

**COMP4680/COMP8650: Advanced Topics
in SML**

**Assignment #6: Deep Learning
Programming Assignment**

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1.

- a) The size of the output: (32, 16, 64, 64) 32 batch sizes 16 channels image 64*64
- b) The size of the output: (32, 128, 3, 3) 32 batch sizes 128 channels image 3*3
- c) The size of the output: (32, 32, 65, 65) 32 batch sizes 32 channels image 65*65
- d) The number of learnable parameters: 448 (16*3*3+16)

2.

- a) The source code show as below:

```
def __init__(self, z_dim=64, img_size=64):  
    super(Encoder, self).__init__()  
  
    self.z_dim = z_dim  
    self.hidden_layer = nn.Sequential(  
        nn.Linear(3*img_size*img_size, z_dim),  
        nn.BatchNorm1d(z_dim),  
        nn.ReLU(),  
    )  
    self.output_layer = nn.Sequential(  
        nn.Linear(z_dim, z_dim),  
        nn.Tanh(),  
    )
```

- b) The source code show as below:

```
def __init__(self, z_dim=64, img_size=64):
    super(Decoder, self).__init__()

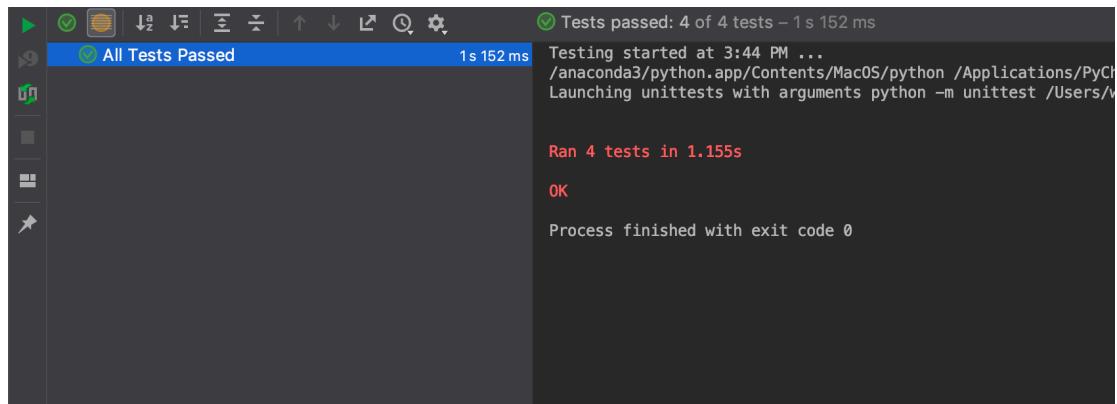
    assert img_size==64
    self.z_dim = z_dim
    self.layer1 = nn.Sequential(
        nn.ConvTranspose2d(
            in_channels=z_dim,
            out_channels=128,
            kernel_size=7,
            stride=1,
        ),
        nn.BatchNorm2d(128),
        nn.ReLU(),
        nn.Conv2d(
            in_channels=128,
            out_channels=128,
            kernel_size=3,
            stride=1,
            padding=1,
        ),
        nn.BatchNorm2d(128),
        nn.ReLU(),
    )
    self.layer2 = nn.Sequential(
        nn.ConvTranspose2d(
            in_channels=128,
            out_channels=64,
            kernel_size=3,
            stride=2,
        ),
        nn.BatchNorm2d(64),
        nn.ReLU(),
        nn.Conv2d(
            in_channels=64,
            out_channels=64,
            kernel_size=3,
            stride=1,
            padding=1,
        ),
        nn.BatchNorm2d(64),
        nn.ReLU(),
    )

```

```
    self.layer3 = nn.Sequential(
        nn.ConvTranspose2d(
            in_channels=64,
            out_channels=64,
            kernel_size=3,
            stride=2,
        ),
        nn.BatchNorm2d(64),
        nn.ReLU(),
        nn.Conv2d(
            in_channels=64,
            out_channels=64,
            kernel_size=3,
            stride=1,
            padding=1,
        ),
        nn.BatchNorm2d(64),
        nn.ReLU(),
    )
    self.layer4 = nn.Sequential(
        nn.ConvTranspose2d(
            in_channels=64,
            out_channels=32,
            kernel_size=3,
            stride=2,
        ),
        nn.BatchNorm2d(32),
        nn.ReLU(),
        nn.Conv2d(
            in_channels=32,
            out_channels=32,
            kernel_size=3,
            stride=1,
            padding=1,
        ),
        nn.BatchNorm2d(32),
        nn.ReLU(),
    )
    self.layer5 = nn.Sequential(
        nn.ConvTranspose2d(
            in_channels=32,
            out_channels=32,
            kernel_size=2,
            stride=1,
        ),
        nn.BatchNorm2d(32),
        nn.ReLU(),
        nn.Conv2d(
            in_channels=32,
            out_channels=3,
            kernel_size=1,
            stride=1,
            padding=0,
        ),
        nn.Tanh(),
    )
```

3.

- a) All the models layers are correctly sized. The check results show as below:



A screenshot of a terminal window titled "All Tests Passed". The status bar at the top right shows "Tests passed: 4 of 4 tests - 1 s 152 ms". The main text area displays the following output:
Testing started at 3:44 PM ...
/anaconda3/python.app/Contents/MacOS/python /Applications/PyCh
Launching unittests with arguments python -m unittest /Users/

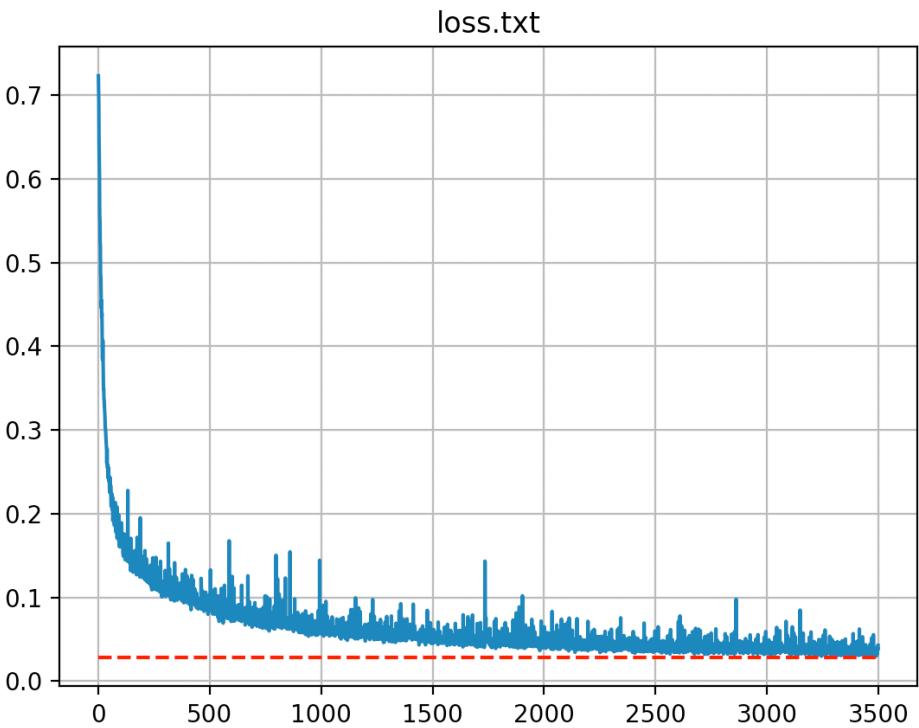
Ran 4 tests in 1.155s
OK
Process finished with exit code 0

- b) The results are stored in the loss.txt after training. The following images shows a part of results(loss.txt).

1	0.7229053974151611	3484	0.03444317355751991
2	0.7074987292289734	3485	0.03161868825554848
3	0.690818727016449	3486	0.034617651253938675
4	0.6431910991668701	3487	0.03284025937318802
5	0.6239379048347473	3488	0.02973368763923645
6	0.6041852235794067	3489	0.03808487579226494
7	0.5571762919425964	3490	0.03285076841711998
8	0.5508760809898376	3491	0.03781181946396828
9	0.5224028825759888	3492	0.030032457783818245
10	0.5199987292289734	3493	0.040223587304353714
11	0.4869062602519989	3494	0.03645233437418938
12	0.48292914032936096	3495	0.031237980350852013
13	0.4623061716556549	3496	0.03583160415291786
14	0.44585180282592773	3497	0.03383314236998558
15	0.45462584495544434	3498	0.04063452407717705
16	0.4349321126937866	3499	0.03721265494823456
17	0.4391724765300751	3500	0.042742013931274414

- c) The loss function show as below. According to the following image, the loss function seems like converged.

Figure 1

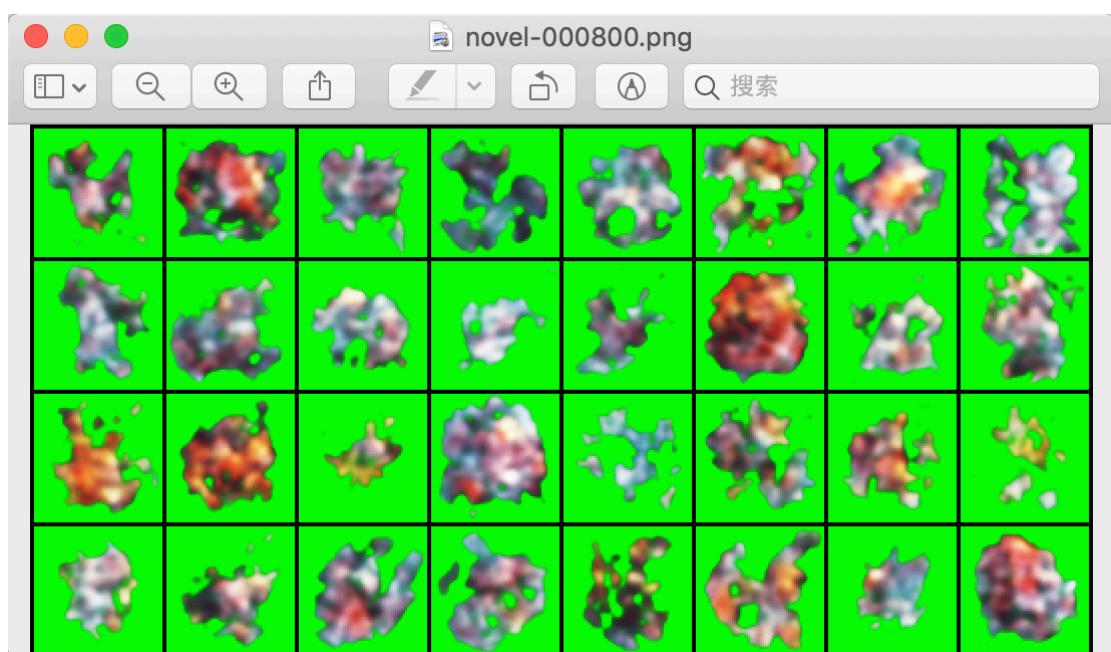
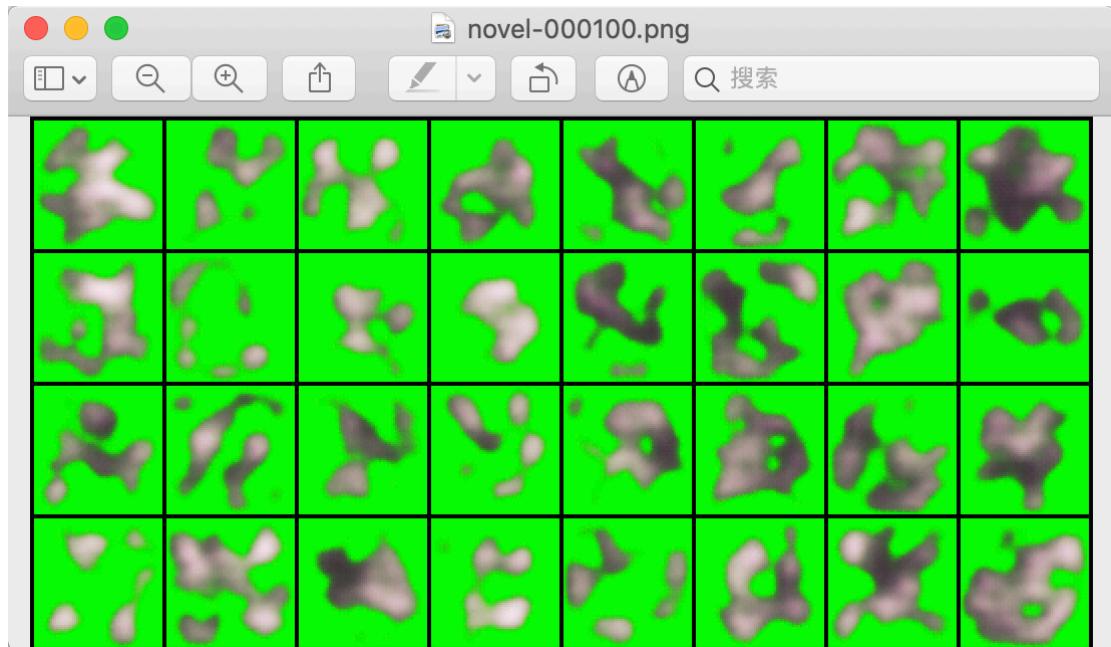


- d) According to the images of sample-*.png, the quality of the denoised images become higher as the training processs. Therefore, this can deal with this situation. The following images are the results after 100 iterations, 800 iterations, 2300 iterations, 3400 iterations respectively.



- e) According to the images of novel-*.png, the quality of the images generated from

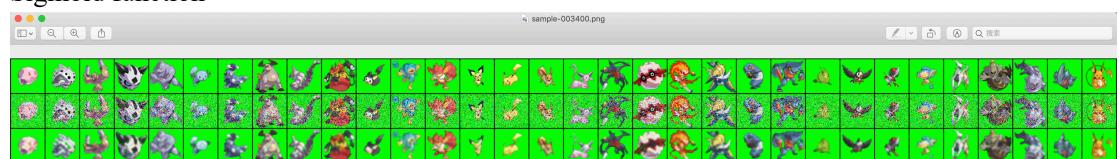
random noise injected into the decoder model become higher. But the final results are still hard to recognize. Therefore, this decoder model can not deal with this situation. The following images are the results after 100 iterations, 800 iterations, 2300 iterations, 3400 iterations respectively.



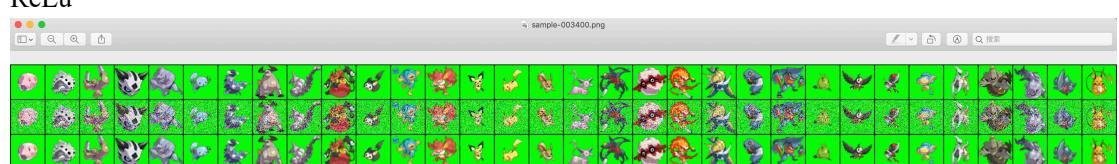


- f) 1) I changed the activation function from ReLU to Sigmoid, the results become bad. The following images show the result.

Sigmoid function



ReLU



2)The size of results are changed through the change of batch sizes(32 -> 128).

