

Advanced Human-System Interfaces Third assignment

Add the data sets and the modify the paths to run the assignment, the assignment is working completely fine and correct.

1.Data Loading,

2. Task Segmentation

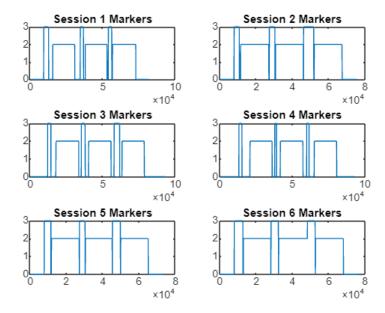
The event markers in the new dataset are standardized to align with the task definitions:

```
disp('Third Assignment')
```

Third Assignment

```
% Load the task datasets for each session
 %Paths of the datasets, Change the paths for all path in this code in your
 % path to coded to work professseur
 data1 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session1_Shimmer_5E8A.mat');
 data2 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session2 Shimmer 5E8A.mat');
data3 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session3 Shimmer 5E8A.mat');
 data4 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session4_Shimmer_5E8A.mat');
 data5 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session5 Shimmer 8965.mat');
 data6 = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Session6_Shimmer_8965.mat');
 % Function to standardize markers
 function standardizedMarkers = standardizeMarkers(markers, session)
     standardizedMarkers = zeros(size(markers));
     switch session
         case {1, 2}
             for i = 1:length(markers)
                 switch markers(i)
                     case 1
                         standardizedMarkers(i) = 3; % MathCalculus=3
                         standardizedMarkers(i) = 0; % BaseLine=0
                     case 4
                         standardizedMarkers(i) = 2; % AudioListening=2
                 end
             end
         case {3, 4}
             for i = 1:length(markers)
```

```
switch markers(i)
                    case 16
                        standardizedMarkers(i) = 2; % Listening Audio=2
                    case 4
                        standardizedMarkers(i) = 0; % Baseline=0
                    case 8
                        standardizedMarkers(i) = 3; % Math=3
                end
            end
        case {5, 6}
            for i = 1:length(markers)
                switch markers(i)
                    case 1
                        standardizedMarkers(i) = 2; % Audio=2
                    case 4
                        standardizedMarkers(i) = 0; % Baseline=0
                    case 8
                        standardizedMarkers(i) = 3; % CognLoad=3
                end
            end
    end
end
% Standardize markers for each session
data1.markers = standardizeMarkers(data1.Shimmer_5E8A_Event_Marker_CAL, 1);
data2.markers = standardizeMarkers(data2.Shimmer 5E8A Event Marker CAL, 2);
data3.markers = standardizeMarkers(data3.Shimmer_5E8A_Event_Marker_CAL, 3);
data4.markers = standardizeMarkers(data4.Shimmer 5E8A Event Marker CAL, 4);
data5.markers = standardizeMarkers(data5.Shimmer 8965 Event Marker CAL, 5);
data6.markers = standardizeMarkers(data6.Shimmer 8965 Event Marker CAL, 6);
% Visualize the event markers for validation
figure;
subplot(3, 2, 1); plot(data1.markers); title('Session 1 Markers');
subplot(3, 2, 2); plot(data2.markers); title('Session 2 Markers');
subplot(3, 2, 3); plot(data3.markers); title('Session 3 Markers');
subplot(3, 2, 4); plot(data4.markers); title('Session 4 Markers');
subplot(3, 2, 5); plot(data5.markers); title('Session 5 Markers');
subplot(3, 2, 6); plot(data6.markers); title('Session 6 Markers');
```



```
% Initialize the combined struct array with all fields to avoid dissimilar
structures
CLData = repmat(struct('GSRraw', [], 'PPGraw', [], 'GSRPPGmarker', []), 6, 1);
% Fill in the data for the sessions
CLData(1).GSRraw = data1.Shimmer_5E8A_GSR_Skin_Conductance_CAL;
CLData(1).PPGraw = data1.Shimmer_5E8A_PPG_A13_CAL;
CLData(1).GSRPPGmarker = data1.markers;
CLData(2).GSRraw = data2.Shimmer_5E8A_GSR_Skin_Conductance_CAL;
CLData(2).PPGraw = data2.Shimmer 5E8A PPG A13 CAL;
CLData(2).GSRPPGmarker = data2.markers;
CLData(3).GSRraw = data3.Shimmer 5E8A GSR Skin Conductance CAL;
CLData(3).PPGraw = data3.Shimmer_5E8A_PPG_A13_CAL;
CLData(3).GSRPPGmarker = data3.markers;
CLData(4).GSRraw = data4.Shimmer_5E8A_GSR_Skin_Conductance_CAL;
CLData(4).PPGraw = data4.Shimmer_5E8A_PPG_A13_CAL;
CLData(4).GSRPPGmarker = data4.markers;
CLData(5).GSRraw = data5.Shimmer_8965_GSR_Skin_Conductance_CAL;
CLData(5).PPGraw = data5.Shimmer_8965_PPG_A13_CAL;
CLData(5).GSRPPGmarker = data5.markers;
CLData(6).GSRraw = data6.Shimmer_8965_GSR_Skin_Conductance_CAL;
```

```
CLData(6).PPGraw = data6.Shimmer_8965_PPG_A13_CAL;
CLData(6).GSRPPGmarker = data6.markers;

% Display the combined struct to verify
disp(CLData);

6x1 struct array with fields:

    GSRraw
    PPGraw
    GSRPPGmarker

% Save the combined sessions
% change the path if you want to run it in you pc
save('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/CLDatav4.mat', 'CLData');
```

Signal Preprocessing

Each subject's signals are processed to remove noise and normalize the amplitude:

```
% Load the combined dataset
%change the path if you want to run it in you pc
load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/CLDatav4.mat');
% Define the filter functions
 Fs = 128; % Sampling frequency
 function GSRfilt = filterSegment GSR(GSRraw)
     Fs = 128; % Sampling frequency
     [b, a] = butter(2, 0.5 / (Fs / 2), 'low');
     GSRfilt = filtfilt(b, a, GSRraw);
 end
 function PPGfilt = filterSegment_PPG(PPGraw)
     Fs = 128; % Sampling frequency
     [b, a] = butter(2, 3 / (Fs / 2), 'low');
     PPGfilt = filtfilt(b, a, PPGraw);
 end
 % Preprocess and normalize signals
```

```
for subject = 1:length(CLData)
    disp(subject)
CLData(subject).GSRfilt = filterSegment_GSR(CLData(subject).GSRraw);
CLData(subject).PPGfilt = filterSegment_PPG(CLData(subject).PPGraw);

% Signal amplitude normalization using z-score
CLData(subject).GSRnorm = zscore(CLData(subject).GSRfilt);
CLData(subject).PPGnorm = zscore(CLData(subject).PPGfilt);
end

1
2
3
4
5
```

Feature Extraction

Features are extracted from the preprocessed signals for both PPG and GSR signals:

```
function features = extractGSRFeatures(signal)
    if isempty(signal)
        features = NaN(1, 10); % Return NaNs for empty signals
        return;
    end
    max_gsr = max(signal);
    min_gsr = min(signal);
   mean_gsr = mean(signal);
   var_gsr = var(signal);
    phasic gsr = diff(signal);
    max_gsr_phas = max(phasic_gsr);
    min_gsr_phas = min(phasic_gsr);
    mean gsr phas = mean(phasic gsr);
    var_gsr_phas = var(phasic_gsr);
    [peaks, ~] = findpeaks(signal);
    rate_peaks_gsr = length(peaks) / length(signal) * 128;
    reg_coef_gsr = polyfit(1:length(signal), signal, 1);
    reg_coef_gsr = reg_coef_gsr(1);
```

```
features = [max_gsr, min_gsr, mean_gsr, var_gsr, ...
                 max_gsr_phas, min_gsr_phas, mean_gsr_phas, var_gsr_phas, ...
                rate_peaks_gsr, reg_coef_gsr];
end
function features = extractPPGFeatures(signal)
    if isempty(signal)
         features = NaN(1, 7); % Return NaNs for empty signals
         return;
    end
    max_ppg = max(signal);
    min_ppg = min(signal);
    mean ppg = mean(signal);
    var_ppg = var(signal);
     [peaks, locs] = findpeaks(signal);
     rate_peaks_ppg = length(peaks) / length(signal) * 128;
    IBI = diff(locs) / 128;
    IBI_mean = mean(IBI);
    SDNN = std(IBI);
    features = [max_ppg, min_ppg, mean_ppg, var_ppg, rate_peaks_ppg, IBI_mean,
SDNN];
end
```

EXtracting features, from segmented data

Applying the Pre-Trained Classifier from second Assignment and evaluating it bestModel_final.mat

```
% Initialize feature arrays
features_GSR = [];
features_PPG = [];
labels = [];
subjects = [];

% Extract features from the segmented data
for subject = 1:length(CLData)
    markers = CLData(subject).GSRPPGmarker;
    segments = find(diff(markers) ~= 0);
    segments = [1; segments + 1; length(markers) + 1];
    for i = 1:length(segments) - 1
        segment = segments(i):segments(i + 1) - 1;
        marker = markers(segment(1));
```

```
if marker == 3 % MathCalculus or CognLoad
             gsr features =
extractGSRFeatures(CLData(subject).GSRnorm(segment));
             ppg features =
extractPPGFeatures(CLData(subject).PPGnorm(segment));
             features GSR = [features GSR; gsr features];
             features_PPG = [features_PPG; ppg_features];
             labels = [labels; {'CognLoad'}];
             subjects = [subjects; subject];
         elseif marker == 2 % AudioListening or relax
             gsr_features =
extractGSRFeatures(CLData(subject).GSRnorm(segment));
             ppg_features =
extractPPGFeatures(CLData(subject).PPGnorm(segment));
             features GSR = [features GSR; gsr features];
             features_PPG = [features_PPG; ppg_features];
             labels = [labels; {'relax'}];
             subjects = [subjects; subject];
         end
     end
end
% Combine the features from GSR and PPG signals
combined features = [features PPG, features GSR];
% Load the trained model
model = load('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/bestModel final.mat');
trainedModel = model.trainedModel final;
% Convert the combined features to a table with appropriate variable names
requiredVars = trainedModel.RequiredVariables;
 combined_features_table = array2table(combined_features, 'VariableNames',
requiredVars);
% Convert labels to categorical
labels = categorical(labels);
% Initialize arrays for LOSO cross-validation results
predicted_labels_LOSO = [];
actual labels LOSO = [];
% Perform LOSO cross-validation
unique_subjects = unique(subjects);
for i = 1:length(unique subjects)
    % Split the data into training and test sets based on subjects
```

```
test_idx = (subjects == unique_subjects(i));
    train_idx = ~test_idx;
    train_data = combined_features_table(train_idx, :);
    test_data = combined_features_table(test_idx, :);
    train labels = labels(train idx);
    test_labels = labels(test_idx);
    % Use the pre-trained model for predictions
    predictions = trainedModel.predictFcn(test_data);
    % Convert numeric predictions to categorical
    predicted_labels = categorical(cellstr(predictions));
    % Append the results
    predicted_labels_LOSO = [predicted_labels_LOSO; predicted_labels];
    actual_labels_LOSO = [actual_labels_LOSO; test_labels];
end
% Evaluate the model's performance
confMat = confusionmat(actual labels LOSO, predicted labels LOSO);
accuracy = sum(diag(confMat()) / sum(confMat(:));
fprintf('Overall Accuracy: %.2f%\\n', accuracy * 100);
Overall Accuracy: 66.67%
% Compute other evaluation metrics
precision = diag(confMat) ./ sum(confMat, 2);
recall = diag(confMat) ./ sum(confMat, 1)';
F1 = 2 * (precision .* recall) ./ (precision + recall);
disp(['Precision: ', num2str(mean(precision, 'omitnan'))])
Precision: 0.66667
disp(['Recall: ', num2str(mean(recall, 'omitnan'))])
Recall: 0.8
disp(['F1-score: ', num2str(mean(F1, 'omitnan'))])
F1-score: 0.625
% Save the extracted features and labels
```

```
save('/MATLAB Drive/Third Assignment Material/Third Assignment
Material/Cognitive Load tasks/Features_LOSO.mat', 'combined_features', 'labels',
'predicted_labels_LOSO', 'actual_labels_LOSO');
disp('Finish')
```

Finish

```
disp("codes finished with 0 errores")
```

codes finished with 0 errores

%created by anass nassiri