

Assignment 2 bonus

DD2424

1. Optimize the performance of the network

1.1. Grid search on the length of the cycles and the number of the cycles.

($\lambda = 10^{**}(-2.82638)$)

| Accuracy(test) | ns=1epoch | ns=2epoch | ns=3epoch | ns=4epoch | ns=5epoch | ns=6epoch | ns=7epoch | ns=8epoch |
|----------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
| 1cycle | 47.72 | 50.1 | 50.54 | 51.12 | 51.15 | 51.9 | 51.79 | 51.8 |
| 2cycles | 50.08 | 51.38 | 51.7 | 51.91 | 51.46 | 52.2 | 52.41 | 52.16 |
| 3cycles | 50.87 | 51.68 | 51.93 | 52.21 | 51.77 | 51.98 | 52.55 | 52.58 |
| 4cycles | 51.42 | 52.3 | 52.35 | 51.97 | 51.81 | 52.15 | 52.75 | 52.15 |
| 5cycles | 51.57 | 51.83 | 52.21 | 52.05 | 52.05 | 52.04 | 52.61 | 52.72 |

When the ns increases and cycles increase, the accuracy on test data is increasing overall.

So I choose the hyper parameters which achieve the highest accuracy—na=7epoch, 4cycles.

1.2 Apply more hidden nodes

($\lambda = 10^{**}(-2.82638)$, cycle=2, ns=7epoch)

| Hidden nodes | 50 | 75 | 100 | 150 | 200 | 300 |
|---------------|-------|-------|-------|-------|-------|-------|
| Test accuracy | 52.01 | 53.24 | 53.38 | 53.38 | 52.01 | 52.02 |

We can see from the results that increasing hidden nodes can improve the accuracy at some extend, but we will encounter overfitting if the hidden node are too many.

1.3 Apply drop out to training

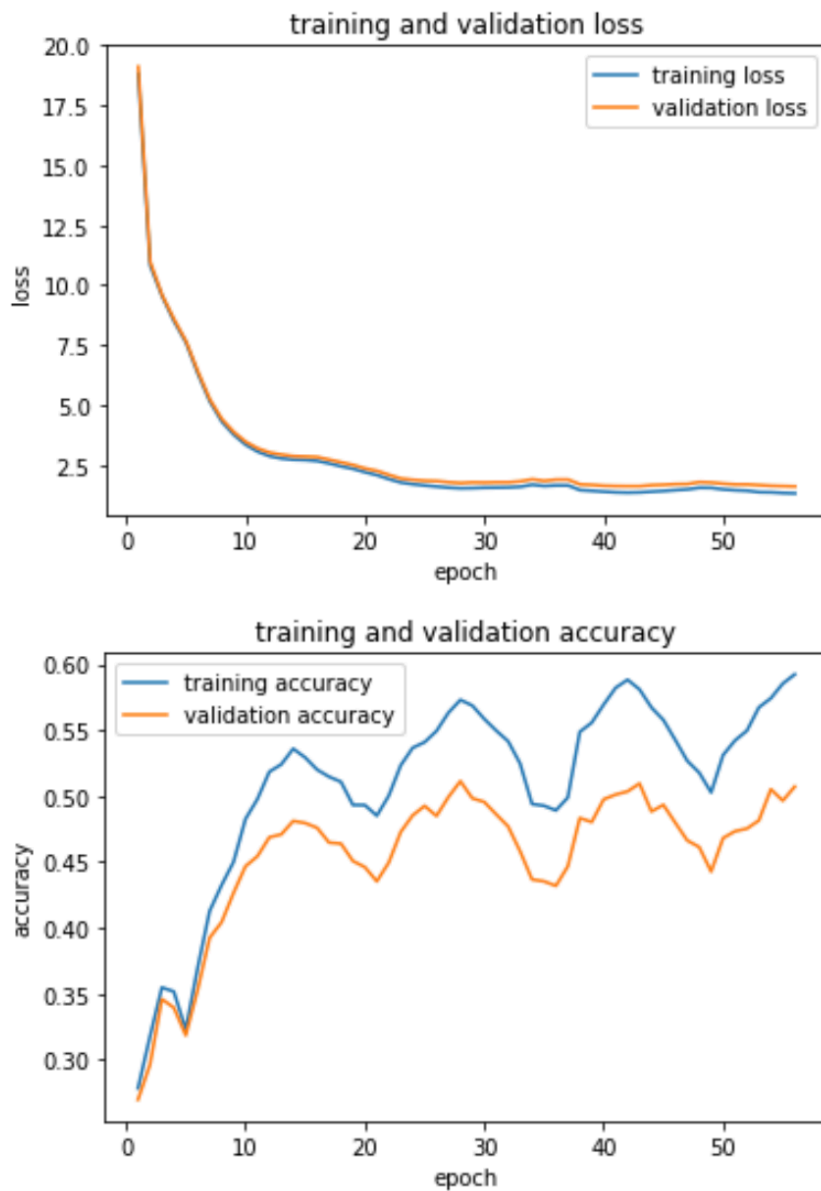
($\lambda = 10^{**}(-2.82638)$, cycle=2, ns=7epoch, hidden nodes=100)

| Drop out rate | 0 | 0.2 | 0.3 | 0.5 |
|-------------------|-------|-------|-------|-------|
| Training accuracy | 71.81 | 57.30 | 52.87 | 45.94 |
| Test accuracy | 53.38 | 54.17 | 53.41 | 50.91 |

We can see that apply proper dropout can increase the performance of test data, but decrease the performance of training data, because we add more redundancy and noise to the data at training time, thus reduces overfitting.

1.4 The best accuracy achieved

-lambda= $10^{*(-2.82638)}$, cycle=4, ns=7epoch, hidden nodes=100, dropout=0.2

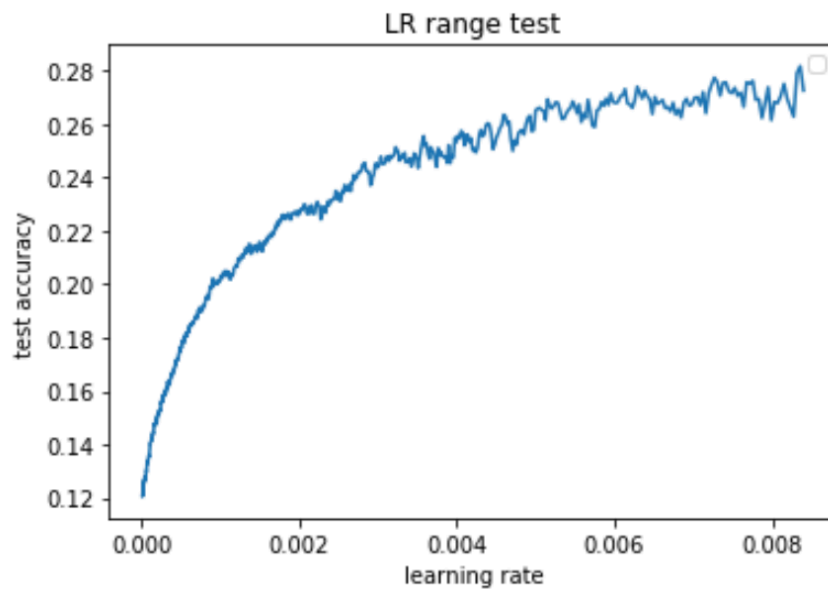


| accuracy | |
|------------|-------|
| Training | 59.24 |
| Validation | 50.72 |
| Test | 54.83 |

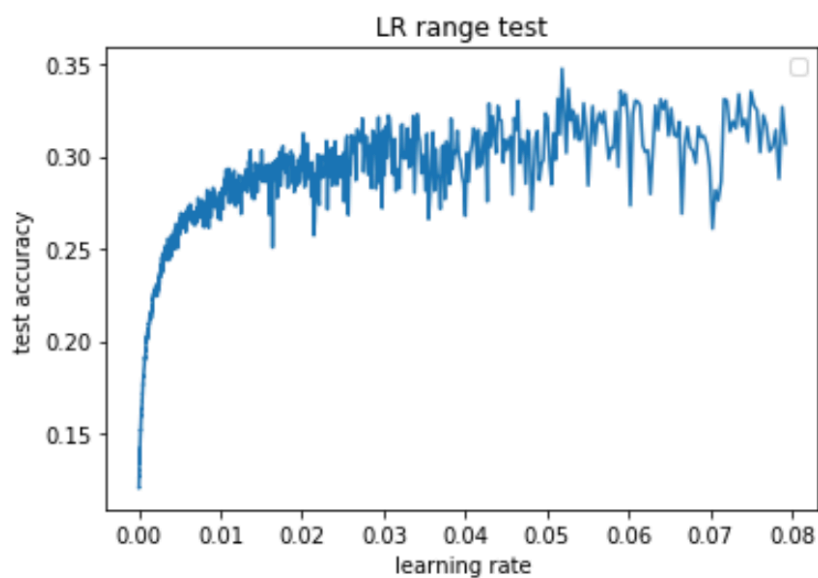
2. Learning rate range test

I use LR range test to find the best boundary for cyclic learning rate. I run my model for several epochs while letting the learning rate increase by multiply by factor 1.004 after every iteration. The other parameters are the same in 1.4. The initial value for learning rate is $1e-5$, and run for 3-5 epochs. Then plot the test accuracy versus learning rate.

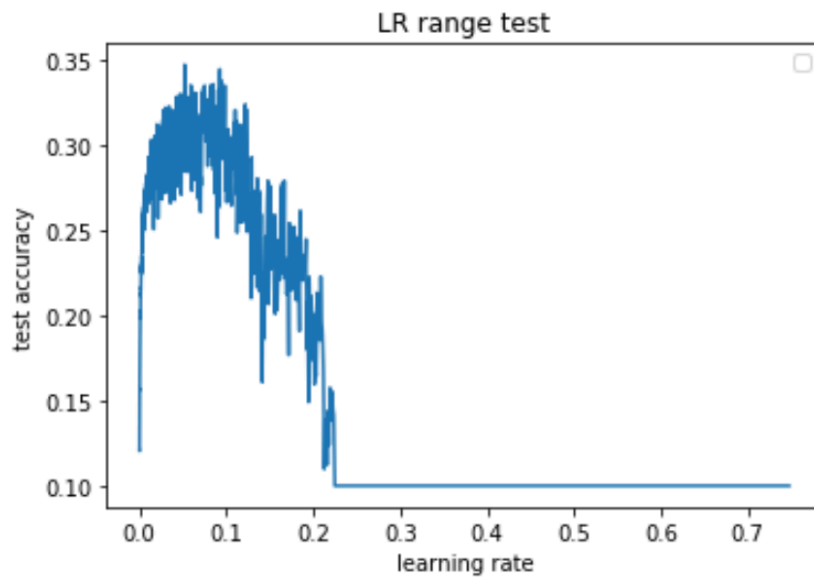
3 epoch:



4 epoch:



5 epoch:



From these results, we take learning rate values when the accuracy starts to increase and when the accuracy slows, became ragged, or starts to fall. I chose max learning rate=0.1 and min learning rate=0.02.