In the name of God

Blind Source Separation

HW #5

Maryam Riazi

810197518

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Introduction

```
TrainData_class1 = zeromean(TrainData_class1);
TrainData_class2 = zeromean(TrainData_class2);
TestData = zeromean(TestData);
```

Q1)

```
trainC1_size = size(TrainData_class1);
trainC2 size = size(TrainData class2);
Rx1 = zeros(trainC1_size(1),trainC1_size(1));
Rx2 = zeros(trainC2 size(1),trainC2 size(1));
%For loop over over each Subject %Class1
for k = 1:trainC1 size(3)
    X = TrainData_class1(:,:,k);
    Rx1 = Rx1+X*X.':
end
%For loop over over each Subject %Class2
for k = 1:trainC2 size(3)
    X = TrainData_class2(:,:,k);
    Rx2 = Rx2+X*X.';
end
Rx1 = (1/trainC1 size(3))*Rx1;
Rx2 = (1/trainC2_size(3))*Rx2;
[W, lambda] = eig(Rx1, Rx2);
[~,series] = sort(diag(lambda), 'descend');
```

```
%Sorting from the highest eigenvalue to the lowest
W = W(:,series);
```

Normalize Column wise:

Here I am not sure about the function because without normalizing the variances are more different. However as it is mentioned in the Question, I normalized W column wise...

Here if the Questions order was normalizing we should use the normalize function but I am not sure about that and I think It means only rescaling!

```
% W = normalize(W,2);
W1 = W(:,1);
W30 = W(:,30);
% Normalizing if it means scaling
W1 = W1/max(W1);
W30 = W30/max(W30);
filtered1 = [W1.';W30.']*TrainData_class1(:,:,49);
fprintf('var{wT1.X1(49)} = %f',var(filtered1(1,:)));
```

```
var{wT1.X1(49)} = 10.601210

fprintf('var{wT30.X1(49)} = %f', var(filtered1(2,:)));
```

```
var\{wT30.X1(49)\} = 1.922711
```

```
figure
tiledlayout(1,2)
nexttile
plot(filtered1(1,:),'color','red')
xlim([0,256])
ylim([-50,50])
xlabel('Time')
ylabel('w_{1}^T.X_{49}^{(1)}')

nexttile
plot(filtered1(2,:),'green')
xlim([0,256])
ylim([-50,50])
xlabel('Time')
ylabel('w_{30}^T.X_{49}^{(1)}')
```

```
50
                                              50
40
                                              40
30
                                              30
20
                                              20
 10
                                              10
                                               0
-10
                                             -10
-20
                                             -20
-30
                                             -30
-40
                                             -40
                                             -50
0
-50 L
              100
                          200
                                                            100
                                                                        200
                                                               Time
                 Time
```

```
filtered2 = [W1.';W30.']*TrainData_class2(:,:,49);
fprintf('var(wT1.X2(49)) = %f',var(filtered2(1,:)));
```

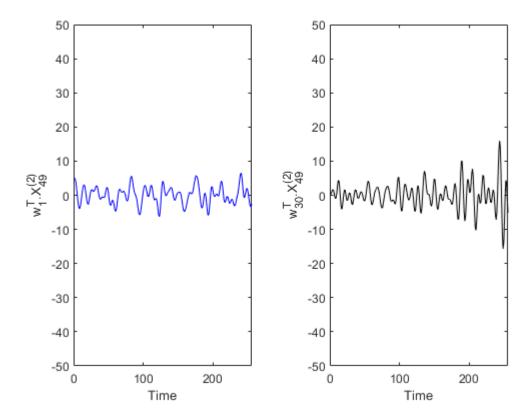
var(wT1.X2(49)) = 6.550035

```
fprintf('var(wT30.X2(49)) = %f',var(filtered2(2,:)));
```

var(wT30.X2(49)) = 14.569599

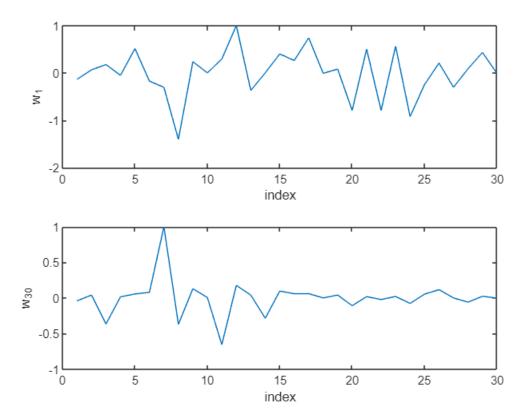
```
figure
tiledlayout(1,2)
nexttile
plot(filtered2(1,:),'color','blue')
xlim([0,256])
ylim([-50,50])
xlabel('Time')
ylabel('w_{1}^T.X_{49}^{(2)}')

nexttile
plot(filtered2(2,:),'color','black')
xlim([0,256])
ylim([-50,50])
xlabel('Time')
ylabel('w_{30}^T.X_{49}^{(2)}')
```



Q2)

```
figure
tiledlayout(2,1)
nexttile
plot(W1)
xlabel('index')
ylabel('w_1')
nexttile
plot(W30)
xlabel('index')
ylabel('index')
```



As it is shown above, During task1 the part of the brain to which sensor 8 is connected, is the most active. Similarly the most active part of the brain when doing task 2 is the one connected to sensor 7.

Q3)

```
W_14channels = [W(:,1:7) W(:,24:30)].';

train_set1 = cspfilter(W_14channels,TrainData_class1);
train_set2 = cspfilter(W_14channels,TrainData_class2);

mean1 = mean(train_set1).';
mean2 = mean(train_set2).';

cov1 = cov(train_set1);
cov2 = cov(train_set2);

[WLDA,landa] = eig((mean1-mean2)*(mean1-mean2).',cov1+cov2);
%Sorting from the highest eigenvalue to the lowest
[~,seriesLDA] = sort(diag(landa),'descend');
WLDA = WLDA(:,seriesLDA(1))
```

WLDA = 14×1 1.9530 312.7866 -1.3676 157.5076

```
199.7564

47.2718

-53.6375

-342.8187

:

% WLDA = normalize(WLDA)

WLDA=WLDA/max(WLDA);

C = (WLDA.'*mean1 + WLDA.'*mean2)/2

c = 0.0074
```

Q4)

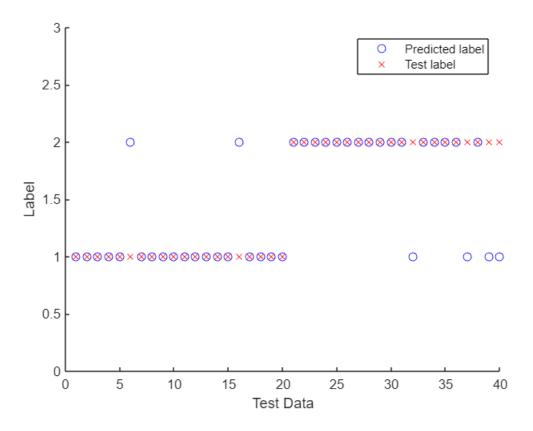
-4.4032 158.3735

Q5)

```
[Accuracy, Sum_of_correct_predictions] = acc(prediction,TestLabel)
```

```
Accuracy = 85
Sum_of_correct_predictions = 34
```

```
figure
hold on
scatter(1:length(prediction), prediction, 'Blue')
scatter(1:length(TestLabel), TestLabel, 'red', 'x')
ylim([0,3])
ylabel('Label')
xlabel('Test Data')
legend('Predicted label', 'Test label', 'Location', 'northeast')
hold off
```



Functions.

```
function zm_data = zeromean(data)
data_size = size(data);
zm_data = zeros(data_size);
for k = 1:data_size(3)
    avg = mean(data(:,:,k),2);
    zm_{data}(:,:,k) = data(:,:,k)-avg;
end
end
function filtered_set = cspfilter(W,Data)
data_size = size(Data);
filtered_set = zeros(data_size(3),size(W,1));
for k = 1:data_size(3)
    filtered = W*Data(:,:,k);
    avg = mean(filtered,2);
    variance = sum((filtered-avg).^2,2)/data_size(2);
    filtered_set(k,:) = variance.';
end
end
function label = predict(test,WLDA,c)
test_size = size(test);
label = zeros(1,test_size(1));
```

```
for i = 1:test_size(1)
   X = test(i,:).';
    if WLDA.'*X > c
        label(i) = 1;
    else
        label(i) = 2;
    end
end
end
function [accuracy, sumofcorrect] = acc(prediction, TestLabel)
sumofcorrect = 0;
for i = 1:length(TestLabel)
if TestLabel(i) == prediction(i)
    sumofcorrect = sumofcorrect+1;
end
accuracy = sumofcorrect/length(TestLabel)*100;
end
end
```