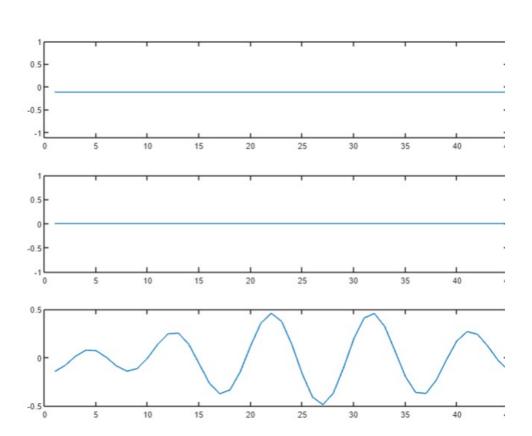
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Android App Classification

30.09.2021

18EE35003 Divyansh Singh

Outline

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- Using the DFT data of several readings train the SVM to classify input data
- Test the SVM over more input data
- Provide the user with output classes, viz,x-axis motion, y-axis motion, z-axis motion

Improvement

The task is to classify whether the data represents motion along x-axis, y-axis, z-axis or circular motion (on a horizontal surface).

The key observation

During motion along a single axis, the data from other axes should represent a near still waveform. While during circular motion the data from x and y directions should indicate oscillations while the data from the third direction would be more or less still.

Thus, all we need to do is train a SVM to classify whether there is oscillatory motion along one axis or not.

<u>Advantages</u>

- More data (triple) to train the SVM
- Faster training
- Faster runtime

<u>AIM</u>

Aim is to classify the type of motion via individual classification along each axis. The idea is to sample certain (here, 40 - found via successive observations) acceleration values and process them to classify the type of motion

Getting the data



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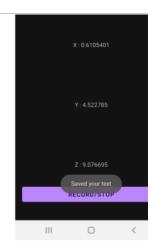
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CODE:

```
package com.example.firstapp;
import androidx.annotation.RequiresApi;
import androidx.appcompat.app.AppCompatActivity;
import android.content.Context;
import android.content.Intent;
import android.hardware.Sensor;
import android.hardware.SensorEvent;
import android.hardware.SensorEventListener;
import android.hardware.SensorManager;
import android.os.Build;
import android.os.Bundle;
import android.os.Environment;
import android.util.Log;
import android.view.View;
import android.widget.EditText;
import android.widget.TextView;
import android.widget.Toast;
import java.io.File;
import java.io.FileNotFoundException;
import java.io.FileOutputStream;
import java.io.FileWriter;
import java.io.IOException;
import java.io.OutputStreamWriter;
import java.nio.charset.StandardCharsets;
public class MainActivity extends AppCompatActivity
implements SensorEventListener {
  public static final String EXTRA MESSAGE =
"com.example.myfirstapp.MESSAGE";
  private static final String TAG = "MainActivity";
  private Sensor accelero;
  private SensorManager SM;
  private TextView xtext, ytext, ztext;
  private int record_status;
  String fileName="sensordata.txt";
  String baseDir =
Environment.getExternalStorageDirectory().getAbsolutePath();
  String pathDir = baseDir +
"/Android/data/com.mypackage.firstApp/";
  File file:
   File gpxfile;
   FileWriter writer;
  private int t;
  private OutputStreamWriter outputWriter;
```

NET - (DELIBOTERATIONET)



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```
getSystemService(Context.SENSOR SERVICE);
//
       accelero =
SM.getDefaultSensor(Sensor.TYPE_LINEAR_ACCELERATION);
      accelero =
SM.getDefaultSensor(Sensor.TYPE ACCELEROMETER);
       SM.registerListener(this, accelero,
SensorManager. SENSOR DELAY NORMAL);
       Log.d(TAG, "onCreate: registered accelerometer
listener");
      xtext = (TextView) findViewById(R.id.xtext);
       ytext = (TextView)findViewById(R.id.ytext);
       ztext = (TextView)findViewById(R.id.ztext);
       record status = 0;
       t = 0:
       file=new
File(MainActivity.this.getFilesDir(),"Kalman");
       if (!file.exists()) {
           file.mkdir():
   }
   @Override
  public void onSensorChanged(SensorEvent event) {
      xtext.setText("X : " + event.values[0]);
       ytext.setText("Y : " + event.values[1]);
       ztext.setText("Z : " + event.values[2]);
       if(record status == 1) {
           String str = event.values[0] + "," +
event.values[1] + "," + event.values[2] + "," + ++t + "\n";
           try {
               writer.append(str);
               writer.flush();
           } catch (Exception e) {
               e.printStackTrace();
  public void onAccuracyChanged(Sensor sensor, int i) {
  public void send message(View view) {
       Intent intent = new Intent(this,
DisplayMessageActivity.class);
       EditText editText = (EditText)
findViewById(R.id.editTextTextPersonName);
       String message = editText.getText().toString();
       intent.putExtra(EXTRA_MESSAGE, message);
       startActivity(intent);
  public void do record stop(View view) {
       EditText editText = (EditText)
findViewById(R.id.editTextTextPersonName);
       if(record_status%2 == 0 ) {
           editText.setText("Recording");
           try {
               gpxfile=new File(file, "sensordata.txt");
               writer=new FileWriter(gpxfile);
           } catch(Exception e) {
               editText.setText(e.getMessage());
               e.printStackTrace();
```



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```
Toast.LENGTH_SHORT) .show();
           } catch (Exception e) {
               editText.setText(e.getMessage());
               e.printStackTrace();
           record_status = 2;
   }
}
```

Processing the data in MATLAB to observe usability of the same

To save the data as csv for analysis in other apps, convert to a matfile for further use

```
d = load('long1.txt');
writematrix(d, 'd.csv')
T=readtable('d.csv');
a=T\{:,1:3\};
t=T\{:,4\};
save('testmat.mat','a','t')
```

To run the Kalman filter and save the data (in multiples of 40)

```
load testmat.mat
% Initial Guess
% state
x=randn(9,1);
% Covariance
P=eye(9);
% Process Noise covariance Q
q = 0.0001;
Q= q*eye(9);
% Measurement Noise covariance R
r = 0.01;
R= r*eye(3);
X=[]; Z=[]; Gain=[]; Err=[];
N=length(t);
% Construct H matrix
  H=[zeros(3,6) eye(3)];
for n=1:N-1
  h=t(n+1)-t(n);
  h2=h^2/2;
  % Construct Phi matrix
  phi=[eye(3) 	 h*eye(3) 	 h2*eye(3)
          zeros(3) eye(3)
h*eye(3)
          zeros(3) zeros(3)
 eye(3)];
  % Compute the Kalman Gain K
  K=P*H'*inv(H*P*H'+R);
  \mbox{\%} Update the states
  z=a(n,:)';
  err=(z-H*x);
  x=x+K*err;
```



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```
% Project Ahead
  x=phi*x;
  P=phi*P*phi'+Q;
ae=[X(:,7:9);a(end,:)];
subplot (311)
plot([a(:,1) ae(:,1)])
ylabel("a x")
legend("Sampled Data", "Filtered
Data", Location="best")
grid on
title("Q = " + q + " R = " + r)
subplot(312)
plot([a(:,2) ae(:,2)])
ylabel("a_y")
legend("Sampled Data", "Filtered
Data", Location="best")
grid on
subplot(313)
plot([a(:,3)-9.8 ae(:,3)-9.8])
plot([a(:,3) ae(:,3)])
ylabel("a z")
legend("Sampled Data", "Filtered
Data", Location="best")
grid on
figure()
plot([ae(:,1) ae(:,2) ae(:,3)-9.8])
plot([ae(:,1) ae(:,2) ae(:,3)])
legend()
shg
ae = ae(1:40*floor(length(ae)/40), :);
save('otput.mat','ae')
x = ae(1:40, :);
save('unopoint.mat', 'x')
```

THE KEY POINT IS TO OBSERVE THE **DIFFERENCE FIRST (WHETHER THE DATA IS** THEORETICALLY SUFFICIENT TO CLASSIFY OR NOT)

I observed such distinctions produced by:

- Taking fourier transform of accelerometer data along each axis
- Derive the magnitudes of fourier coefficients using log operation
- Threshold the Fourier Space data using magnitude as criteria
- Taking the Inverse fourier transform

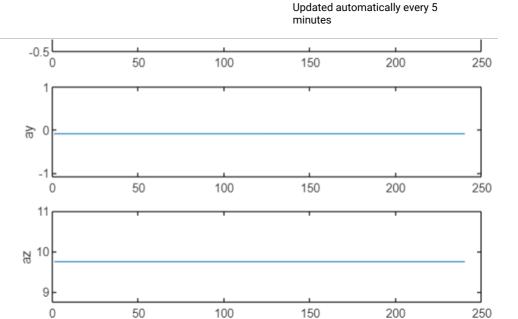
For x-axis motion:

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CODE:

```
load otput.mat
thresh = 1;
x = ae(:, 1);
y = fft(x);
mag = real(log(y));
%y (mag \sim max (mag (1:length (mag)))) =
y(mag < thresh) = 0;
ne = ifft(y);
subplot(311)
plot(ne)
ylabel('ax')
x = ae(:, 2);
y = fft(x);
mag = real(log(y));
%y(mag \sim = max(mag(1:length(mag)))) =
0;
y(mag < thresh) = 0;
ne = ifft(y);
subplot(312)
plot(ne)
ylabel('ay')
x = ae(:, 3);
y = fft(x);
mag = real(log(y));
%y(mag ~= max(mag(1:length(mag)))) =
0;
y(mag < thresh) = 0;
ne = ifft(y);
subplot (313)
plot(ne)
ylabel('az')
%save('p.mat', 'ne')
```

KEY INFERENCE:

- The Fourier space data is sufficient for proper classification
- Only the magnitude of Fourier space data can be used for complete classification

Now the data is sorted in lengths of 40



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```
is z = -1;
t = ae(:, 1);
z = reshape(t, [40]
floor(length(t)/40)]);
z = transpose(z);
%z(1, :)
%z(2, :)
T = table(z);
a = rowfun(@fft, T);
outa = 1:height(a);
outa = outa*0 + is x;
a = a\{:, :\};
a = log(a);
a = real(a);
t = ae(:, 2);
z = reshape(t, [40]
floor(length(t)/40)]);
z = transpose(z);
T = table(z);
b = rowfun(@fft, T);
outb = 1:height(b);
outb = outb*0 + is y;
b = b\{:, :\};
b = log(b);
b = real(b);
t = ae(:, 3);
z = reshape(t, [40]
floor(length(t)/40)]);
z = transpose(z);
T = table(z);
c = rowfun(@fft, T);
outc = 1:height(c);
outc = outc*0 + is_z;
c = c\{:,:\};
c = log(c);
c = real(c);
out = transpose([outa outb outc]);
input_data = [a ; b ; c];
save('p.mat', 'input data', 'out')
```

Next the SVM is designed and solved using quadprog

```
%% Design SVM
H = zeros(n,n);
for i=1:n
   for j=i:n
       H(i,j) =
y(i)*y(j)*x(:,i)'*x(:,j);
       H(j,i) = H(i,j);
   end
end
f = -ones(n, 1);
Aeq=y;
beq=0;
lb=zeros(n,1);
ub=C*ones(n,1);
Alg{1}='trust-region-reflective';
Alg{2}='interior-point-convex';
options=optimset('Algorithm', Alg{2},...
   'Display', 'off'
   'MaxIter',20);
alpha=quadprog(H, f, [], [], Aeq, beq, lb, ub,
[],options)';
AlmostZero=(abs(alpha)
<max(abs(alpha))/1e5);
alpha(AlmostZero)=0;
```

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