Objective:

This Project involves applying supervised learning techniques, specifically **Logistic Regression** and **Support Vector Machines** (SVM), to the C-stick Fall Prediction Dataset. You can read about the dataset here. You will then perform predictions On the data using each model, then finally you will compare the performance of these two algorithms and interpret their efficacy in making predictions based on the data.

Steps

This Project is divided into 2 parts: **Section 1** and **Section 2**. Section 1 is implemented for you, and it involves loading the data, loading the necessary libraries and the separation of the data in features (inputs/attributes/Predictor variables) and labels (output/target/Predicted variables). You will implement Section 2, and it involves the following tasks:

- 1. Split the data into training sets and test sets
- 2. Implement the SVM Model
- 3. Implement predictions on the SVM model using an accuracy metric
- 4. Implement the Log Reg Model
- 5. Implement predictions on the Log Reg Model an accuracy metric

```
# -*- coding: utf-8 -*-
"""fall-prediction-using-ml.ipynb
Automatically generated by Colab.
Original file is located at
    https://colab.research.google.com/drive/1T9CbfJCYEVTJ E26h07H9SQtujtMeh4V
#Training the model:
# Import necessary libraries
import numpy as np
import pandas as pd
from sklearn.model selection import train test split
from sklearn.svm import SVC
from sklearn.metrics import accuracy score
df Cstick = pd.read csv("/content/cStick.csv")
print(df Cstick)
'''Output:
                                    Sugar level
                                                         Accelerometer
                                                                          Decision
      Distance Pressure
                              HRV
                                                   Sp02
        25.540
                     1.0 101.396
                                         61.080
                                                 87.770
                                                                              1
                                                                    1.0
         2.595
                     2.0 110.190
                                         20.207
                                                 65.190
                                                                    1.0
        68.067
                     0.0
                           87.412
                                         79.345
                                                 99.345
                                                                    0.0
        13.090
                     1.0
                           92.266
                                         36.180
                                                 81.545
                                                                              1
                                                                    1.0
        69.430
                     0.0
                           89.480
                                         80.000
                                                 99.990
                                                                    0.0
                                                                              0
```

```
2034
       5.655
                 2.0 116.310
                                162.242
                                        71.310
                                                       1.0
                                                               2
2035
      9.660
                 2.0 124.320
                                177.995 79.320
                                                       1.0
2036
     15.220
                                                               1
                 1.0 93.828
                                40.440 82.610
                                                       1.0
2037
      9.120
                 2.0 123.240
                                175.871 78.240
                                                       1.0
                                                               2
                                                               0
2038
      62.441
                 0.0 78.876
                                76.435 96.435
                                                       0.0
[2039 rows x 7 columns]
# Separate features and labels
features = df_Cstick[['Distance', 'Pressure', 'HRV', 'Sugar level', 'SpO2',
'Accelerometer']]
print(features)
'''Output:
     Distance Pressure
                         HRV Sugar level Sp02 Accelerometer
      25.540
                 1.0 101.396
                                 61.080 87.770
                                                       1.0
0
      2.595
                 2.0 110.190
                                 20.207 65.190
                                                       1.0
2
      68.067
                 0.0 87.412
                                 79.345 99.345
                                                       0.0
                 1.0 92.266
                                 36.180 81.545
                                                       1.0
      13.090
4
      69.430
                 0.0 89.480
                                80.000 99.990
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2034
      5.655
                 2.0 116.310
                                162.242 71.310
                                                       1.0
2035
      9.660
                 2.0 124.320
                                177.995 79.320
                                                       1.0
2036
      15.220
                 1.0 93.828
                                40.440 82.610
                                                       1.0
2037
      9.120
                 2.0 123.240
                                175.871 78.240
                                                       1.0
2038
      62.441
                                76.435 96.435
                                                       0.0
                 0.0 78.876
[2039 rows x 6 columns]
df Cstick = df Cstick.rename(columns={"Decision": "Decision"})
labels = df Cstick['Decision']
# Split the data into training and testing sets
X train, X test, y train, y test = train test split(features, labels, test size=0.2,
random state=42)
# Create and fit SVM model
svm model = SVC()
svm_model.fit(X_train, y_train)
# Predict on the test set
```

```
y pred = svm model.predict(X test)
# Calculate accuracy
accuracy = accuracy score(y test, y pred)
print('The accuracy of the SVM Model is:', accuracy)
Output:
The accuracy of the SVM Model is: 1.0
#Testing using the svm model
import joblib
# Save the trained SVM model to a .pkl file
joblib.dump(svm model, 'svm model.pkl')
from google.colab import files
files.download('svm model.pkl')
svm model = joblib.load('/content/svm model.pkl')
# Define a function to predict fall based on sensor data
def predict fall(sensor data):
   # Make prediction using the trained SVM model
   prediction = svm model.predict([sensor data])[0]
   if prediction == 2:
       return 'Take care of your child/patient.'
   else:
       return 'No significant abnormal variations detected.'
# Generate some random sensor data for testing
random sensor data = [2.6, 3, 111, 20.207, 66, 1.0] # Replace with your random sensor
data
# Test the model with the random sensor data
prediction = predict fall(random sensor data)
print(prediction)
```

```
Output:
```

Take care of your child/patient. /usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have

valid feature names, but SVC was fitted with feature names

warnings.warn(