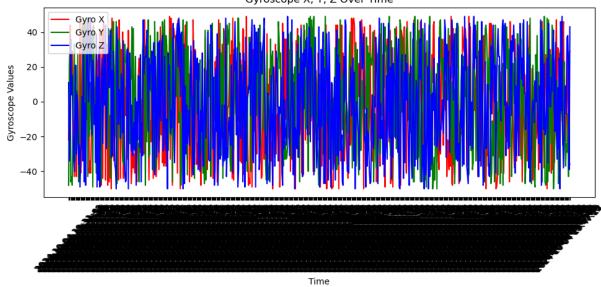












Student name: Rammaka Iddamalgoda

Student ID: s223496576

SIT225: Data Capture Technologies

Video link: https://deakin.au.panopto.com/Panopto/Pages/Viewer.aspx?id=d64650d9-319a-4ffa-96b4-b1ff009e7626

```
Py script
import serial
import time
import json
import firebase_admin
from firebase_admin import credentials, db
# Initialize Firebase with the correct credentials
cred_obj = firebase_admin.credentials.Certificate(
    r'C: \label{local-control} In the control of the 
firebase_admin.initialize_app(cred_obj, {
     'databaseURL': 'https://sit225-5-default-rtdb.firebaseio.com//'
})
# Set up Serial communication
ser = serial.Serial('COM3', 9600, timeout=1)
time.sleep(2) # Allow some time for Arduino to initialize
# Firebase database reference
ref = db.reference('mpu6050 data')
while True:
    data = ser.readline().decode('utf-8').strip()
    if data:
                # Parse the labeled data
               if "Gyro X:" in data:
                     parts = data.replace("Gyro X: ", "").replace("Y: ", "").replace("Z: ", "").split(",")
                     gyro_x, gyro_y, gyro_z = map(float, parts)
                    timestamp = time.strftime('%Y-%m-%d %H:%M:%S')
                     # Prepare data in JSON format
                     gyro_data = {
                           "timestamp": timestamp,
                           "gyro_x": gyro_x,
                           "gyro_y": gyro_y,
                           "gyro_z": gyro_z
                     # Push data to Firebase
                     ref.push(gyro_data)
                     print(f"Data sent: {gyro_data}")
```

except ValueError:

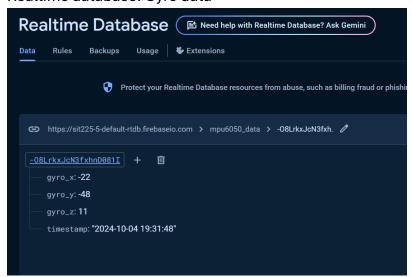
print(f"Error parsing data: {data}")

time.sleep(1) # Adjust the data rate

Arduino Sketch

```
#include <Wire.h>
#include <Adafruit_MPU6050.h>
#include <Adafruit_Sensor.h>
Adafruit_MPU6050 mpu;
void setup() {
Serial.begin(9600);
 Wire.begin();
 // Initialize the MPU6050 sensor
 if (!mpu.begin()) {
 Serial.println("Failed to find MPU6050 chip");
 while (1) {
  delay(10);
 }
// Setup the gyro and accelerometer ranges
 mpu.setGyroRange(MPU6050_RANGE_250_DEG);
mpu.setAccelerometerRange(MPU6050_RANGE_2_G);
mpu.setFilterBandwidth(MPU6050_BAND_21_HZ);
Serial.println("MPU6050 Initialized");
}
void loop() {
sensors_event_t a, g, temp;
mpu.getEvent(&a, &g, &temp);
// Print gyroscope data to the Serial Monitor
Serial.print("Gyro X: "); Serial.print(g.gyro.x);
 Serial.print(", Y: "); Serial.print(g.gyro.y);
 Serial.print(", Z: "); Serial.println(g.gyro.z);
 delay(500); // Adjust the data rate
```

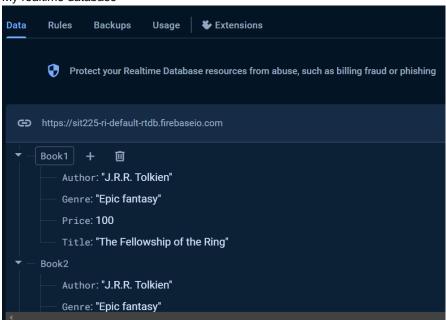
Realtime database: Gyro data

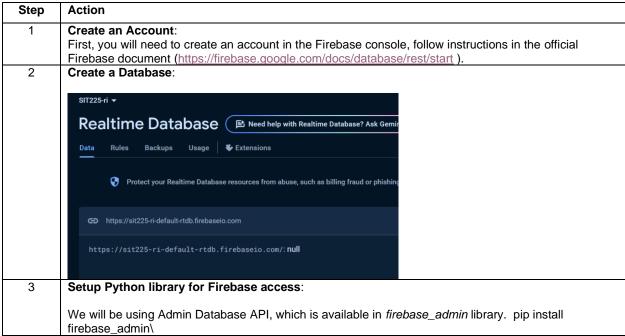


Activity 5.1: Firebase Realtime database

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.

My realtime database





```
wnloading cachecontrol-0.14.0-py3-none-any.whl (22 kB)
wnloading google_cloud_firestore-2.19.0-py2.py3-none-any.whl (336 kB)
 Downloading google_cloud_storage-2.18.2-py2.py3-none-any.whl (130 kB)
                                                                                                                                                                                    eta 0:00:00
 ownloading grpcio-1.66.2-cp311-cp311-win_amd64.whl (4.3 MB)
                                                                                                                                                                        eta 0:00:00
Downloading protobuf-4.25.5-cp310-abi3-win_amd64.wh (413 kB)

Installing collected packages: pyjwt, protobuf, grpcio, google-cr32c, google-resumable-media, us, google-cloud-core, google-cloud-croape, google-cloud-firestore, firebase_admin

Attempting uninstall: pyjwt 2.4.0

Uninstalling PyJWT-2.4.0:

Successfully uninstalled PyJWT-2.4.0

Attempting uninstall: protobuf
Found existing installation: protobuf 3.20.3

Uninstalling protobuf-3.20.3:
Successfully uninstalled PyJWT-2.4.0

Successfully uninstalled Successfully uninstalled protobuf-3.20.3

Successfully uninstalled cachecontrol-0.14.0 firebase_admin-6.5.0 google-cloud-core-2.4.1 google-cloud-storage-2.18.2 google-cr32c-1.6.0 google-resumable-media-2.7.2 grpcio-1.66.2 grpcio-2.5.5 pyjwt-2.9.0
```

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.



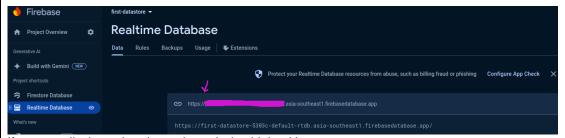
駅 sit225-5-firebase-adminsdk-0982n-f4f8bea959...

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 (https://github.com/deakin-deepdreamer/sit225/blob/main/week_5/firebase_explore.ipynb).

```
import firebase admin
   databaseURL = 'https://XXX.firebasedatabase.app/'
4
   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.

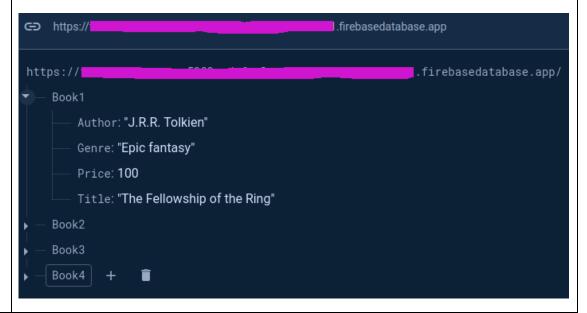
Write to database Using the set() Function:

5

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.

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.

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.



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.

8 Update data:

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.



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:
```

```
{
    "sensor_name": "DHT22",
    "timestamp": "2024-10-03 15:01:23",
    "temperature": 24.5,
    "humidity": 60
}
```

to handle multiple data values (like temperature and humidity) in the same JSON object, you can store data for multiple sensors in the database under different nodes. For example:

```
json
Copy code
{
    "DHT22": {
        "sensor_name": "DHT22",
        "timestamp": "2024-10-03 15:01:23",
        "temperature": 24.5,
        "humidity": 60
    },
    "SR04": {
        "sensor_name": "SR04",
```

```
"timestamp": "2024-10-03 15:01:23",
         "distance": 150
        }
11
12
       Question: Generate some random data for DHT22 sensor, insert data to database, query all data and
       screenshot the output here.
       Question: Generate some random data for the SR04 Ultrasonic sensor, insert data to database, query
       all data and screenshot the output here.
       Answer:
       For both 11 and 12
                        default_app = firebase_admin.initialize_app(cred_obj, {
                            'databaseURL': 'https://sit225-ri-default-rtdb.firebaseio.com/'
                       })
                   # Generate random data for DHT22 sensor
                   dht_data = {
                        "sensor name": "DHT22",
                        "timestamp": time.strftime('%Y-%m-%d %H:%M:%S'),
                        "temperature": round(random.uniform(20, 30), 2),
                        "humidity": random.randint(40, 70)
                   # Push DHT22 data to Firebase
                   ref = db.reference("/DHT22 data")
                   ref.push(dht_data)
                   # Generate random data for SR04 sensor (Ultrasonic sensor)
                   sr04 data = {
                        "sensor_name": "SR04",
                        "timestamp": time.strftime('%Y-%m-%d %H:%M:%S'),
                        "distance": random.randint(100, 200)
                   # Push SR04 data to Firebase
                   ref = db.reference("/SR04 data")
                   ref.push(sr04_data)
                   Data successfully pushed to Firebase
13
       Question: Firebase Realtime database generates events on data operations. You can refer to section
       'Handling Realtime Database events' in the document
       (https://firebase.google.com/docs/functions/database-events?gen=2nd ). 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: Firebase Realtime Database generates events on data operations (insert, update, delete). These events can be handled using listeners in Python to trigger actions when data changes.

In Python, you can use the listen() function to monitor changes in real-time, but Firebase Functions (in JavaScript) are more commonly used for handling database events directly on the server side.

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.

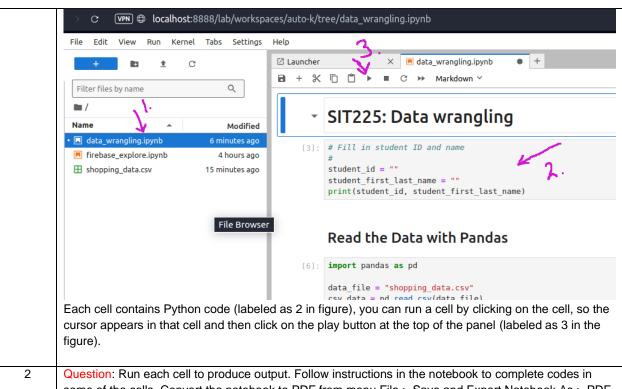
Software Required

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).



Question: Run each cell to produce output. Follow instructions in the notebook to complete codes in some of the cells. Convert the notebook to PDF from menu File > Save and Export Notebook As > PDF. Convert this activity sheet to PDF and merge with the notebook PDF.

Answer: There is no answer to write here. You have to answer in the Jupyter Notebook.

Question: Once you went through the cells in the Notebook, you now have a basic understanding of data wrangling. Pandas are a powerful tool and can be used for reading CSV data. Can you use Pandas in reading sensor CSV data that you generated earlier? Describe if any modification you think necessary?

Answer:

Yes, I can use Pandas to read the sensor data that I generated earlier. The pd.read_csv() function is useful for loading the CSV file. If there are missing or noisy values, I can handle those by dropping them. Or, I could use methods like fillna() to fill missing values with the mean or median of the data to ensure consistency.

If the CSV file has inconsistent formatting or extra columns, I might need to specify the correct delimiter or column names while loading the file with pd.read_csv(). Additionally, it may be necessary to check the data types of each column and convert them appropriately, especially for the timestamp and numeric data.

Question: What do you understand of the Notebook section called Handling Missing Value? Discuss in group and briefly summarise different missing value imputation methods and their applicability on different data conditions.

Answer:

4

The Notebook section on Handling Missing Values discusses various imputation techniques that help fill in missing data points in a dataset. The common methods include:

Mean Imputation: This method fills missing values with the mean of the available data. It is suitable for data with no major outliers or when data points are evenly distributed around the mean.

Median Imputation: This method fills missing values with the median of the dataset. It works well for skewed data or when the data has significant outliers, as the median is less affected by extreme values.

Mode Imputation: This is used for categorical data where the most frequent value (mode) is used to replace missing values.

Linear Interpolation: This is applied for time-series data where the trend and continuity are important. It estimates missing values by interpolating between known data points.

Dropping Missing Values: Sometimes, if there are only a few missing data points, it is better to drop the rows or columns with missing values, especially if their absence does not significantly affect the analysis.

These methods are applied depending on the type of data and how much missing data exists. For example, mean or median imputation is suitable for numerical data, while linear interpolation is preferred in time-series data.

```
In [11]: # Fill in student ID and name
#
    student_id = "s223496576"
    student_first_last_name = "Rammaka Iddamalgoda"
    print(student_id, student_first_last_name)
    s223496576 Rammaka Iddamalgoda
```

```
In [13]: import firebase admin
         from firebase admin import credentials, db
         # Initialize Firebase Admin SDK with the correct certificate path
         cred obj = firebase admin.credentials.Certificate(
             r'C:\Users\ramma\Downloads\sit225-ri-firebase-adminsdk-octoc-af69f7a7ab.
         # Initialize Firebase app (ensure only one initialization)
         if not firebase admin. apps:
             default app = firebase admin.initialize app(cred obj, {
                  'databaseURL': 'https://sit225-ri-default-rtdb.firebaseio.com/'
             })
         # A reference point is always needed to be set before any operation is carri
         ref = db.reference("/")
         # JSON format data (key/value pair)
         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
             }
         # JSON format data is set (overwritten) to the reference point set at /, whi
         ref.set(data)
```

```
# 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 a specific book (/Book1)
         ref = db.reference("/Book1")
         book1 = ref.get()
         print(book1)
         # Write data using push()
         ref = db.reference("/Books/Best Sellers")
         for key, value in data.items():
             ref.push().set(value)
         # Update data (reducing price for Tolkien books)
         ref = db.reference("/Books/Best Sellers/")
         best sellers = ref.get()
         for key, value in best sellers.items():
             if value["Author"] == "J.R.R. Tolkien":
                 ref.child(key).update({"Price": 80})
         # Delete all Tolkien's books from the Best Sellers list
         ref = db.reference("/Books/Best Sellers/")
         for key, value in best sellers.items():
             if value["Author"] == "J.R.R. Tolkien":
                 ref.child(key).set({})
        {'Bookl': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 10
        0, 'Title': 'The Fellowship of the Ring'}, 'Book2': {'Author': 'J.R.R. Tolki
        en', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Two Towers'}, 'Boo
        k3': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Ti
        tle': 'The Return of the King'}, 'Book4': {'Author': 'Paulo Coelho', 'Genr
        e': 'Fiction', 'Price': 100, 'Title': 'Brida'}}
        <class 'dict'>
        Bookl: {'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'}
In [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': 10
        0, 'Title': 'The Fellowship of the Ring'}, 'Book2': {'Author': 'J.R.R. Tolki
        en', 'Genre': 'Epic fantasy', 'Price': 100, 'Title': 'The Two Towers'}, 'Boo
        k3': {'Author': 'J.R.R. Tolkien', 'Genre': 'Epic fantasy', 'Price': 100, 'Ti
        tle': 'The Return of the King'}, 'Book4': {'Author': 'Paulo Coelho', 'Genr
        e': 'Fiction', 'Price': 100, 'Title': 'Brida'}, 'Books': {'Best Sellers':
        {'-08LLuZGE4Tvg7RdVbAC': {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Pri
        ce': 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'}
        Books: {'Best Sellers': {'-08LLuZGE4Tvg7RdVbAC': {'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'}
In [15]: # 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})
        {'-08LLuZGE4Tvg7RdVbAC': {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Pri
        ce': 100, 'Title': 'Brida'}}
```

```
In [16]: # Delete all best_seller data.
#
    ref = db.reference("/Books/Best_Sellers/")
    best_sellers = ref.get()
    print(best_sellers)
    print(type(best_sellers))

{'-08LLuZGE4Tvg7RdVbAC': {'Author': 'Paulo Coelho', 'Genre': 'Fiction', 'Price': 100, 'Title': 'Brida'}}
    <class 'dict'>
In []:
```

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```
In [6]: import random
        import time
        import firebase admin
        from firebase admin import credentials, db
        # Initialize Firebase Admin SDK with the correct certificate path
        cred obj = firebase admin.credentials.Certificate(
            r'C:\Users\ramma\Downloads\sit225-ri-firebase-adminsdk-octoc-af69f7a7ab.
        # Initialize Firebase app (ensure only one initialization)
        if not firebase admin. apps:
            default app = firebase admin.initialize app(cred obj, {
                'databaseURL': 'https://sit225-ri-default-rtdb.firebaseio.com/'
            })
        # Generate random data for DHT22 sensor
        dht data = {
            "sensor name": "DHT22",
            "timestamp": time.strftime('%Y-%m-%d %H:%M:%S'),
            "temperature": round(random.uniform(20, 30), 2),
            "humidity": random.randint(40, 70)
        }
        # Push DHT22 data to Firebase
        ref = db.reference("/DHT22 data")
        ref.push(dht data)
        # Generate random data for SR04 sensor (Ultrasonic sensor)
        sr04 data = {
            "sensor name": "SR04",
            "timestamp": time.strftime('%Y-%m-%d %H:%M:%S'),
            "distance": random.randint(100, 200)
        }
        # Push SR04 data to Firebase
        ref = db.reference("/SR04 data")
        ref.push(sr04 data)
```

Data successfully pushed to Firebase

In []:

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SIT225: Data wrangling

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

```
In [1]: # Fill in student ID and name
#
    student_id = ""
    student_first_last_name = ""
    print(student_id, student_first_last_name)
```

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.

```
In [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
                         Male
                                19
                                                    15
                    1
                    2
                         Male 21
                                                    15
                                                                            81
       1
                    3 Female 20
                                                    16
                                                                             6
                    4 Female
                                                                            77
       3
                                23
                                                    16
                    5 Female 31
                                                    17
                                                                            40
                           . . . . . . . . .
                   . . .
                                                   . . .
                                                                            . . .
       195
                   196 Female 35
                                                                            79
                                                   120
       196
                  197 Female 45
                                                   126
                                                                            28
       197
                                                                            74
                   198
                         Male
                                32
                                                   126
       198
                   199
                         Male
                                32
                                                   137
                                                                            18
                   200
                         Male
                                30
                                                   137
                                                                            83
       199
       [200 rows x 5 columns]
                                                      Spending Score (1-100)
          CustomerID Genre Age Annual Income (k$)
                      Male 19
      0
                  1
                                                                          39
       1
                       Male
                              21
                                                   15
                                                                          81
                  3 Female 20
                                                  16
                                                                           6
                  4 Female
                                                                          77
       3
                              23
                                                  16
                  5 Female
                              31
                                                  17
                                                                          40
```

Access the Column

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

```
In [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
       198
              32
       199
              30
       Name: Age, Length: 200, dtype: int64
```

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.

```
In [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 the column "Age" at the first row to the column to the column "Age" at the first row to the column to th
```

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

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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

```
In [5]:
        print("Shows data to 5th to less than 10th in a row:")
        print(csv data.iloc[5:10])
       Shows data to 5th to less than 10th in a row:
         CustomerID Genre Age Annual Income (k$)
                                                      Spending Score (1-100)
       5
                  6 Female
                              22
                                                  17
                                                                          76
                  7 Female
                              35
                                                  18
                                                                           6
      7
                  8 Female
                                                  18
                                                                          94
                              23
                  9
                                                  19
                                                                           3
       8
                       Male
                              64
      9
                 10 Female
                              30
                                                  19
                                                                          72
```

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.

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

```
200 200.000000
               200.000000
                                                        200.000000
       count
                                2
       unique
                      NaN
                                           NaN
                                                               NaN
       top
                      NaN Female
                                           NaN
                                                               NaN
       freq
                      NaN
                              112
                                           NaN
                                                               NaN
               100.500000
                              NaN
                                    38.850000
                                                         60.560000
       mean
       std
               57.879185
                              NaN
                                    13.969007
                                                         26.264721
                              NaN
       min
                 1.000000
                                     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
               200.000000
                              NaN
                                    70.000000
                                                        137.000000
       max
               Spending Score (1-100)
       count
                           200.000000
       unique
                                  NaN
                                  NaN
       top
                                  NaN
       freq
                            50.200000
       mean
       std
                            25.823522
       min
                             1.000000
       25%
                            34.750000
       50%
                            50.000000
       75%
                            73.000000
       max
                            99.000000
        Handling Missing Value
In [7]: # For the first step, we will figure out if there is missing value.
        print(csv data.isnull().values.any())
        print()
       False
In [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
                                                         Spending Score (1-100)
                       Genre
                               Age
                                    Annual Income (k$)
       0
                   1
                        Male 19.0
                                                   15.0
                                                                           39.0
                   2
                        Male
                               NaN
                                                   15.0
                                                                           81.0
       1
       2
                   3 Female 20.0
                                                    NaN
                                                                             6.0
       3
                   4
                      Female 23.0
                                                                           77.0
                                                   16.0
                      Female 31.0
                                                   17.0
                                                                            NaN
       Missing? True
In [ ]:
```

Age

Annual Income (k\$)

CustomerID

Genre

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)

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

print(data missing.mean())

Question: This code will generate error. Can you explain why and how it can be solved?

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

Answer:

The error occurs because the mean() function in Pandas is attempting to calculate the mean for columns that contain non-numeric data, specifically the 'Genre' column, which contains strings such as 'Male' and 'Female'. Since these values are not numeric, Pandas cannot compute a mean for this column, leading to the TypeError.

Solution

To solve this issue, I think dropping the column before applying would work, but in the case that non numeric data is also required it can just be excluded from

```
In [10]: # 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
                        Age Annual Income (k$) Spending Score (1-100)
                   1 19.0
                                           15.0
                                                                   39.0
                    2
                                           15.0
                       NaN
                                                                   81.0
        1
        2
                    3 20.0
                                            NaN
                                                                    6.0
        3
                    4 23.0
                                           16.0
                                                                   77.0
                    5 31.0
                                           17.0
                                                                    NaN
In [11]: print(data missing wo genre.mean())
        CustomerID
                                  100.500000
        Age
                                   38.939698
        Annual Income (k$)
                                   61.005051
        Spending Score (1-100)
                                   50.489899
        dtype: float64
In [12]: 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! :
               Age Annual Income (k$)
                                         Spending Score (1-100)
   CustomerID
0
            1 19.0
                                   15.0
                                                           39.0
1
            2
               NaN
                                   15.0
                                                           81.0
            3
              20.0
2
                                   NaN
                                                            6.0
3
            4
              23.0
                                   16.0
                                                           77.0
            5
              31.0
4
                                   17.0
                                                            NaN
5
            6
              22.0
                                    NaN
                                                           76.0
            7
6
              35.0
                                   18.0
                                                            6.0
7
            8
              23.0
                                   18.0
                                                           94.0
8
            9 64.0
                                   19.0
                                                            NaN
9
           10 30.0
                                   19.0
                                                           72.0
Dataset that has been processed Handling Missing Values with Mean :
   CustomerID
                     Age Annual Income (k$)
                                             Spending Score (1-100)
            1 19.000000
0
                                   15.000000
                                                           39.000000
            2
              38.939698
                                   15.000000
                                                           81.000000
1
2
            3
              20.000000
                                   61.005051
                                                            6.000000
3
              23.000000
                                   16.000000
                                                           77.000000
            5 31.000000
                                                           50.489899
4
                                   17.000000
5
            6
              22.000000
                                   61.005051
                                                           76.000000
           7
6
              35.000000
                                   18.000000
                                                            6.000000
7
           8 23.000000
                                   18.000000
                                                           94.000000
8
           9
              64.000000
                                   19.000000
                                                           50.489899
           10 30.000000
                                   19.000000
                                                           72.000000
```

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.

```
In [13]: 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$)
                           62.0
Spending Score (1-100)
                           50.0
dtype: float64
Dataset with empty values! :
                Age Annual Income (k$) Spending Score (1-100)
   CustomerID
0
            1
               19.0
                                   15.0
                                                            39.0
1
            2
                NaN
                                   15.0
                                                            81.0
2
            3
               20.0
                                                             6.0
                                    NaN
3
            4
               23.0
                                   16.0
                                                            77.0
            5
4
               31.0
                                   17.0
                                                             NaN
5
            6
               22.0
                                    NaN
                                                            76.0
6
            7
                                                             6.0
               35.0
                                   18.0
7
            8
               23.0
                                   18.0
                                                            94.0
8
            9
              64.0
                                   19.0
                                                             NaN
9
           10
               30.0
                                   19.0
                                                            72.0
Dataset that has been processed Handling Missing Values with Median :
                                         Spending Score (1-100)
   CustomerID
                Age Annual Income (k$)
0
            1
               19.0
                                   15.0
                                                            39.0
1
            2
               36.0
                                   15.0
                                                            81.0
2
            3
               20.0
                                   62.0
                                                             6.0
3
            4
               23.0
                                   16.0
                                                            77.0
4
            5
               31.0
                                   17.0
                                                            50.0
5
               22.0
                                   62.0
                                                            76.0
            6
            7
                                                             6.0
6
               35.0
                                   18.0
7
            8
               23.0
                                   18.0
                                                            94.0
8
            9
               64.0
                                   19.0
                                                            50.0
9
           10 30.0
                                   19.0
                                                            72.0
```

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In []: