

## CS535 Deep Learning , Assignment 2, Dongkyu Kim

1) and 2) → please see the source code.

### 3) The best hyper-parameter tuned.

Below parameters were decided by the results of 5) Turing Parameters.

```
mini_batch_size = 64
learning_rate = 0.0008
hidden_units = 2000
seed = 10
num_epochs = 100
momentum = 0.9
```

```
Train[Epoch 75 mini-batch 156 Cost = 5.849]
Train[Avg.cost: 0.0005849087498108696 Accuracy: 99.77%]
Test[Cost: 0.4332420980283863 Accuracy: 85.75%]
```

.....

```
Train[Epoch 99 mini-batch 156 Cost = 3.335]
Train[Avg.cost: 0.00033352909675913553 Accuracy: 99.97%]
Test[Cost: 0.4684568376083723 Accuracy: 85.35%]
```

### 4) below is a part of the results

.....

```
Train[Epoch 62 mini-batch 156 Cost = 9.902]
Train[Avg.cost: 0.00099018636592678 Accuracy: 99.15%]
Test[Cost: 0.4216214591585031 Accuracy: 84.55%]
```

```
Train[Epoch 63 mini-batch 156 Cost = 9.712]
Train[Avg.cost: 0.0009711819528149317 Accuracy: 99.13%]
Test[Cost: 0.41480091597693647 Accuracy: 85.30%]
```

```
Train[Epoch 64 mini-batch 156 Cost = 9.411]
Train[Avg.cost: 0.0009411072369252203 Accuracy: 99.23%]
Test[Cost: 0.41709766101330514 Accuracy: 85.10%]
```

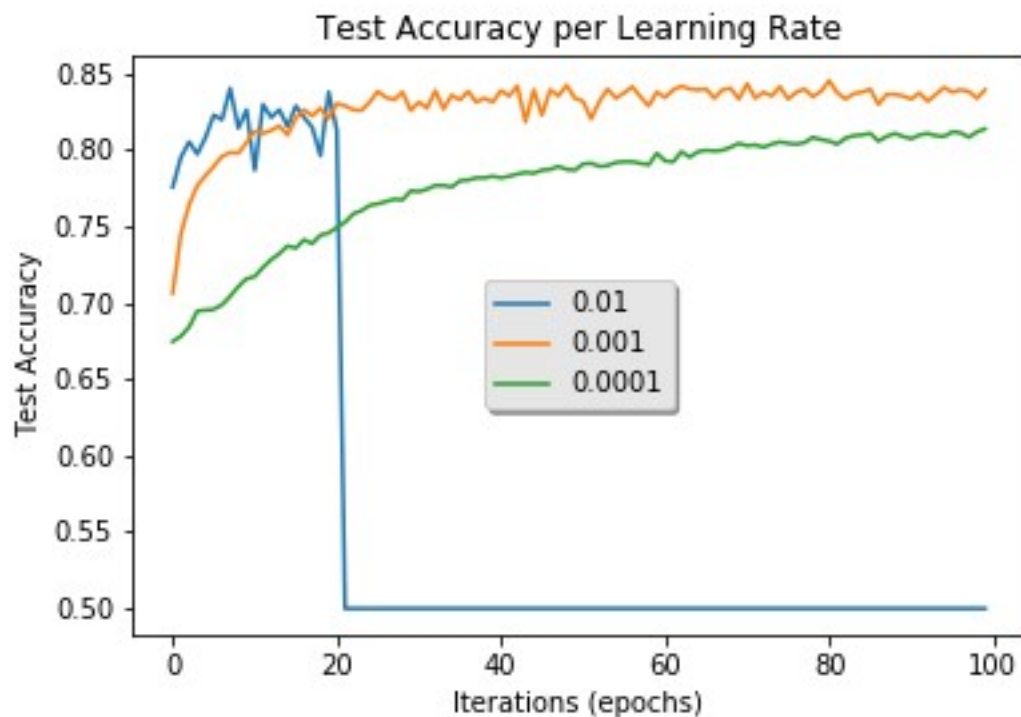
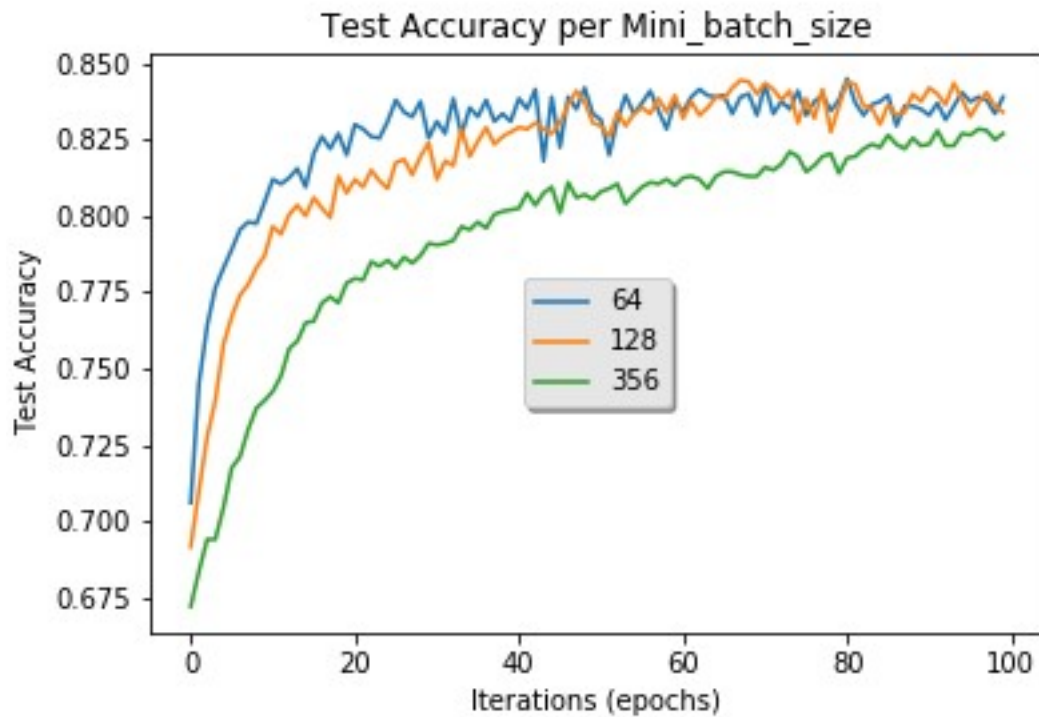
```
Train[Epoch 65 mini-batch 156 Cost = 9.002]
Train[Avg.cost: 0.0009001835106140099 Accuracy: 99.36%]
Test[Cost: 0.42023565611280844 Accuracy: 84.30%]
```

```
Train[Epoch 66 mini-batch 156 Cost = 8.834]
Train[Avg.cost: 0.000883444180390107 Accuracy: 99.33%]
Test[Cost: 0.4296430719418355 Accuracy: 84.35%]
```

```
Train[Epoch 67 mini-batch 156 Cost = 8.584]
Train[Avg.cost: 0.0008584292426807811 Accuracy: 99.42%]
Test[Cost: 0.421883786647761 Accuracy: 85.25%]
```

.....

### 5) Tuning Parameters





## 6) Discussion

- Vanishing gradient

As you can see above figures regarding to learning rates and hidden units in 5), the accuracy drops to 50%. Also, the information of training-testing monitoring shows as below:

```
Train[Epoch 22 mini-batch 156 Cost = nan]
Train[Avg.cost: nan Accuracy: 50.00%]
Test[Cost: nan Accuracy: 50.00%]
```

```
Train[Epoch 99 mini-batch 156 Cost = nan]]
Train[Avg.cost: nan Accuracy: 62.13%]
Test[Cost: nan Accuracy: 50.00%]
```

Cost values become NaN, and it could be an evidence of vanishing gradient problem. The problem is that in some cases, the gradient will be vanishingly small, effectively preventing the weight from changing its value. In the worst case, this may completely stop the neural network from further training.

- Since it performs mini-batch gradient descent, graphs do not show linear lines. Each batch has a different a cost(loss) value, thus its accuracy could be fluctuated. When mini-batch size becomes the total sample size, it works as a batch gradient descent, and it will take too long per iteration. On the other hand, when mini-batch size is too small, it works as a stochastic gradient descent, and it will lose speed up from the vectorization. The result also shows a tendency that the accuracy goes up until certain points as iteration(epochs) increases.

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- In terms of the learning rate change, the bigger a learning rate is, then the wider fluctuation is, but it reaches to a minimum area but not exactly the minimum earlier. It might oscillate around the minimum(optimal) point. On the other hand, when a learning rate is small, it searches the minimum point sophisticatedly. Thus, it reaches the higher accuracy slowly as shown in the figure, and it reduces the effect of the momentum.
- Regarding hidden units, the last figure in 5) shows that when the number of hidden units increases, the neural network's performance will be improved.
- Considering these findings and several experiments with other parameters such as momentum and number of epochs, I derived the best tuned parameters as shown in 3)