Topic: Clustering-Classification Approach For Human Activity Detection Using Smartphone Dataset

Shantanu Agrawal 1901182

Course: CS360 - Machine Learning Lab

Problem Definition

Problem Definition

The objective of this project is to recognize 6 activity of a human (Sitting, Standing, Laying, Walking, Walking Downstairs, Walking Upstairs) by using gyroscope and accelerometer sensor data using smart phone

Topics/Models Covered

- Classification
- Logistic Regression
- Perceptron
- Single Layer Perceptron
- Multi Layer Perceptron

- Clustering
- K-Means
- o K-Medoid
- Fuzzy
- SOM

Paper: Human Activity Recognition on Smartphones for Mobile Context Awareness
By Davide Anguita, Alessandro Ghio, Luca Oneto, Xavier Parra, Jorge L. Reyes-Ortiz

Setup Feature Selection Classification

- Support Vector Machine(SVM) Algorithm used.
- Uses One-VS-All Approach for classification.
- Reason of using smartphone for collecting data is the high number of growth in the user in last decade and the needed sensors like gyroscope, accelerometer hardware are already present.
- In this paper, researchers are improving SVM to be used on resource limited devices by extending SVM to Hardware-Friendly SVM (HF-SVM) then to Multiclass HF-SVM (MC-HF-SVM).
- Use of MC-HF-SVM can be smartwatch since it has very small resources available.

- Paper: A Public Domain Dataset for Real-Life Human Activity Recognition Using Smartphone Sensors By Daniel, Daniel, Enrique and Miguel
- The main idea is to make activity recognition use more real world data than the model data provided under UCI Human Activities Recognition (HAR) repository
- Support Vector Machine(SVM) Algorithm used due to fixed-point arithmetic data points. .
- Uses One-VS-All Approach for classification.
- Since the real world data is more unorganized, the first step is to organize it and make it easier to work with.
- More sensors are used to collect data, like GPS, Magnetometer and more types of activities recognition is aimed for Use of MC-HF-SVM can be smartwatch since it has very small resources available.
- Accuracy is slightly lower than previous works due to more real world data.

Paper: A Smartphone-Based Adaptive Recognition and Real-Time Monitoring System for Human Activities By Wen, Hang and Andrea

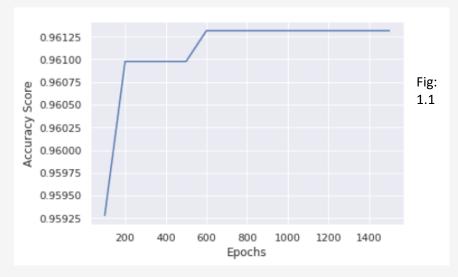
- Main idea is introduction of real-time human activity monitoring system and a adaptive learning environment.
- Unsupervised learning algorithm as we want to classify new unrecognized activity and add a label to it for future classification.
 This new label will be added to the class.
- Researchers are aiming for online learning algorithm which will be efficient in the terms of resources
- Hierarchical clustering has been used to learning

Result Analysis

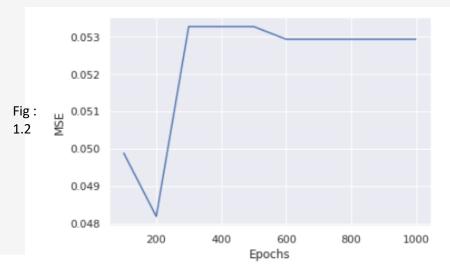
Classification: Logistic Regression

- ☐ As shown in fig 1.1 by on varying epoch accuracy score
- ☐ Around 600 epochs convergence occur
- ☐ After 1000 epochs iteration
 - ❖ Train Accuracy Score: 99.25%❖ Test Accuracy Score: 96.13%
 - ❖ Log Loss Error: 0.119389576982268
 - ❖ Mean Squared Error : 0.05293518832711232
- ☐ Faster compare to other models

■ Model is performing well since it is achieving 96.13% accuracy

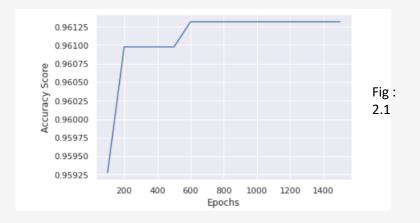


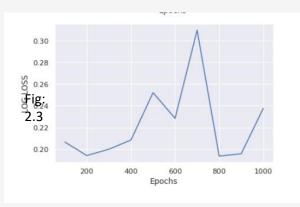
Epoch vs Mean Squared error (Fig: 1.2)

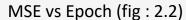


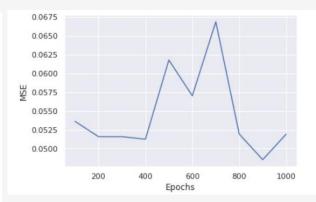
Classification: Stochastic Gradient Descent (ONE Vs ALL)

- ☐ As shown in Fig 2.1 we plot Accuracy vs epoch graph
- ☐ Around 800 epoch accuracy is similar
- After 1000 iterations
 - ❖ Train Accuracy Score: 98.59%
 - Test Accuracy Score: 96.13%
 - Mean Squared Error: 0.04784526637258229
 - **❖** Log Loss Error: 0.2194638846322902
- ☐ K-Fold Cross-validation is a statistical method used to estimate the skill of machine learning models.
- ☐ After K-Fold Validation
 - ❖ Accuracy Score: 94.63%









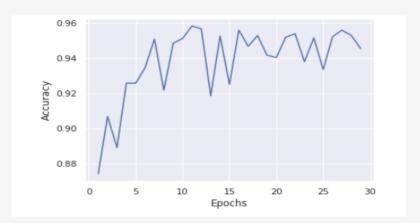
Log Loss vs Epoch (fig: 2.3)

Classification: Perceptron

- ☐ It does not require a learning rate
- ☐ It updates its model only on mistakes.
- ☐ As shown in fig 3.1 by on varying epoch accuracy score
- ☐ After 30 epochs iteration
 - ❖ Train Accuracy Score: 95.06%❖ Test Accuracy Score: 90.73%
 - ❖ Mean Squared Error : 0.10960298608754666

☐ In MSE vs Epoch (fig : 3.2)
After 20 epoch MSE is almost constant





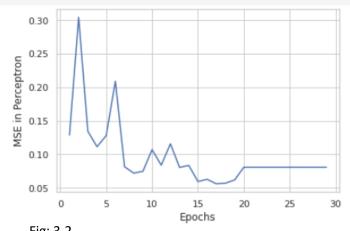
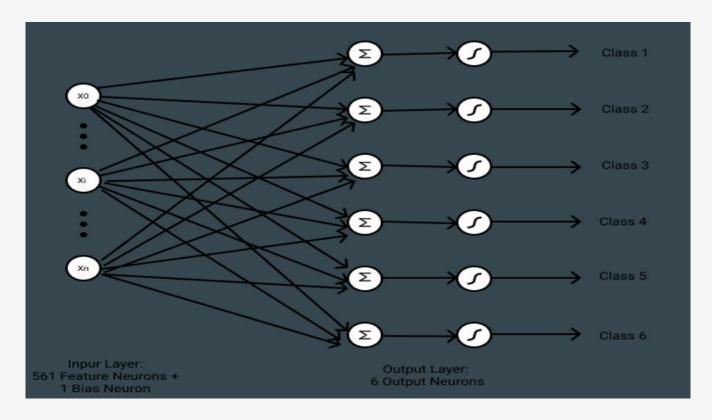


Fig: 3.2

Classification: Single Layer Perceptron (SLP)

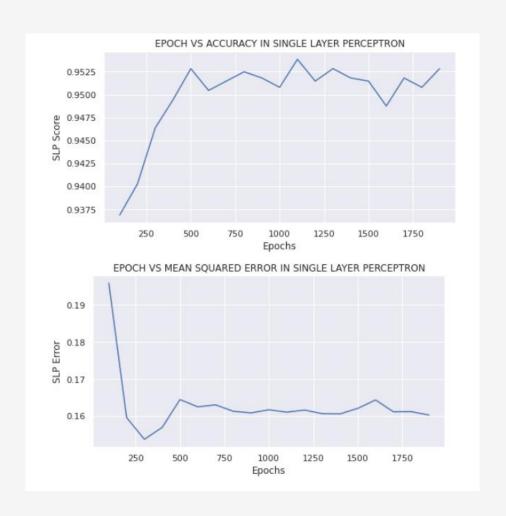
- ☐ It is a feed-forward neural network
- ☐ It has a single layer of perceptrons or artificial neurons. It updates its model only on mistakes.

Architecture



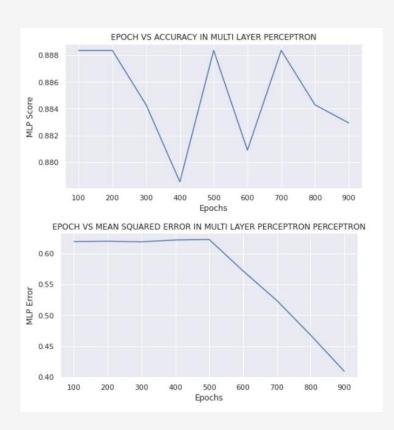
Classification: Single Layer Perceptron (SLP)

- ☐ Slower compared to other models
- ☐ Sigmoid activation function is used
- ☐ Model converges with high accuracy
- ☐ Higher accuracy achieved 95.35%



Classification: Multi Layer Perceptron (MLP)

- □ It has one input layer, one output layer, and multiple hidden layers (feed-forward architecture).
- ■No of hidden neuron taken is 562
- □Slow to train for increased number of layers and nodes
- □Slow to predict in case of large number of layers
- ■Not suitable for real time accurate computations
- ☐ Highest accuracy achieved 88.89%



Clustering

- ☐ Fast compared to other models.
- □ Algorithms Used K-Means, K-Medoids, Fuzzy Clustering, SOM
- ☐ Model converges with very low accuracy
- ☐ Highest Accuracy achieved in K-Means 32.37%
- ☐ Highest Accuracy achieved in K-Medoid 20.0543%
- ☐ Highest Accuracy achieved in Fuzzy 24.8728%
- ☐ Highest Accuracy achieved in SOM 9.3482%
- ☐ Poor accuracy even with the reduced features.

Refernces

- https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LogisticRegression.html
- https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.SGDClassifier.html
- https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.Perceptron.html
- https://scikit-learn.org/stable/modules/generated/sklearn.neural_network.MLPClassifier.html
- https://scikit-learnextra.readthedocs.io/en/stable/generated/sklearn_extra.cluster.KMedoids.html
- https://scikit-learn.org/stable/modules/generated/sklearn.cluster.KMeans.html
- https://pypi.org/project/fuzzy-c-means/