Lab3 Report

Task 1(a)

Sum-of-squares error vs. cross-entropy error function

Sum-of-squares error:

Step1: do experiment with different values of parameters:

- 1) number of hidden layers,
- 2) number of hidden units in each layer,
- 3) learning rates,
- 4) momentum rates,
- 5) epochs

	Parameters							
Hidden	Units	Epochs	Learning	Momentum	Loss & Accuracy			
layer			rates	rates	(train data)			
2	100/50	500	0.01	0.90	Loss: 0.018			
					Acc: 0.9155			
2	100/50	1000	0.01	0.90	Loss: 0.007			
					Acc: 0.9589			
2	100/50	1000	0.02	0.95	Loss: 0.0025			
					Acc: 0.9882			
2	70/35	1000	0.02	0.95	Loss: 0.0042			

					Acc: 0.9772
3	100/50/25	500	0.02	0.95	Loss: 0.0028
					Acc: 0.9856
3	100/50/25	1000	0.02	0.95	Loss: 0.0018
					Acc: 0.9892

Step3: Best parameters config for •Sum-of-squares error function

Config	Results
Num of hidden layers = 3	Execution time: 43s
Units each layer = 100/50/25	Acc(train): .0.9892
Learning rates = 0.02	Loss(train): 0.0018
Momentum rates = 0.95	Acc(test): 0.9482
Epochs = 1000	Loss(test): 0.0084

Step4: Confusion matrix for train data

[[3	373	0	0	0	1	0	1	0	1	0]
[0	384	0	0	0	3	0	0	0	2]
[0	0	377	2	0	0	1	0	0	0]
[0	1	0	383	0	5	0	0	0	0]
[1	0	0	0	384	0	2	0	0	0]
[0	0	0	0	0	376	0	0	0	0]
[0	2	0	0	1	0	374	0	0	0]
[0	0	0	1	0	1	0	384	0	1]
[0	3	2	0	0	2	0	0	372	1]
[1	1	0	3	1	0	0	0	1	375]]

Step5: Class accuracies

	precision	recall	f1-score	support
0	0.99	0.99	0.99	376
1	0.98	0.99	0.98	389
2	0.99	0.99	0.99	380
3	0.98	0.98	0.98	389
4	0.99	0.99	0.99	387

5	0.97	1.00	0.99	376
6	0.99	0.99	0.99	377
7	1.00	0.99	1.00	387
8	0.99	0.98	0.99	380
9	0.99	0.98	0.99	382

Step6: Confusion matrix for test data

[37	4	0	0	0	1	0	0	0	1	0]
[0	385	0	0	0	0	1	0	0	3]
[0	1	377	1	0	0	1	0	0	0]
[0	0	0	383	0	5	0	0	0	1]
[0	0	0	0	385	0	2	0	0	0]
[0	0	0	0	0	376	0	0	0	0]
[0	1	0	0	0	0	374	0	2	0]
[0	0	0	1	0	0	0	385	0	1]
[0	1	0	0	0	0	0	0	379	0]
[0	0	0	0	1	1	0	0	1	379]

Step7: Class accuracies

	precision	recall	f1-score	support
0	1.00	0.99	1.00	376
1	0.99	0.99	0.99	389
2	1.00	0.99	1.00	380
3	0.99	0.98	0.99	389
4	0.99	0.99	0.99	387
5	0.98	1.00	0.99	376
6	0.99	0.99	0.99	377
7	1.00	0.99	1.00	387
8	0.99	1.00	0.99	380
9	0.99	0.99	0.99	382

cross-entropy error:

Step1: do experiment with different values of parameters:

1) number of hidden layers,

- 2) number of hidden units in each layer,
- 3) learning rates,
- 4) momentum rates,
- 5) epochs

	Results				
Hidden	Units	Epochs	Learning	Momentum	Loss & Accuracy
layer			rates	rates	(train data)
2	100/50	50	0.01	0.90	Loss: 0.1312
					Acc: 0.9623
2	100/50	100	0.01	0.90	Loss: 0.0864
					Acc: 0.9746
2	100/50	100	0.02	0.95	Loss: 0.0536
					Acc: 0.9847
2	70/35	100	0.02	0.95	Loss: 0.0443
					Acc: 0.9892
3	100/50/25	50	0.02	0.95	Loss: 0.0723
					Acc: 0.9772
3	100/50/25	100	0.02	0.95	Loss: 0.0493
					Acc: 0.9840

Step3: Best parameters config for •Sum-of-squares error function

Config	Results
Num of hidden layers = 3	Execution time: 5s
Units each layer = 100/50/25	Acc(train): .0.9840
Learning rates = 0.02	Loss(train): 0.0493
Momentum rates = 0.95	Acc(test): 0.9321
Epochs = 100	Loss(test): 0.2391

Step4: Confusion matrix for train data

[37	4	0	0	0	1	0	1	0	0	0]
[0	384	0	0	0	0	0	1	1	3]
[0	1	377	0	0	0	1	1	0	0]
[0	1	2	379	0	1	0	5	1	0]
[0	0	0	0	385	0	1	0	0	1]
[1	0	1	3	1	368	0	0	0	2]
[0	0	0	0	0	0	377	0	0	0]
[0	0	0	0	0	0	0	387	0	0]
[0	3	1	0	0	0	0	0	376	0]
[0	1	0	8	1	0	0	15	2	355]

Step5: Class accuracies

	precision	recall	f1-score	support
0	1.00	0.99	1.00	376
1	0.98	0.99	0.99	389
2	0.99	0.99	0.99	380
3	0.97	0.97	0.97	389
4	0.99	0.99	0.99	387
5	1.00	0.98	0.99	376
6	0.99	1.00	1.00	377
7	0.95	1.00	0.97	387
8	0.99	0.99	0.99	380
9	0.98	0.93	0.96	382

Step6: Confusion matrix for test data

[37	4	0	0	0	1	0	1	0	0	0]
[0	372	0	0	12	0	4	1	0	0]

```
1 374
               0
                       0
                                1
                                    0
                                        0]
                   0
                                        0]
           0 380
  0
           0
               0 386
                       0
                               0
                                    0
                                        0]
                   0 376
                                        0]
               0 13
                       0 364
                                        0]
      0
           0
                                0
              1
                   1
                       0
                           0 385
                                        0]
[ 10
       5
           1
              2
                   5
                       2
                           9
                               0 346
                                        0]
[ 2
      1
           0
                   5
                       7
                           0
                               3
                                   1 356]
```

Step7: Class accuracies

	precision	recall	f1-score	support
0	0.97	0.99	0.98	376
-				
1	0.98	0.96	0.97	389
2	1.00	0.98	0.99	380
3	0.97	0.98	0.98	389
4	0.91	1.00	0.95	387
5	0.96	1.00	0.98	376
6	0.95	0.97	0.96	377
7	0.98	0.99	0.99	387
8	1.00	0.91	0.95	380
9	1.00	0.93	0.96	382

Compare results:

The performance of the cross-entropy loss function is much better than the mse loss function under the same number of epochs. For the similar performance, the cross-entropy loss function needs less epochs. The cross-entropy loss function is suitable for the multiple-class classification.

Task1 (b)

tanh vs relu hidden units

tanh:

Step1: do experiment with different values of parameters:

- 1) number of hidden layers,
- 2) number of hidden units in each layer,
- 3) learning rates,
- 4) momentum rates,
- 5) epochs

	Parameters						
Hidden	Units	Epochs	Learning	Momentum	Loss & Accuracy		
layer			rates	rates	(train data)		
2	100/50	20	0.01	0.90	Loss: 0.2546		
					Acc: 0.9357		
2	100/50	50	0.01	0.90	Loss: 0.1200		
					Acc: 0.9689		
2	100/50	50	0.02	0.95	Loss: 0.0832		
					Acc: 0.9741		
2	70/35	50	0.02	0.95	Loss: 0.0785		
					Acc: 0.9778		

3	100/50/25	20	0.02	0.95	Loss: 0.1132
					Acc: 0.9623
3	100/50/25	50	0.02	0.95	Loss: 0.0492
					Acc: 0.9838

Step3: Best parameters config for •Sum-of-squares error function

Config	Results
Num of hidden layers = 3	Execution time: 3s
Units each layer = 100/50/25	Acc(train): .0.9838
Learning rates = 0.02	Loss(train): 0.0492
Momentum rates = 0.95	Acc(test): 0.9543
Epochs = 50	Loss(test): 0.1624

Step4: Confusion matrix for train data

[3	74	0	0	0	1	1	0	0	0	0]
[0	379	0	0	0	0	0	0	8	2]
[0	1	371	2	0	2	1	1	1	1]
[0	1	0	380	0	4	0	0	0	4]
[0	0	0	0	380	0	2	0	3	2]
[0	0	1	0	0	374	0	0	0	1]
[0	1	0	0	1	0	374	0	1	0]
[0	0	0	2	0	1	0	377	1	6]
[0	3	0	0	0	1	0	0	376	0]
[0	0	0	1	2	1	0	0	2	376]

Step5: Class accuracies

	precision	recall	f1-score	support
0	1.00	0.99	1.00	376
1	0.98	0.97	0.98	389
2	1.00	0.98	0.99	380
3	0.99	0.98	0.98	389
4	0.99	0.98	0.99	387
5	0.97	0.99	0.98	376

6	0.99	0.99	0.99	377
7	1.00	0.97	0.99	387
8	0.96	0.99	0.97	380
9	0.96	0.98	0.97	382

Step6: Confusion matrix for test data

[3	76	0	0	0	0	0	0	0	0	0]
[0	377	0	0	0	0	4	1	5	2]
[0	0	379	0	0	0	1	0	0	0]
[0	1	0	385	0	2	0	0	0	1]
[1	0	0	0	383	0	3	0	0	0]
[0	0	1	1	0	372	0	0	0	2]
[1	0	0	0	0	0	376	0	0	0]
[0	0	0	1	0	0	0	385	0	1]
[0	1	0	0	1	0	1	0	377	0]
[0	0	0	4	2	1	0	1	2	372]

Step7: Class accuracies

	precision	recall	f1-score	support
0	0.00	1 00	1 00	27.6
0	0.99	1.00	1.00	376
1	0.99	0.97	0.98	389
2	1.00	1.00	1.00	380
3	0.98	0.99	0.99	389
4	0.99	0.99	0.99	387
5	0.99	0.99	0.99	376
6	0.98	1.00	0.99	377
7	0.99	0.99	0.99	387
8	0.98	0.99	0.99	380
9	0.98	0.97	0.98	382

relu:

Step1: do experiment with different values of parameters:

- 1) number of hidden layers,
- 2) number of hidden units in each layer,

- 3) learning rates,
- 4) momentum rates,
- 5) epochs

Step2:

		Paramete	rs		Results
Hidden	Units	Epochs	Learning	Momentum	Loss & Accuracy
layer			rates	rates	(train data)
2	100/50	20	0.01	0.90	Loss: 0.3054
					Acc: 0.9189
2	100/50	50	0.01	0.90	Loss: 0.1312
					Acc: 0.9623
2	100/50	50	0.02	0.95	Loss: 0.0979
					Acc: 0.9689
2	70/35	50	0.02	0.95	Loss: 0.1427
					Acc: 0.9610
3	100/50/25	20	0.02	0.95	Loss: 0.1759
					Acc: 0.9456
3	100/50/25	50	0.02	0.95	Loss: 0.1144
					Acc: 0.9597

Step3: Best parameters config for •Sum-of-squares error function

Config	Res	sults
--------	-----	-------

Num of hidden layers = 3	Execution time: 3s
Units each layer = 100/50/25	Acc(train): .0.9597
Learning rates = 0.02	Loss(train): 0.1144
Momentum rates = 0.95	Acc(test): 0.9232
Epochs = 50	Loss(test): 0.3124

Step4: Confusion matrix for train data

[3	75	0	0	0	0	0	0	0	1	0]
[0	304	4	1	1	0	16	3	55	5]
[1	0	378	0	0	0	1	0	0	0]
[0	0	1	377	0	5	0	1	2	3]
[0	0	0	0	365	0	11	0	5	6]
[0	0	2	1	0	371	0	0	0	2]
[0	0	0	0	0	0	376	0	1	0]
[0	0	0	1	0	0	0	385	0	1]
[1	1	1	0	0	0	0	0	377	0]
[0	0	1	5	1	1	0	7	6	361]

Step5: Class accuracies

	precision	recall	f1-score	support
0	0.99	1.00	1.00	376
1	1.00	0.78	0.88	389
2	0.98	0.99	0.99	380
3	0.98	0.97	0.97	389
4	0.99	0.94	0.97	387
5	0.98	0.99	0.99	376
6	0.93	1.00	0.96	377
7	0.97	0.99	0.98	387
8	0.84	0.99	0.91	380
9	0.96	0.95	0.95	382

Step6: Confusion matrix for test data

[3	64	0	2	0	1	4	0	0	1	4]
[0	368	0	3	1	0	0	0	11	6]
[1	3	363	9	0	0	0	0	4	0]
[0	1	0	379	0	4	0	0	0	5]
Γ	1	2	Ο	Ο	372	0	3	0	1	81

```
[ 0
          1
              0 370
                     0
                        0
                              5]
                              0]
                  0 363
                       0
     0 0 5 0
                  3
                     0 364
                           0 15]
    1 2 1 0
                 1
[ 0
     0
        0 7
                 1
                     0
                        0
                           0 373]
              1
```

Step7: Class accuracies

	precision	recall	f1-score	support
0	0.99	0.97	0.98	376
1	0.97	0.95	0.96	389
2	0.99	0.96	0.97	380
3	0.94	0.97	0.95	389
4	0.99	0.96	0.98	387
5	0.97	0.98	0.97	376
6	0.99	0.96	0.98	377
7	1.00	0.94	0.97	387
8	0.95	0.97	0.96	380
9	0.88	0.98	0.93	382

Compare the results:

The performance of tanh is slightly better than relu for the same setting. The runtime is similar for the two activations.

Task2:

Convolutional networks

Step1: do experiment with different values of parameters:

- 1) number of convolutional layers,
- 2) number of filters,
- 3) kernel sizes,
- 4) strides,

5) epochs

		Parameters			Results
Convolutional	filters	Epochs	Kernel	strides	Loss &
layer			sizes		Accuracy
					(train data)
1	32	20	3	1	Loss: 0.9020
					Acc: 0.8837
1	32	50	3	1	Loss: 0.2764
					Acc: 0.9406
1	16	50	3	1	Loss: 0.3792
					Acc: 0.9233
1	32	50	5	1	Loss: 0.2638
					Acc: 0.9395
1	32	50	3	2	Loss: 0.8516
					Acc: 0.8093
2	32/64	20	3	1	Loss: 0.5201
					Acc: 0.8843
2	32/64	50	3	1	Loss: 0.1984
					Acc: 0.9476
2	32/64	50	5	1	Loss: 0.1553
					Acc: 0.9626

2	16/32	50	3	1	Loss: 0.3253
					Acc: 0.9147
2	32/64	50	7	1	Loss: 0.1281
					Acc: 0.9665

Step3: Best parameters config for •Sum-of-squares error function

Config	Results
Num of Convolutional layers= 2	Execution time: 62s
filters= 32/64	Acc(train): .0.9665
kernel size = 7	Loss(train): 0.1281
strides = 1	Acc(test): 0.9388
Epochs = 50	Loss(test): 0.2011

Step4: Confusion matrix for train data

[3	71	0	0	0	3	0	1	0	1	0]
[0	375	1	0	0	0	1	2	6	4]
[0	3	373	0	0	1	1	2	0	0]
[0	2	1	376	0	5	0	1	2	2]
[0	0	0	0	373	0	9	0	0	5]
[0	0	2	1	1	361	0	0	0	11]
[0	2	0	0	1	0	374	0	0	0]
[1	2	0	1	1	0	0	379	1	2]
[0	21	1	0	2	0	0	2	354	0]
[0	3	0	8	6	1	0	4	1	359]

Step5: Class accuracies

	precision	recall	f1-score	support
0	1.00	0.99	0.99	376
1	0.92	0.96	0.94	389
2	0.99	0.98	0.98	380
3	0.97	0.97	0.97	389
4	0.96	0.96	0.96	387
5	0.98	0.96	0.97	376

6	0.97	0.99	0.98	377
7	0.97	0.98	0.98	387
8	0.97	0.93	0.95	380
9	0.94	0.94	0.94	382

Step6: Confusion matrix for test data

[3	71	0	0	0	4	0	1	0	0	0]
[0	373	2	0	0	0	0	2	8	4]
[0	3	370	2	0	1	1	2	1	0]
[0	0	2	378	0	5	0	0	2	2]
[0	0	0	0	372	0	9	0	1	5]
[0	0	1	2	0	367	0	0	1	5]
[0	2	0	0	1	0	374	0	0	0]
[0	1	0	1	1	0	0	379	1	4]
[0	18	4	3	1	2	0	1	350	1]
[0	2	0	10	3	1	0	5	2	359]

Step7: Class accuracies

	precision	recall	f1-score	support
0	1.00	0.99	0.99	376
1	0.93	0.96	0.95	389
2	0.98	0.97	0.97	380
3	0.95	0.97	0.96	389
4	0.97	0.96	0.97	387
5	0.98	0.98	0.98	376
6	0.97	0.99	0.98	377
7	0.97	0.98	0.98	387
8	0.96	0.92	0.94	380
9	0.94	0.94	0.94	382

Compare the results:

Filters: the more filters in a certain range, the better the performance

Kernel size: the larger the kernel size in a certain range, the better the performance

Convolutional layers: the more convolutional layers in a certain range, the better the performance.

Stride: the smaller the stride, the better the performance.

Epochs: the more the epochs, the better the performance.