# **Prototypical Networks:**

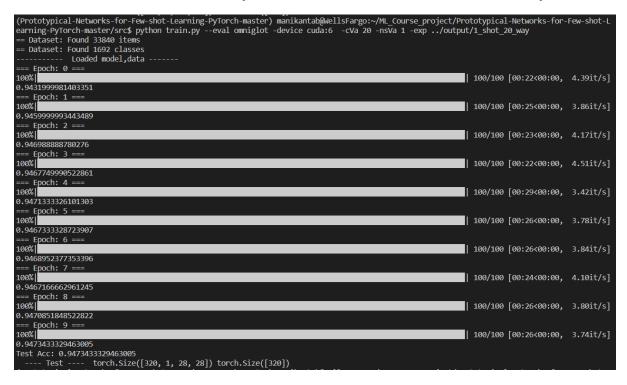
# Omniglot (1-shot-5-way):

```
=== Epoch: 1 ===
100%|
0.98273333378791809
                                                                                                  | 100/100 [00:07<00:00, 12.71it/s]
=== Epoch: 2 ==
100%|
                                                                                                  | 100/100 [00:08<00:00, 11.87it/s]
0.9812888936201731
=== Epoch: 3 ===
100%| 0.9811000046133995
                                                                                                  | 100/100 [00:06<00:00, 15.52it/s]
0.9811000046133995
=== Epoch: 4 ===
100%
0.9813333381414413
=== Epoch: 5 ===
100%
0.9807777825991313
                                                                                                  | 100/100 [00:06<00:00, 14.40it/s]
                                                                                                  | 100/100 [00:08<00:00, 11.26it/s]
  == Epoch: 6 =
100%|
0.9817714335237231
                                                                                                  | 100/100 [00:08<00:00, 11.83it/s]
=== Epoch: 7 =
100%|
                                                                                                  | 100/100 [00:10<00:00, 9.53it/s]
0.9817000047117471
=== Epoch: 8 ===
100%
0.9819407454464171
                                                                                                  | 100/100 [00:08<00:00, 11.12it/s]
=== Epoch: 9 ===
100%|
0.9820133380889893
                                                                                                 | 100/100 [00:09<00:00, 11.06it/s]
Test Acc: 0.9820133380889893
                   torch.Size([80, 1, 28, 28]) torch.Size([80])
```

## Omniglot (5-shot-5-way):

```
=== Epoch: 0 ==
100%|
                                                                                                      | 100/100 [00:07<00:00, 12.68it/s]
0.9932000035047531
=== Epoch: 1 ===
100%|
0.9939333361387253
                                                                                                      | 100/100 [00:06<00:00, 16.63it/s]
=== Epoch: 2 ===
100%|
0.994444446961085
                                                                                                      | 100/100 [00:08<00:00, 12.44it/s]
=== Epoch: 3 ===
100% 0.9945000027120113 === Epoch: 4 ===
                                                                                                      | 100/100 [00:07<00:00, 12.79it/s]
100%| 0.994880002617836
                                                                                                      | 100/100 [00:07<00:00, 14.16it/s]
0.9951333359877268
=== Epoch: 6 ===
                                                                                                      | 100/100 [00:08<00:00, 11.86it/s]
                                                                                                      | 100/100 [00:08<00:00, 11.58it/s]
100%|
0.9952571455069951
=== Epoch: 7 ===
100%|
0.9954000027477741
                                                                                                     | 100/100 [00:05<00:00, 17.33it/s]
=== Epoch: 8 =
100%|
                                                                                                      | 100/100 [00:08<00:00, 11.55it/s]
0.9956000027391646
=== Epoch: 9 ===
| 100/100 [00:05<00:00, 17.26it/s]
```

### Omniglot (1-shot-20-way)



## Omniglot (5-shot-20-way)

```
== Epoch: 0 ==
100%|
0.9835333448648452
                                                                                                                                | 100/100 [00:33<00:00, 2.96it/s]
=== Epoch: 1 ==
100%|
                                                                                                                                 | 100/100 [00:30<00:00, 3.24it/s]
0.98325001090765
=== Epoch: 2 ===
100% 0.9834111217657725
                                                                                                                                 | 100/100 [00:30<00:00, 3.33it/s]
=== Epoch: 3 ==
100%|
                                                                                                                                 | 100/100 [00:27<00:00, 3.64it/s]
100%|

0.9839333441853523

=== Epoch: 4 ===

100%|

0.9836466773748398

=== Epoch: 5 ===

1000%|
                                                                                                                                | 100/100 [00:26<00:00, 3.71it/s]
=== Epoch: 5 ===
100%|
0.9837222332755725
=== Epoch: 6 ===
100%|
                                                                                                                                | 100/100 [00:22<00:00, 4.45it/s]
                                                                                                                                 | 100/100 [00:16<00:00, 6.11it/s]
0.9840523917334421
=== Epoch: 7 ===
=== Epoch: 7 ===
100% | 0.9840791773796081
=== Epoch: 8 ===
100% | 0.9840791773796081
                                                                                                                                | 100/100 [00:19<00:00, 5.11it/s]
                                                                                                                                 | 100/100 [00:17<00:00, 5.65it/s]
0.9842296403646469
=== Epoch: 9 ===
100%| 0.9843966774344445
                                                                                                                                | 100/100 [00:14<00:00, 6.84it/s]
v.964390077434444
Test Acc: 0.9843966774344445
---- Test ---- torch.Size([400, 1, 28, 28]) torch.Size([400])
```

## Adaptation of Omniglot to Mnist (5-shot-5-way)

```
=== Epoch: 0
100%|
                                                                                                          | 100/100 [00:10<00:00, 9.97it/s]
0.8990666687488555
=== Epoch: 1 =
100%|
                                                                                                          | 100/100 [00:08<00:00, 11.62it/s]
0.9001333364844322
  == Epoch: 2 ==
100%
                                                                                                          | 100/100 [00:09<00:00, 10.33it/s]
0.8989333365360895
                                                                                                          | 100/100 [00:09<00:00, 10.87it/s]
100%|
0.8987000036239624
=== Epoch: 4 ==
100%|
                                                                                                          | 100/100 [00:08<00:00, 12.12it/s]
0.8988266695737839
=== Epoch: 5 ==
100%|
                                                                                                          | 100/100 [00:08<00:00, 11.71it/s]
0.9002444471915563
=== Epoch: 6 ===
100%|
                                                                                                          | 100/100 [00:08<00:00, 11.54it/s]
0.8995428597075599
  == Epoch: 7 :
100%|
                                                                                                          | 100/100 [00:08<00:00, 11.85it/s]
0.9000000027567148
=== Epoch: 8 ===
100%|
0.8993037064870198
                                                                                                          | 100/100 [00:08<00:00, 12.09it/s]
=== Epoch: 9 =
                                                                                                          | 100/100 [00:08<00:00, 12.33it/s]
 0.9003066692352295
0.9003000092332295
---- Test ---- torch.Size([100, 1, 28, 28]) torch.Size([100])
(Prototypical-Networks-for-Few-shot-Learning-PyTorch-master) manikantab@WellsFargo:~/ML_Course_project/Prototypical-Networks-for-Few-shot-Learning-PyTorch-master/src$
```

## Adaptation of Omniglot to Mnist (1-shot-5-way)

```
(Prototypical-Networks-for-Few-shot-Learning-PyTorch-master) manikantab@WellsFargo:~/ML_Course_project/Prototypical-Networks-for-Few-shot-Learning-PyTorch-master/src$ python train.py --eval -device cuda:6 -cVa 5 -nsVa 1 -exp ../output/1_shot_5_way
=== Epoch: 0 ===
100%
                                                                                                                                          | 100/100 [00:09<00:00, 10.54it/s]
0.7486666685342789
=== Epoch: 1 ===
100%
                                                                                                                                          | 100/100 [00:10<00:00, 9.51it/s]
0.7514666682481765
 == Epoch: 2 ==
100%|
                                                                                                                                          | 100/100 [00:08<00:00, 11.74it/s]
100%|

0.7503555572032928

=== Epoch: 3 ===

100%|

0.7493000017851591
                                                                                                                                          | 100/100 [00:08<00:00, 11.84it/s]
=== Epoch: 4 ===
100%|
0.7473066682815552
                                                                                                                                          100/100 [00:07<00:00, 12.61it/s]
=== Epoch: 5 ===
100%|
0.7447777799268563
                                                                                                                                          | 100/100 [00:09<00:00, 10.37it/s]
=== Epoch: 6 ==
100%|
                                                                                                                                           | 100/100 [00:08<00:00, 11.46it/s]
0.7454285732337407
=== Epoch: 7 ===
                                                                                                                                          | 100/100 [00:08<00:00, 11.64it/s]
0.7450333350151778
=== Epoch: 8 ===
100%|
                                                                                                                                          | 100/100 [00:08<00:00, 12.22it/s]
0.7460888904333115
   = Epoch: 9 =
100%
                                                                                                                                          | 100/100 [00:09<00:00, 10.95it/s]
0.7452133347094059
Test Acc: 0.7452133347094059
---- Test ---- torch.Size([80, 1, 28, 28]) torch.Size([80])
```

## **MAML**: Omniglot dataset

#### 1-shot-5-way

```
(MAML-Pytorch) manikantab@wellsFargo:~/virtual_env/MAML-Pytorch$ python omniglot_train.py --eval omniglot --k_spt 1 --model _path models/omniglot/1_shot_5_way/best_model.pth
Namespace(epoch=40000, eval='omniglot', imgc=1, imgsz=28, k_qry=15, k_spt=1, meta_lr=0.001, model_path='models/omniglot/1_s
hot_5_way/best_model.pth', n_way=5, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, update_lr=0.4, update_step=5, update_step_test=10)
Loading Pre-trained model at path == models/omniglot/1_shot_5_way/best_model.pth
load from omniglot.npy.
DB: train (1200, 40, 1, 28, 28) test (423, 40, 1, 28, 28)
     -- Test -
x_spt.shape == torch.Size([32, 5, 1, 28, 28])
y_spt.shape == torch.Size([32, 5])
x_qry.shape == torch.Size([32, 75, 1, 28, 28])
y_qry.shape == torch.Size([32, 75])
step: 0 test acc: [0.2258 0.9307 0.9326 0.9326 0.933 0.933 0.933 0.933 0.933 0.933 0.933
 0.933 ]
                       test acc: [0.2103 0.932 0.9365 0.938 0.9385 0.9385 0.9385 0.939 0.939
step: 5
 0.939 ]
step: 10
                       test acc: [0.2063 0.9307 0.936 0.937 0.937 0.938 0.938 0.938 0.938 0.938
 0.9385]
                       test acc: [0.2059 0.933 0.939 0.94 0.9404 0.9404 0.9404 0.941 0.941 0.941
step: 15
 0.941 ]
                       test acc: [0.2062 0.9316 0.938 0.939 0.9395 0.9395 0.9395 0.94 0.94 0.94
step: 20
 0.94041
                       step: 25
 0.94 ]
step: 30
                       test acc: [0.2058 0.932 0.938 0.939 0.9395 0.9395 0.94 0.94 0.94 0.9404
 0.9404]
Test Accuracy === [0.2058 0.932 0.938 0.939 0.9395 0.9395 0.94 0.94 0.94 0.9404
```

#### 5-shot-5-way

#### Trained on 5 classes\*5 examples and tested on 5 classes\*15 examples

```
(MAML-Pytorch) manikantab@WellsFargo:∼/virtual env/MAML-Pytorch$ python omniglot train.py --eval omniglot --k spt 5 --n way
CMWML-Pytorch; marikantangweiisrargo: //irtua_env/mami-Pytorch; python omnigiot_train.py --eval omnigiot --k_spt 5 --h_way 5 --model_path models/omnigiot/5_shot_5_way/best_model.pth

Namespace(epoch=40000, eval='omnigiot', imgc=1, imgsz=28, k_qry=15, k_spt=5, meta_lr=0.001, model_path='models/omnigiot/5_shot_5_way/best_model.pth', n_way=5, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, update_lr=0.4, update_step=5, update_step_test=10)

Loading Pre-trained model at path == models/omniglot/5_shot_5_way/best_model.pth
 load from omniglot.npy.
DB: train (1200, 40, 1, 28, 28) test (423, 40, 1, 28, 28)
        Test
x_spt.shape == torch.Size([32, 25, 1, 28, 28])
y_spt.shape == torch.Size([32, 25])
x_qry.shape == torch.Size([32, 75, 1, 28, 28])
y_qry.shape == torch.Size([32, 75])
                      test acc: [0.2308 0.9814 0.982 0.982 0.982 0.9814 0.9814 0.9814 0.9814 0.9814
 step: 0
 0.9814]
                      step: 5
 0.981 ]
step: 10
                      test acc: [0.203  0.9814  0.982  0.982  0.982  0.982  0.982  0.982  0.982  0.982
0.982 ]
step: 15
                      test acc: [0.2013 0.9805 0.981 0.9814 0.9814 0.9814 0.9814 0.9814 0.9814 0.9814
 0.9814]
 step: 20
                       test acc: [0.2029 0.981 0.9814 0.982 0.982 0.982 0.982 0.982 0.982 0.982
 0.9824]
step: 25
                      test acc: [0.2002 0.981 0.982 0.9824 0.9824 0.9824 0.9824 0.9824 0.9824 0.9824 0.9824
 0.9824]
                      test acc: [0.2019 0.9814 0.9824 0.9824 0.983 0.983 0.983 0.983 0.983 0.983
step: 30
Test Accuracy === [0.2019 0.9814 0.9824 0.9824 0.983 0.983 0.983 0.983 0.983 0.983
 0.983 ]
```

#### 1-shot-20-way

```
(MAML-Pytorch) manikantab@WellsFargo:~/virtual_env/MAML-Pytorch$ python omniglot_train.py --eval omniglot --k spt 1 --n way
CHAMLE-Pyton Chin manikantangwellsrango. "Virtual_env/Manie-Pyton Chin python Chininglot_chall.py "-eval Chininglot --k_spt 1 --h_way 20 --model_path models/omniglot/1_shot_20_way/best_model.pth

Namespace(epoch=40000, eval='omniglot', imgc=1, imgsz=28, k_qry=15, k_spt=1, meta_lr=0.001, model_path='models/omniglot/1_s hot_20_way/best_model.pth', n_way=20, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, upda te_lr=0.4, update_step=5, update_step_test=10)

Loading Pre-trained model at path == models/omniglot/1_shot_20_way/best_model.pth
load from omniglot.npy.
DB: train (1200, 40, 1, 28, 28) test (423, 40, 1, 28, 28)
x_spt.shape == torch.Size([32, 20, 1, 28, 28])
y_spt.shape == torch.Size([32, 20])
x_qry.shape == torch.Size([32, 300, 1, 28, 28])
y_qry.shape == torch.Size([32, 300])
y_qry.shape ==
step: 0
                     test acc: [0.05176 0.792
                                                      0.812
                                                                 0.812 0.8125 0.8125 0.8125 0.8135 0.8135
 0.8135 0.8135 ]
                      .
test acc: [0.05313 0.8135 0.838 0.8384 0.8384 0.839 0.839 0.8394 0.8394
 0.8394 0.8394
                     step: 10
 0.8364 0.8364 ]
step: 15
                     test acc: [0.0504 0.8076 0.832 0.8325 0.8325 0.833 0.833 0.833 0.8335 0.8335
 0.8335]
                     test acc: [0.04977 0.808 0.833 0.834 0.834 0.834 0.8345 0.8345 0.8345
step: 20
 0.835 0.835
step: 25
                     0.834 0.834
step: 30
                     test acc: [0.05127 0.807 0.832
                                                                 0.833
                                                                           0.833 0.8335 0.8335 0.8335 0.8335
 0.834 0.834
Test Accuracy = 0.834 0.834
                        [0.05127 0.807 0.832 0.833 0.833 0.8335 0.8335 0.8335 0.8335
```

#### 5-shot-20-way

```
(MAML-Pytorch) manikantab@wellsFargo:~/virtual_env/MAML-Pytorch$ python omniglot_train.py --eval omniglot --k_spt 5 --n_way 20 --model_path models/omniglot/5_shot_20_way/best_model.pth

Namespace(epoch=40000, eval='omniglot', imgc=1, imgsz=28, k_qry=15, k_spt=5, meta_lr=0.001, model_path='models/omniglot/5_s hot_20_way/best_model.pth', n_way=20, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, update_lr=0.4, update_step=5, update_step_test=10)

Loading Pre-trained model at path == models/omniglot/5_shot_20_way/best_model.pth
load from omniglot.npy.
DB: train (1200, 40, 1, 28, 28) test (423, 40, 1, 28, 28)
    -- Test --
0.9263]
step: 5
                    test acc: [0.05365 0.908 0.9204 0.9224 0.923 0.9233 0.924 0.9243 0.925
 0.9253 0.9253 ]
 step: 10
                    0.9277 0.928
step: 15
                    test acc: [0.05127 0.912 0.924 0.9253 0.9263 0.927 0.9272 0.9277 0.9277
0.928 0.928
step: 20
                    test acc: [0.05054 0.9116 0.923 0.9243 0.9253 0.926 0.9263 0.927 0.927
 0.9272 0.9272 ]
step: 25
                    test acc: [0.0508 0.912 0.923 0.925 0.926 0.9263 0.927 0.927 0.9272 0.9272
                    test acc: [0.05066 0.912 0.9233 0.925 0.926 0.9263 0.927 0.9272 0.9272
 0.9277 0.9277 ]
                       [0.05066 0.912 0.9233 0.925 0.926 0.9263 0.927 0.9272 0.9272
Test Accuracy ==
 0.9277 0.9277 ]
```

### Adaptation of Omniglot to Mnist (1-shot-5-way)

```
(MAML-Pytorch) manikantab@wellsFargo:~/virtual_env/MAML-Pytorch$ python omniglot_train.py --eval mnist --k_spt 1 --model_pa th models/omniglot/1_shot_5_way/best_model.pth

Namespace(epoch=40000, eval='mnist', imgc=1, imgsz=28, k_qry=15, k_spt=1, meta_lr=0.001, model_path='models/omniglot/1_shot_5_way/best_model.pth', n_way=5, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, update_lr=0.4, update_step=5, update_step_test=10)

Loading Pre-trained model at path == models/omniglot/1_shot_5_way/best_model.pth
 (5421, 1, 28, 28)
(5923, 1, 28, 28)
(5923, 1, 28, 28)
(5842, 1, 28, 28)
(6742, 1, 28, 28)
(5949, 1, 28, 28)
(5958, 1, 28, 28)
(6131, 1, 28, 28)
(5918, 1, 28, 28)
 (6265, 1, 28, 28)
(5851, 1, 28, 28)
data shape: (10, 5421, 1, 28, 28)
DB: test (10, 5421, 1, 28, 28)
       -- Test -
x_spt.shape == torch.size([32, 5, 1, 28, 28])
y_spt.shape == torch.size([32, 5])
x_qry.shape == torch.size([32, 75, 1, 28, 28])
y_qry.shape == torch.size([32, 75])
step: 0 test acc: [0.1821 0.5625 0.579 0.5796 0.5806 0.58 0.5786 0.579 0.5796 0.5996
 0.5806]
step: 5 0.583 ]
                           test acc: [0.1989 0.5547 0.576 0.579 0.58 0.581 0.581 0.5815 0.5825 0.5825
 step: 10
                           0.58 ]
step: 15
                           test acc: [0.1971 0.548 0.575 0.578 0.579 0.5796 0.58 0.5806 0.581 0.5815
  0.582 ]
                           test acc: [0.1973 0.547 0.5737 0.576 0.577 0.5776 0.5786 0.579 0.579 0.5796
 step: 20
0.58 ]
step: 25
                           test acc: [0.1975 0.548 0.575 0.577 0.578 0.5786 0.5796 0.58 0.5806 0.5806
  0.581 ]
 step: 30
                           test acc: [0.1967 0.548 0.5747 0.5767 0.5776 0.578 0.579 0.5796 0.5796 0.58
  0.5806]
 Test Accuracy === [0.1967 0.548 0.5747 0.5767 0.5776 0.578 0.579 0.5796 0.5796 0.58
  0.5806]
```

### Adaptation of Omniglot to Mnist (5-shot-5-way)

### Trained on 5 classes\*5 examples and tested on 5 classes\*15 examples

```
torch) manikantab@WellsFargo:~/virtual_env/MAML-Pytorch$ python omniglot_train.py --eval mnist --k_spt 5 --n_way 5
-model_path models/omniglot/5_shot_5_way/best_model.pth
Namespace(epoch=40000, eval='mnist', imgc=1, imgsz=28, k_qry=15, k_spt=5, meta_lr=0.001, model_path='models/omniglot/5_shot_5_way/best_model.pth', n_way=5, save_path='/data4/home/manikantab/virtual_env/MAML-Pytorch/models', task_num=32, update_lr=0.4, update_step=5, update_step_test=10)
Loading Pre-trained model at path == models/omniglot/5_shot_5_way/best_model.pth
 (5421, 1, 28, 28)
(5923, 1, 28, 28)
 (5842, 1, 28, 28)
(6742, 1, 28, 28)
 (5949, 1, 28, 28)
 (5958, 1, 28, 28)
 (6131, 1, 28, 28)
  (5918, 1, 28, 28)
 (6265, 1, 28, 28)
(5851, 1, 28, 28)
data shape: (10, 5421, 1, 28, 28)
DB: test (10, 5421, 1, 28, 28)
---- Test ----
x_spt.shape == torch.Size([32, 25, 1, 28, 28])
y_spt.shape == torch.Size([32, 25])
x_qry.shape == torch.Size([32, 75, 1, 28, 28])
y_qry.shape == torch.Size([32, 75, 1, 28, 28])
step: 0 torch.Size([32, 75])
step: 0 test acc: [0.212  0.7627  0.787  0.788  0.789  0.789  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.7896  0.
  0.79 ]
step: 5
                                               test acc: [0.2009 0.773 0.7915 0.7925 0.793 0.794 0.7944 0.795 0.7954 0.7954
   0.7954]
 step: 10
                                               test acc: [0.1997 0.7686 0.786 0.787 0.7876 0.788 0.7886 0.789 0.7896 0.79
0.79 ] step: 15
                                               test acc: [0.1996 0.769 0.7866 0.7876 0.788 0.7886 0.789 0.7896 0.79 0.79
  0.7905]
                                                test acc: [0.2     0.7705     0.7876     0.7886     0.789     0.7896     0.79     0.7905     0.791     0.7915
 step: 20
   0.7915]
 step: 25
                                               test acc: [0.1997 0.769 0.787 0.788 0.789 0.7896 0.79 0.7905 0.7905 0.791
  0.7915]
                                               test acc: [0.1996 0.77  0.788  0.789  0.7896  0.79  0.7905  0.791  0.7915  0.792
step: 30
  0.792 ]
 Test Accuracy === [0.1996 0.77 0.788 0.789 0.7896 0.79 0.7905 0.791 0.7915 0.792
   0.792 ]
```

#### Adaptation Omniglot to Mnist (1-shot-20-way)

#### Trained on 20 classes\*1 examples and tested on 20 classes\*15 examples

 Since the MNIST dataset has only 10 classes, it was not possible to evaluate 1-shot 20-way model

#### Adaptation Omniglot to Mnist (5-shot-20-way)

#### Trained on 20 classes\*5 examples and tested on 20 classes\*15 examples

 Since the MNIST dataset has only 10 classes, it was not possible to evaluate 5-shot 20-way model

# MAML: MiniImageNet

#### 1\_shot\_5\_way:

- update\_step = 5
- update\_step\_test = 15
- best model path

```
Val acc: [0.1912 0.405 0.4097 0.412 0.412 0.4146 0.4148 0.416 0.416 0.417
   0.4167 0.4167 0.417 0.4172 0.4182 0.4185]
     ----- Saving the best model -----
                                          step: 2010
step: 2040
step: 2070
                                       step: 2100
                                      step: 2130
step: 2160
step: 2190
step: 2220
                                       training acc: [0.26666667 0.59 0.59333333 0.59 0.59666667 0.59333333
step: 2250
                                       training acc: [0.16666667 0.516666667 0.53333333 0.53666667 0.53333333 0.52666667] training acc: [0.25333333 0.56666667 0.58333333 0.58 0.59 0.59333333] training acc: [0.16333333 0.59333333 0.61 0.61333333 0.60666667 0.61333333]
step: 2280
step: 2310
step: 2340
                                       training acc: [0.24666667 0.6 0.58 0.58 0.57333333 0.57666667] training acc: [0.13666667 0.51666667 0.51 0.51333333 0.51 0.51333333] training acc: [0.23 0.60333333 0.62333333 0.62666667 0.62333333 0.62333333] training acc: [0.18333333 0.486666667 0.51666667 0.52333333 0.53333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.5333333 0.533333 0.5333333 0.5333333 0.5333333 0.533333 0.5333333 0.5333333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.53333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.5333333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.5333333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.533333 0.53333 0.5333
step: 2370
step: 2400
step: 2430
step: 2460
step: 2490
                                         training acc: [0.13333333 0.576666667 0.56666667 0.57333333 0.57333333 0.57666667]
Test acc: [0.2107 0.4348 0.4348 0.4353 0.4353 0.438 0.4385 0.4385 0.439 0.44
  0.4397 0.4397 0.4404 0.4404 0.4404 0.4402]
```

## 5\_shot\_5\_way:

python miniimagenet\_train.py --device cuda:7 --update\_step 5 --update\_step\_test 15 --k\_spt 5 -- save path models/5 shot 5 way

```
Val acc: [0.2139 0.5396 0.553 0.561 0.5625 0.5625 0.5625 0.564 0.5645 0.5635
0.564 0.5645 0.564 0.565 0.565 0.565 ]
  ----- Saving the best model -----
step: 2010
             training acc: [0.20333333 0.63666667 0.63333333 0.67
                                                                0.68333333 0.683333333]
step: 2040
             training acc: [0.21333333 0.76333333 0.79333333 0.79333333 0.78
                                                                          0.78666667
             training acc: [0.11666667 0.69333333 0.71 0.72666667 0.73333333 0.74 ]
training acc: [0.24333333 0.69 0.72333333 0.73 0.72333333 0.72333333]
training acc: [0.22 0.64666667 0.64333333 0.65 0.64666667 0.65
step: 2070
step: 2100
step: 2130
            training acc: [0.18333333 0.71333333 0.74333333 0.74666667 0.75666667 0.76333333]
step: 2160
             training acc: [0.18666667 0.63666667 0.63 0.64666667 0.64333333 0.64
step: 2190
step: 2220
            0.71666667 0.71
step: 2250
            training acc: [0.25333333 0.66
                                                             0.71 0.71
0.71666667 0.72
                                                                          0.71666667
             training acc: [0.19333333 0.63666667 0.67333333 0.71
step: 2280
step: 2310
            step: 2340
             training acc: [0.14333333 0.75333333 0.74333333 0.76333333 0.75333333 0.75333333
             step: 2370
             step: 2400
step: 2430
             training acc: [0.21
                                    0.73333333 0.75666667 0.76333333 0.77333333 0.773333333
step: 2460
             training acc: [0.17333333 0.68666667 0.69666667 0.72
                                                                 0.72
                                                                           0.72666667
step: 2490 training acc: [0.15666667 0.7 0.71666667 0.75333333 0.75333333 0.75333333]
Test acc: [0.2064 0.5625 0.58 0.588 0.5884 0.5913 0.5923 0.5938 0.5938 0.594
0.5947 0.5947 0.594 0.5947 0.595 0.5957]
```

### 1\_shot\_20\_way:

python miniimagenet\_train.py --device cuda:7 --update\_step 5 --update\_step\_test 15 --k\_spt 1 - n way 20 --save path models/1 shot 20 way

■ Val set not considered(it has only 16 classes)

```
Val acc: [0.04764 0.1458 0.1637 0.167
                                     0.1664 0.167
                                                    0.1671 0.1674 0.1672
0.167 0.1671 0.1672 0.1675 0.1675 0.1674 0.1674 ] ------ Saving the best model
step: 2010
              training acc: [0.0475
                                     0.21666667 0.27833333 0.29166667 0.2925
                                                                             0.2925
step: 2040
              training acc:
                           [0.04
                                     0.27 0.31083333 0.30666667 0.3125
                                                                             0.305
                           [0.04916667 0.275
step: 2070
                                               0.29583333 0.295 0.29666667 0.29916667
              training acc:
            step: 2100
                                                                   0.27666667 0.27666667
step: 2130
                                                                             0.273333333
step: 2160
              training acc:
                           [0.05833333 0.23416667 0.275
                                                        0.28583333 0.28666667 0.28833333
             training acc: [0.05416667 0.2925 0.32583333 0.32833333 0.325
step: 2190
                                                                             0.3225
step: 2220
            training acc:
                           [0.04916667 0.25916667 0.29583333 0.31333333 0.3125
step: 2250
                           training acc:
step: 2280
              training acc:
                           0.06
                                     0.24583333 0.27
                                                         0.2725 0.27666667 0.2775
                           0.03083333 0.22166667 0.27416667 0.28416667 0.2875
step: 2310
              training acc:
                                                                             0.285
             training acc:
                           [0.05166667 0.26166667 0.31666667 0.32916667 0.32833333 0.33083333]
step: 2340
             training acc:
                                                                  0.3475
step: 2370
                           [0.04
                                    0.2825
                                             0.33833333 0.345
                                                                             0.34333333]
                           [0.05916667 0.2475
[0.05083333 0.2725
step: 2400
              training acc:
                                               0.30083333 0.30083333 0.30666667 0.2975
                                             0.31416667 0.3225
step: 2430
              training acc:
                                                                 0.315
                                                                             0.31
                           0.05333333 0.25416667 0.26916667 0.28
step: 2460
              training acc:
                                                                   0.28
                                                                             0.29
                                  0.28083333 0.3275 0.3225
step: 2490
              training acc: [0.065
                                                                   0.31916667 0.32333333]
Test acc: [0.05045 0.1433 0.1603 0.1647 0.165 0.1649 0.1644 0.1648 0.1649
0.165  0.1649  0.1652  0.1649  0.165  0.1653  0.1654 ]
```

#### 5\_shot\_20\_way:

python miniimagenet\_train.py --device cuda:7 --update\_step 5 --update\_step\_test 15 --k\_spt 5 -- n way 20 --save path models/5 shot 20 way

■ Val set not considered(it has only 16 classes)

```
0.2632 0.2632 0.2632 0.2637 0.2634 0.2634]
 ----- Saving the best model -----
step: 2010
              training acc: [0.0475
                                      0.37916667 0.45
                                                          0.465
                                                                     0.485
                                                                               0.49083333]
              training acc: [0.07416667 0.33333333 0.44583333 0.45916667 0.47833333 0.4775
step: 2040
              step: 2070
              training acc: [0.01416667 0.35083333 0.445 0.45916667 0.47416667 0.48 training acc: [0.03166667 0.36833333 0.4275 0.46666667 0.46416667 0.46416667
step: 2100
step: 2130
              training acc: [0.03166667 0.36833333 0.4275
step: 2160
                           [0.04916667 0.36083333 0.43666667 0.46166667 0.47166667 0.4975
              training acc:
step: 2190
              training acc: [0.04333333 0.36833333 0.45083333 0.475 0.47
                                                                             0.49083333
step: 2220
              0.51416667
step: 2250
              training acc: [0.06583333 0.37916667 0.465 0.47416667 0.4825
                                                                              0.49166667
step: 2280
              training acc:
                           [0.04333333 0.36166667 0.46333333 0.49833333 0.4975
                                                                               0.50666667
step: 2310
                           0.05333333 0.32583333 0.40666667 0.4375
                                                                   0.445
                                                                              0.455
              training acc:
step: 2340
              training acc: [0.05666667 0.3625 0.43 0.46833333 0.49916667 0.515
                                                          0.43083333 0.44
step: 2370
              training acc: [0.0625
                                   0.295
                                               0.41
                                                                               0.44583333
              training acc: [0.05916667 0.33416667 0.41416667 0.42583333 0.45583333 0.4725 training acc: [0.06166667 0.34416667 0.436666667 0.466166667 0.46916667 0.47416667
step: 2400
step: 2430
              training acc: [0.01916667 0.36333333 0.4675 0.49916667 0.50916667 0.5125
step: 2460
step: 2490
              training acc: [0.0625 0.33416667 0.42666667 0.43916667 0.46
                                                                               0.46166667]
Test acc: [0.0474 0.1825 0.2369 0.2512 0.2605 0.2605 0.2637 0.264 0.264 0.264
0.2646 0.2644 0.2642 0.2644 0.2646 0.2646]
```