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Problem 1: Feed Forward Neural Network

Verification of gradient using gradcheck() function:

(base) saammmy@saammmy-xps15:~/Deep_Learning/4\$ /home/saammmy/anaconda3/bin
The Gradient check on 5 training examples gives the error as shown below:
1.0302307469074127e-06

Best Hyperparameters:

```
Validation loss and Accuracy: 0.37 , 87.98 %
Hyperparameters Updated
Best Hyperparameter uptill now: [50, 3, 35, 1, 0.01, 16]

Grid Search Completed

Results after performing Grid Search:
Best Hyperparameters:
Hidden Inputs in Each Layer= 50
Hidden Layers= 3
Epochs= 35
Alpha= 1
Learning Rate= 0.01
Mini Batch Size= 16
```

• Training Loss (on training + validation dataset) and Test Accuracy:

```
Training on Training + Validation Dataset:
Training loss and Accuracy at Epoch 1: 0.60, 78.99%
Training loss and Accuracy at Epoch 2: 0.48, 83.34%
Training loss and Accuracy at Epoch 3: 0.42, 85.04%
Training loss and Accuracy at Epoch 4: 0.40, 85.83%
Training loss and Accuracy at Epoch 5: 0.38, 86.49%
Training loss and Accuracy at Epoch 6: 0.36, 87.06%
Training loss and Accuracy at Epoch 7: 0.35, 87.48%
Training loss and Accuracy at Epoch 8: 0.34, 87.70%
Training loss and Accuracy at Epoch 9: 0.32, 88.18%
Training loss and Accuracy at Epoch 10: 0.32, 88.36 %
Training loss and Accuracy at Epoch 11: 0.31, 88.62%
Training loss and Accuracy at Epoch 12: 0.30, 88.84%
Training loss and Accuracy at Epoch 13: 0.30, 89.05%
Training loss and Accuracy at Epoch 14: 0.29, 89.29%
Training loss and Accuracy at Epoch 15: 0.28, 89.61%
Training loss and Accuracy at Epoch 16: 0.28, 89.72%
Training loss and Accuracy at Epoch 17: 0.27, 89.92%
Training loss and Accuracy at Epoch 18: 0.27, 90.14%
Training loss and Accuracy at Epoch 19: 0.26, 90.37%
Training loss and Accuracy at Epoch 20: 0.26, 90.44%
Training loss and Accuracy at Epoch 21: 0.25, 90.49%
Training loss and Accuracy at Epoch 22: 0.25, 90.66%
Training loss and Accuracy at Epoch 23: 0.25, 90.75%
Training loss and Accuracy at Epoch 24: 0.25, 90.83 % Training loss and Accuracy at Epoch 25: 0.24, 90.91 %
Training loss and Accuracy at Epoch 26: 0.24, 91.08%
Training loss and Accuracy at Epoch 27: 0.24, 91.13%
Training loss and Accuracy at Epoch 28: 0.24, 91.08%
Training loss and Accuracy at Epoch 29 : 0.24 , 91.12 %
Training loss and Accuracy at Epoch 30: 0.23, 91.17%
Training loss and Accuracy at Epoch 31: 0.23, 91.27%
Training loss and Accuracy at Epoch 32: 0.23, 91.43%
Training loss and Accuracy at Epoch 33: 0.23, 91.50 %
Training loss and Accuracy at Epoch 34: 0.22, 91.56%
Training loss and Accuracy at Epoch 35 : 0.22 , 91.70 %
Performance Evaluation
Testing loss and Accuracy: 0.37 , 87.67 %
```

Problem 2: Mountains and Valleys

• Visualize the SGD trajectory of our network (with 3 hidden layers and 50 neurons in each hidden layer) when trained on Fashion MNIST dataset.

The procedure for generating this visualization is as follows -

- We first performed Principal Component Analysis (PCA) and convert the original high dimensional weight-space to into a 2-D space (X-Y).
- Next, we find the min and max components in each of the X and Y directions.
- We then generate 100 samples between min and max components in both the directions. This gives us a weight space of 100*100 = 10000 components.
- After, taking the inverse PCA of all the 10000 points to generate a weight space of the original dimensions.
- We then calculate Loss on training data for all the 10000 points using these weights and then plot this Loss on the Z-axis corresponding to X and Y.
- By applying interpolation between these points, we get the surface plot as shown below.
- Now, we again perform the inverse PCA for all the 2-D weight space components we got during the training (the number of 2-D weight space components will be equal to num_of_epoch.)
- And again, calculated Loss on these weights, which is then shown as a "BLUE" scatter plot to represent how the weights and biases of our network vary in the weight space.
- The following two image represent the same plot form two different viewing angles.

