```
from google.colab import drive
drive.mount('/content/drive')
   → Mounted at /content/drive
!git clone https://github.com/VICO-UoE/DatasetCondensation
%cd DatasetCondensation
  ₹ Cloning into 'DatasetCondensation'...
                  remote: Enumerating objects: 271, done.
                  remote: Counting objects: 100% (138/138), done.
                  remote: Compressing objects: 100% (64/64), done.
                  remote: Total 271 (delta 126), reused 74 (delta 74), pack-reused 133 (from 1)
                  Receiving objects: 100% (271/271), 4.54 MiB | 19.53 MiB/s, done.
                  Resolving deltas: 100% (152/152), done.
                  /content/DatasetCondensation
!pip install -r requirements.txt
   → Collecting numpy=1.15.1 (from -r requirements.txt (line 1))
                        Downloading numpy-1.15.1.zip (4.5 MB)
                                                                                                                                                                                                                                                                                 4.5/4.5 MB 15.1 MB/s eta 0:00:00
                       Preparing metadata (setup.py) ... done
                 Collecting scipy==1.1.0 (from -r requirements.txt (line 2))
                       Downloading scipy-1.1.0.tar.gz (15.6 MB)
                                                                                                                                                                                                                                                                                          -- 15.6/15.6 MB 100.6 MB/s eta 0:00:00
                       Preparing metadata (setup.pv) ... done
                  ERROR: Ignored the following versions that require a different python version: 1.6.2 Requires-Python >= 3.7, \langle 3.10; 1.6.3 Requires-Python >= 3.7, \langle 3.10; 1.7.0 Requires-Python >= 3.
                  ERROR: Could not find a version that satisfies the requirement torch=1.2.0 (from versions: 1.11.0, 1.12.0, 1.12.1, 1.13.0, 1.13.1, 2.0.0, 2.0.1, 2.1.0, 2.1.1, 2.1.2, 2.2.0, 2.2.1, 2.2.2, 2.3.0, 2.3.1, 2.4.0, 2.4.1, 2.5.0, 2.5.1)
                  ERROR: No matching distribution found for torch==1.2.0
from torchvision, datasets import MNIST
from torchvision import transforms
# download MNIST dataset
mnist_data = MNIST(root='./data', train=True, download=True, transform=transforms.ToTensor())
  Downloading <a href="http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz">http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz</a>
                  Failed to download (trying next):
                  <urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certificate has expired (_ssl.c:1007)>
                  Downloading \ \underline{https://ossci-datasets.\,s3.\,amazonaws.\,com/mnist/train-images-idx3-ubyte.\,gz}
                 Downloading \ \frac{https://ossci-datasets.\,s3.\,amazonaws.\,com/mnist/train-images-idx3-ubyte.\,gz}{https://ossci-datasets.\,s3.\,amazonaws.\,com/mnist/train-images-idx3-ubyte.\,gz} \ to \ ./data/MNIST/raw/train-images-idx3-ubyte.\,gz
                  100% 9.91M/9.91M [00:11<00:00, 899kB/s]
                  Extracting ./data/MNIST/raw/train-images-idx3-ubyte.gz to ./data/MNIST/raw
                  Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz
                  Failed to download (trying next):
                  <url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><url><ur
                 \label{lower_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_pow
                  100% 28.9k/28.9k [00:00<00:00, 135kB/s]
                  Extracting ./data/MNIST/raw/train-labels-idx1-ubyte.gz to ./data/MNIST/raw
                  Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz
                  Failed to download (trying next):
                  <urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certificate has expired (_ssl.c:1007)>
                \label{lem:bownloading https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz} \\ Downloading \frac{https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz}{to ./data/MNIST/raw/t10k-images-idx3-ubyte.gz} \\ to ./data/MNIST/raw/t10k-images-idx3-ubyte.gz
                  100% 1.65M/1.65M [00:01<00:00, 1.27MB/s]
                  Extracting ./data/MNIST/raw/t10k-images-idx3-ubyte.gz to ./data/MNIST/raw
                  Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz
                  Failed to download (trying next):
                 <urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certificate has expired (_ssl.c:1007)>
                Downloading <a href="https://ossci-datasets.s3.amazonaws.com/mnist/t10k-labels-idxl-ubyte.gz">https://ossci-datasets.s3.amazonaws.com/mnist/t10k-labels-idxl-ubyte.gz</a> to ./data/MNIST/raw/t10k-labels-idxl-ubyte.gz
                 100\% \\ \boxed{ \blacksquare 100\% \\ \blacksquare 200\% \\ \blacksquare 100\% \\ \blacksquare 200\% \\ 200\% \\ \blacksquare 200\% \\ \blacksquare 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 200\% \\ 20
```

!python main.py --dataset MNIST --method DM --ipc 10

====== Exp 0 =======

→ eval_it_pool: [0, 500, 1000]

```
Hyper-parameters:
 ('method': 'DM', 'dataset': 'MNIST', 'model': 'ConvNet', 'ipc': 10, 'eval mode': 'S', 'num expl': 5, 'num expl': 5, 'num expl': 20, 'epoch eval train': 300, 'Iteration': 1000, 'lr img': 0.1, 'lr net': 0.01, 'batch real': 256, 'batch train': 256, 'init': 'noise', 'dsa strategy': 'None', 'data path': 'data, 'save path': 're
Evaluation model nool: ['ConvNet']
class c = 0: 5923 real images
class c = 1: 6742 real images
class c = 2: 5958 real images
class c = 3: 6131 real images
class c = 4: 5842 real images
class c = 5: 5421 real images
class c = 6: 5918 real images
class c = 7: 6265 real images
class c = 8: 5851 real images
class c = 9: 5949 real images
real images channel 0, mean = -0.0001, std = 1.0000
/content/DatasetCondensation/main.py;89: UserWarning: Creating a tensor from a list of numpy.ndarrays is extremely slow. Please consider converting the list to a single numpy.ndarray with numpy.array() before converting to a tensor. (Triggered internally at ../torch/csrc/utils/tensor_new.cpp:278.)
  label svn = torch.tensor([np.ones(args.ipc)*i for i in range(num classes)], dtvpe=torch.long, requires grad=False, device=args, device=args, device), view(-1) # [0,0,0, 1,1,1, ..., 9,9,9]
initialize synthetic data from random noise
[2024-11-03 23:23:29] training begins
Evaluation
model train = ConvNet. model eval = ConvNet. iteration = 0
DC augmentation parameters:
 ('crop': 4, 'scale': 0.2, 'rotate': 45, 'noise': 0.001, 'strategy': 'crop scale rotate')
[2024-11-03 23:24:08] Evaluate 00: epoch = 1000 train time = 36 s train loss = 0.011867 train acc = 1.0000, test acc = 0.0943
[2024-11-03\ 23:24:46] Evaluate_01: epoch = 1000 train time = 35 s train loss = 0.015994 train acc = 1.0000, test acc = 0.1153
[2024-11-03 23:25:23] Evaluate 02: epoch = 1000 train time = 35 s train loss = 0.006754 train acc = 1.0000, test acc = 0.1453
[2024-11-03 23:26:01] Evaluate 03: epoch = 1000 train time = 35 s train loss = 0.009939 train acc = 1.0000, test acc = 0.0762
[2024-11-03 23:26:38] Evaluate 04: epoch = 1000 train time = 34 s train loss = 0.008773 train acc = 1.0000, test acc = 0.0859
[2024-11-03 23:27:15] Evaluate_05: epoch = 1000 train time = 34 s train loss = 0.009646 train acc = 1.0000, test acc = 0.0296
[2024-11-03 23:27:52] Evaluate 06: epoch = 1000 train time = 35 s train loss = 0.010515 train acc = 1.0000, test acc = 0.0811
[2024-11-03 23:28:29] Evaluate 07: epoch = 1000 train time = 34 s train loss = 0.011801 train acc = 1.0000, test acc = 0.1026
[2024-11-03 23:29:07] Evaluate_08: epoch = 1000 train time = 35 s train loss = 0.011601 train acc = 1.0000, test acc = 0.0547
[2024-11-03\ 23:29:44] Evaluate_09: epoch = 1000 train time = 35 s train loss = 0.007835 train acc = 1.0000, test acc = 0.0850
[2024-11-03 23:30:22] Evaluate_10: epoch = 1000 train time = 35 s train loss = 0.011320 train acc = 1.0000, test acc = 0.1140
[2024-11-03 23:30:59] Evaluate 11: epoch = 1000 train time = 35 s train loss = 0.008788 train acc = 1.0000, test acc = 0.0847
[2024-11-03 23:31:37] Evaluate 12: epoch = 1000 train time = 35 s train loss = 0.007590 train acc = 1.0000, test acc = 0.0716
[2024-11-03 23:32:14] Evaluate_13: epoch = 1000 train time = 35 s train loss = 0.010688 train acc = 1.0000, test acc = 0.0988
[2024-11-03 23:32:51] Evaluate 14: epoch = 1000 train time = 34 s train loss = 0.012900 train acc = 1.0000, test acc = 0.1350
[2024-11-03 23:33:29] Evaluate 15: epoch = 1000 train time = 35 s train loss = 0.013129 train acc = 1.0000, test acc = 0.0661
[2024-11-03 23:34:06] Evaluate_16: epoch = 1000 train time = 35 s train loss = 0.016028 train acc = 1.0000, test acc = 0.0656
[2024-11-03 23:34:44] Evaluate 17: epoch = 1000 train time = 35 s train loss = 0.012058 train acc = 1.0000, test acc = 0.0536
[2024-11-03 23:35:21] Evaluate 18: epoch = 1000 train time = 34 s train loss = 0.009164 train acc = 1.0000, test acc = 0.0657
[2024-11-03 23:35:58] Evaluate_19: epoch = 1000 train time = 35 s train loss = 0.011086 train acc = 1.0000, test acc = 0.1449
Evaluate 20 random ConvNet, mean = 0.0885 std = 0.0301
[2024-11-03 23:36:06] iter = 0000, loss = 218.2366
[2024-11-03 23:37:15] iter = 0010, loss = 91.7760
[2024-11-03 23:38:24] iter = 0020, loss = 72.9020
[2024-11-03 23:39:33] iter = 0030, loss = 62,8031
[2024-11-03 23:40:42] iter = 0040, loss = 53,0287
```

[2024-11-03 23:41:50] iter = 0050, loss = 49.4174 [2024-11-03 23:42:59] iter = 0060, loss = 45.8709

!ls /content/

→ DatasetCondensation drive sample_data