# CS6910 PG Section Instructor: Mitesh Khapra Assignment 1

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WandB Report Link: wandb Report Link

GitHub Repository Link: Github Repository

## CS6910 PG Section Instructor: Mitesh Khapra Assignment 1

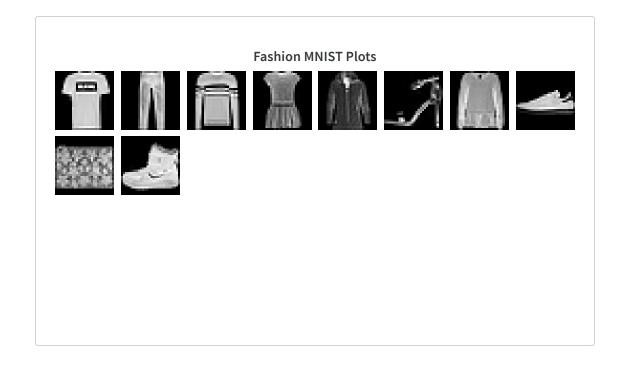
Report for CS6910 Assignment 1

Sandeep Kumar Suresh ee23s059

- Github Link - Github

#### \* Question 1 (2 Marks)

Download the fashion-MNIST dataset and plot 1 sample image for each class as shown in the grid below. Use from keras.datasets import fashion\_mnist for getting the fashion mnist dataset.



#### Question 2 (10 Marks)

Submitted the code in moodle

#### \* Question 3 (24 Marks)

Code Submitted in Moodle

#### \* Question 4 (10 Marks)

Wandb provides different 3 methods to do Hyperparameter Sweeping which are Random, bayes and grid.

Grid Search tries every combination of the hyperparameter

Random Search Select each new combination at random according to provided distributions

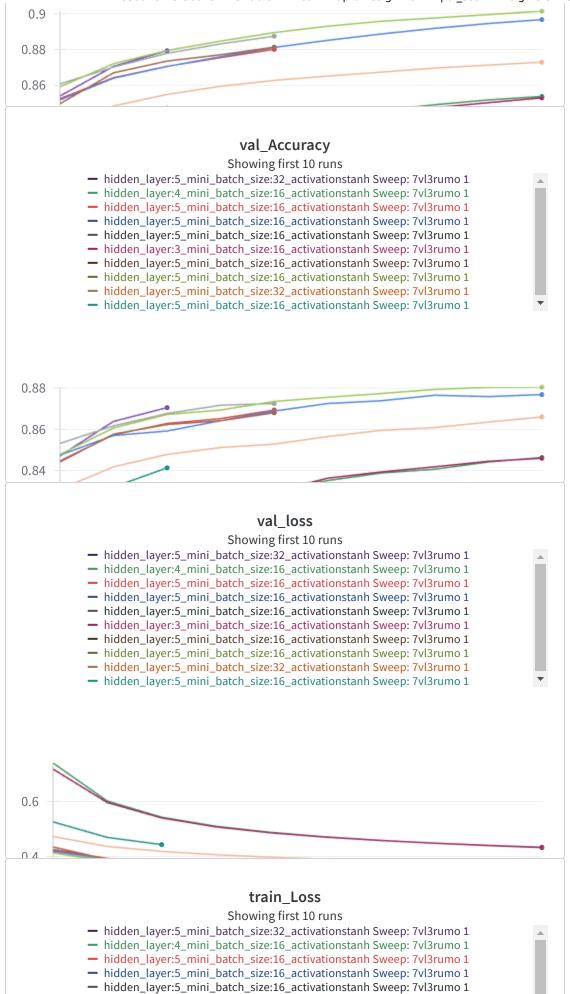
Bayesian Search creates a probabilistic model of metric score as a function of the hyperparameters, and choose parameters with high probability of improving the metric .

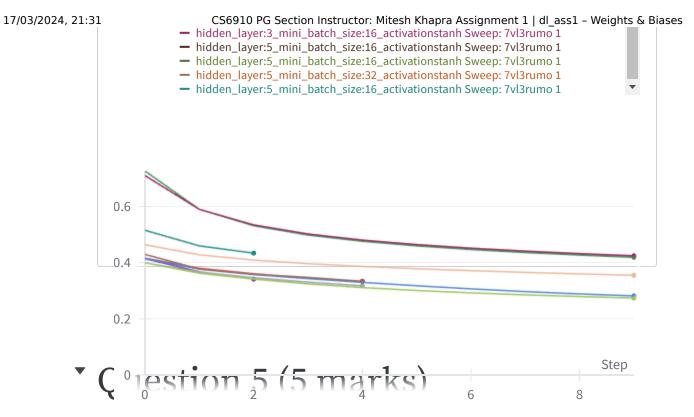
I have choosen bayesian Search as it is computatioally effective and uses a probabilistic approach to find the best set of hyperparameter that can maximize the Validation Accuracy

```
train_Accuracy

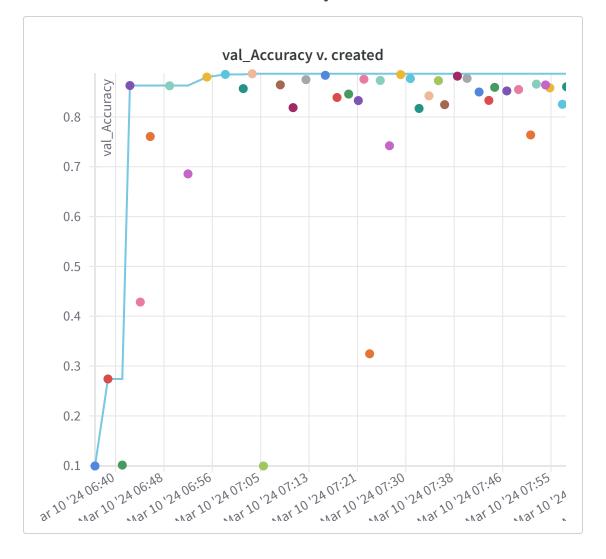
Showing first 10 runs

— hidden_layer:5_mini_batch_size:32_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:4_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:3_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:32_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
— hidden_layer:5_mini_batch_size:16_activationstanh Sweep: 7vl3rumo 1
```



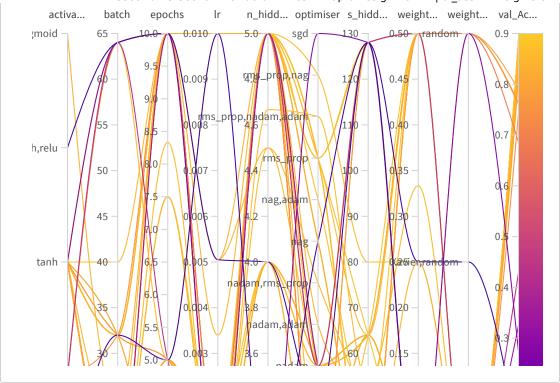


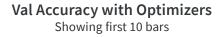
- I got a maximum Validation Accuracy of around 88.82 %.
- The minimum Validation Accuracy is around 9.95 %



#### \* Question 6 (20 Marks)

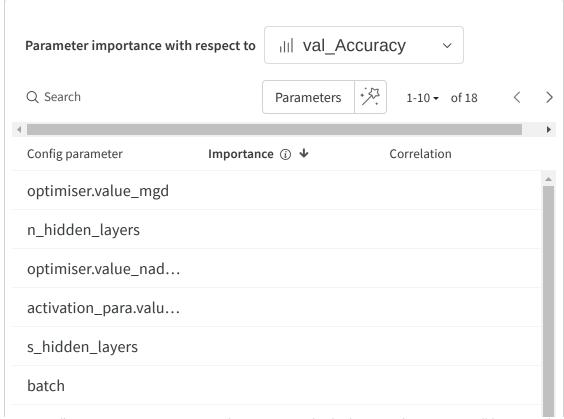
- 1. The highest validation Accuracy is **88.82% without**Regularization. The configuration was nadam,n\_layers = 4
  ,s\_hidden\_layer = 64, batch\_size = 32,lr = 1e-3,Xavier
  Initialization, epochs = 10, tanh activation.
- 2. The lowest Validation Accuracy Observed was around **9.95%** without Regularization . The configuration was mgd,n\_layers = 5 ,s\_hidden\_layer = 128 , batch\_size = 64,lr = 1e-2,Xavier Initialization , epochs = 10 ,tanh activation.
- 3. When I grouped the sweep based on Optimiser that gave the best validation Accuracy, I observed that most gave best result with Xavier Initialization.
- 4. When filtered with group of Validation Accuracy and Optimiser Together, nadam and adam came in the top 20 list. Most of the activation function was tanh, learning rate of 1e-2, hidden layer of 4 or 5 with Xavier Initialization was most common observation. This states the fact that nadam and adam are the best optimizer. (Note: nag and rms\_prop came into the list).
- 5. Based on the Correlation Plot, n\_hidden\_layer is a parameter of importance which necessarily had a positive correleation to maximize the validation Accuracy
- 6. nadam,sgd,nag,mgd were the optimizer which gave validation accuracy below 65%. Eventhough nadam gave highest validation, maybe the learning rate of 1e-4,sigmoid activation and size of hidden layer could have made it performance degrade.
- 7. Definitly nadam, adam can be used as optimizer configuration to get more than 95% with Regularization and Data Augmentation Added. The below two plots are plots that does not involve Regularization.



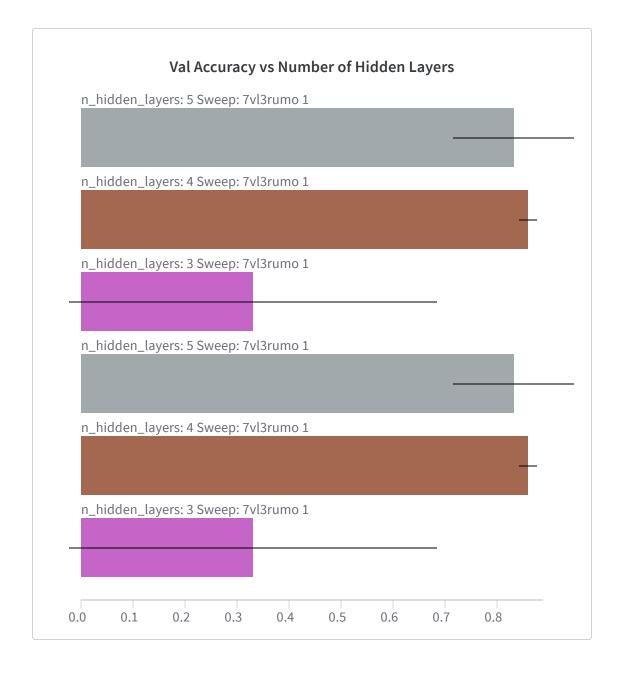


optimiser: nadam Sweep: 7vl3rumo 1

- Now based on the "Correlation Summary", we can see that n\_hidden\_layers is having a positive correlation on the validation Accuracy. Therefore on increasing the n\_hidden\_layers in the network, the accuracy may increase.
- Tanh Activation based on the "Parallel co-ordinates plot" and a "correlation summary" have shown to increase the validation Accuracy. This has a higher positive correlation than n\_hidden\_layers
- The nadam optimizer as expected tends to have positive correlation . This could vary from different configuration used.
- Epoch has little importance and have a negative correlation. This could mean that increasing the number of epoch could be computationally expensive and may end in regions where the loss is very high.
- The adam optimizer tends to work for certain configuration based on "Correlation Summary". But it is of least important. This is also the case with rms\_prop.
- Xavier Initialization works best for all the parameter configuration







#### Question 7 (10 Marks)

For the best model identified above, report the accuracy on the test set of fashion\_mnist and plot the confusion matrix as shown below. More marks for creativity (less marks for producing the plot shown below as it is)

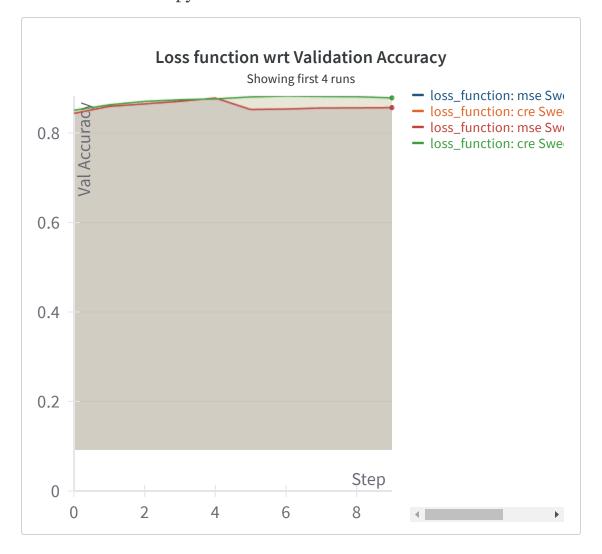
- The best model configuration identified is nadam with n\_layers = 4,s\_hidden\_layer = 64, batch\_size = 32,lr = 1e-3,Xavier Initialization, epochs = 10, cross-entropy loss and tanh activation.
- We can see that most shirt got mispredicted as T-shirt and some as Pullovers and Coats
- T-shirts, Pullovers and Coats are similar to each other. Therfore there are mispredictions for these classes.



### \* Question 8 (5 Marks)

In all the models above you would have used cross entropy loss. Now compare the cross entropy loss with the squared error loss. I would again like to see some automatically generated plots or your own plots to convince me whether one is better than the other.

- I ran Sweep with cross entropy loss and squared error loss as parameters
- We can see that the cross entropy loss performed better than squared error loss wrt to Validation Accuracy. Hence we can say that cross entropy loss is better.



#### \* Question 9 (10 Marks)

https://github.com/sandeepkumarsuresh/cs6910\_assignment1

#### Question 10 (10 Marks)

- Based on my understanding from the experimentation of Fashion MNIST Dataset . I would work on
- 1. Optimiser: Constraining it to adam, nadam and nag
- 2. Activation Function: tanh, relu or sigmoid
- 3. Number of Hidden Layer: 3,4,5
- The highest Validation Accuracy for MNIST Dataset is 97.31%.
- The Configuration-1 for 97.31% is adam optimiser, tanh activation,batch=16,epoch=5,lr=1e-3,Xavier Initialization,n\_hidden layers = 3, s\_hidden\_layer = 128
- The Configuration-2 for 97 % nadam optimiser,tanh activation,batch=64,epoch=5,lr=1e-3,Xavier Initialization,n\_hidden layers =4, s\_hidden\_layer =128
- The Configuration-2 for 96.7% nag optimiser, relu activation,batch=32,epoch=5,lr=1e-3,Xavier Initialization,n\_hidden layers =4, s\_hidden\_layer =128
- Interestingly we can see that the relu function also had a correlation in getting accuracy above 95%.
- From this experiment what I observed is that based on the data we have, certain activation function will provide better result.In the Fashion MNIST Dataset, tanh activation gave better result.

val\_Accuracy v. created

#### \* Self Declaration

I, Sandeep Kumar Suresh, swear on my honour that I have written the code and the report by myself and have not copied it from the internet or other students.

Created with **a** on Weights & Biases.

https://wandb.ai/ee23s059/dl\_ass1/reports/CS6910-PG-Section-Instructor-Mitesh-Khapra-Assignment-1--Vmlldzo3MTc2NjEz