**CSE 572 Activity recognition Project**

In this project, we will attempt to develop a computing system that can **understand** human activities. You will be provided data for a given activity specifically eating action mixed with other unknown activities. Your aim is to identify the eating activities amongst the noise. We will be working with real world wristband data and in the process we will use all the concepts that we will learn in the data mining course. Note that this is only attempt and we do not expect a full solution yet.

In the first phase we will do data cleaning, noise reduction, data organization and feature extraction.

**a) Phase 1 Data cleaning and organization**

The data provided to you is collected using two sources: A) wristband data: where the subject is wearing the wristband and performing eating actions periodically interspersed with non-eating unknown actions. The wristband provides you with i) accelerometer, ii) gyroscope, iii) orientation, and iv) EMG sensors. The sampling rate is 50 Hz for all the sensors. B) Video recording of the person performing eating actions only used for establishing ground truth. In the data, you will be provided with the ground truth in the form of frame numbers where an eating action starts and ends. The actual video data will not be provided for privacy issues. The video data is taken at 30 frames per second. Hence you have to convert the frame numbers into sample numbers for the wristband sensors through some logic that you have to develop. The assumption that you can take is that the start frame and sample #1 of the wristband are synchronized. The output of this step will be a set of data snippets that are labelled eating actions and a set of data snippets that are non-eating.

**b) Phase 2 feature extraction [50 points]**

In this task, you should select and implement five existing feature extraction methods such as Fast Fourier Transform, Discrete Wavelet Transform, a set of statistical features (min, max, avg, std, RMS, energy function), etc. The five types of feature extraction methods can be chosen by you. The aim is to use features that show clear distinction between the actions.

For each type of feature extracted do the following things,

a) write an explanation on how the feature is extracted.

b) Write an intuition on why you use such a feature

c) Write a matlab code to extract that feature from each time series created in task 1.

d) Generate plots each corresponding to eating and non-eating activities . This will give you a better idea of potential patterns in the features.

e) Discuss whether your initial intuition about the features that you selected holds true or not.

**Phase 3: Feature Selection (50 points)**

This step involves reduction of the feature space and keeping only those features which show maximum distance between the two classes (eating and non-eating). We will use Principal Component Analysis technique discussed in class for this purpose. The PCA code is already available in Matlab, hence there is no need to PANIC! Just use it.

Subtask 1: Arranging the feature matrix

You know PCA only takes one matrix. How will you arrange all sensors and their corresponding features into a single matrix such that the eigenvectors of the covariance matrix directly makes sense to your data set? This means that if the PCA results gives you a eigen vector then the new feature matrix can be obtained by simply multiplying the eigen vector with the old feature matrix.

Write your logic of feature matrix arrangement.

Subtask 2: Execution of PCA

Use Matlab’s PCA function to run PCA on your feature matrix. Show all the eigen vectors in a plot.

Subtask 3: Make sense of the PCA eigen vectors

Write an explanation on the reason why the eigen vectors turned out the way they did.

Subtask 4: Results of PCA

Create the new feature matrix.

Subtask 5: Argue whether doing PCA was helpful or not. May be compare the plots generated from subtask d of task 2 and subtask 4 of Task 3.

Submit Code and PDF file with your explanations in Canvas.

**CSE 572 Data Mining Project 2**

**Due Date: February 26th, 2019**

**User dependent analysis**

Consider the new set of features that you obtained by multiplying the PCA output with your feature set.

Divide that new feature set into two parts for each user: a) part 1: training and b) part 2: test. Ideally keep 60% of the data for each user as training and the rest of 40% as test data. Use three types of machines: a) decision trees (fitctree in Matlab), b) support vector machines (fitcsvm in Matlab), and c) neural networks (use the neural network toolbox in Matlab).

Train each machine with the training data and then use the test data to report accuracy. Use the accuracy metrics of Precision, Recall, F1 score. Report each metric for every user.

**User independent analysis**

For a given gesture, consider 60% of total users and use all their feature points of each user as training. Follow the same labelling strategy as considered in previous user dependent analysis. The rest users are testing. Do the same analysis as in previous case and report the same metrics for each of the rest of the test users.

Submit the code and PDF file with your explanations in Canvas by the due date.