

A3) Unsupervised Learning with Autoencoders

```
In [39]: 1 import matplotlib.pyplot as plt
2 import numpy as np
3
4 import requests, pickle, gzip, math
5 from pathlib import Path
6
7 import torch
8 import torch.nn as nn
9 import torch.optim as optim
10 import torch.nn.functional as F
11
12 DATA_PATH = Path("data")
13 PATH = DATA_PATH / "mnist"
14 PATH.mkdir(parents=True, exist_ok=True)
15 URL = "http://deeplearning.net/data/mnist/"
16 FILENAME = "mnist.pkl.gz"
17
18 if not (PATH / FILENAME).exists():
19     content = requests.get(URL + FILENAME).content
20     (PATH / FILENAME).open("wb").write(content)
21
22 with gzip.open((PATH / FILENAME).as_posix(), "rb") as f:
23     ((x_train, y_train), (x_valid, y_valid), _) = pickle.load(f, encoding="latin-1")
24
25 x_train, y_train, x_valid, y_valid = map(torch.tensor, (x_train, y_train, x_valid, y_valid))
26
27 n, c = x_train.shape
```

```
In [40]: 1 import torch
2 import torchvision
3 from torchvision import transforms, datasets
4 import torch.nn as nn
5 import torch.nn.functional as F
6 import torch.optim as optim
7 import matplotlib.pyplot as plt
8 import numpy as np
9
10
11
12 train = datasets.MNIST('', train=True, download=True,
13                       transform=transforms.Compose([
14                           transforms.ToTensor()
15                       ]))
16
17 test = datasets.MNIST('', train=False, download=True,
18                      transform=transforms.Compose([
19                          transforms.ToTensor()
20                      ]))
21
22 trainset = torch.utils.data.DataLoader(train, batch_size=6000, shuffle=False)
23 testset = torch.utils.data.DataLoader(test, batch_size=6000, shuffle=False)
```

```
In [43]: 1 global h
2 h=32
3 epoch_n=20
```

```
In [44]: 1 class linear_Autoencoder(nn.Module):
2         def __init__(self):
3             super().__init__()
4             self.encoder = nn.Linear(28*28, h)
5             self.decoder = nn.Linear(h, 28*28)
6
7         def forward(self, x):
8             x = self.encoder(x)
9             x = self.decoder(x)
10            return x
11
12 net = linear_Autoencoder()
13 loss_function = nn.MSELoss()
14 optimizer = optim.Adam(net.parameters(), lr=0.001)
```

linear_Autoencoder

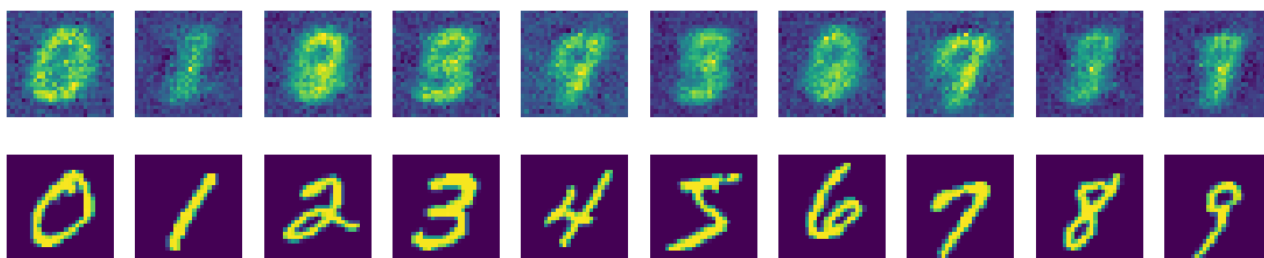
h=32

```
In [45]: 1 h=32
2
3 class linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x = self.decoder(x)
12        return x
13
14 net = linear_Autoencoder()
15 loss_function = nn.MSELoss()
16 optimizer = optim.Adam(net.parameters(), lr=0.001)
17
18 for epoch in range(epoch_n):
19     for data in trainset:
20         x, y = data
21         net.zero_grad()
22         output = net(X.view(-1,784))
23         loss = loss_function(output, X.view(-1,784))
24         loss.backward()
25         optimizer.step()
26     print(loss)
```

```
tensor(0.1122, grad_fn=<MseLossBackward>)
tensor(0.0914, grad_fn=<MseLossBackward>)
tensor(0.0740, grad_fn=<MseLossBackward>)
tensor(0.0658, grad_fn=<MseLossBackward>)
tensor(0.0624, grad_fn=<MseLossBackward>)
tensor(0.0605, grad_fn=<MseLossBackward>)
tensor(0.0589, grad_fn=<MseLossBackward>)
tensor(0.0571, grad_fn=<MseLossBackward>)
tensor(0.0550, grad_fn=<MseLossBackward>)
tensor(0.0528, grad_fn=<MseLossBackward>)
```

```
In [46]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 32



h=64

In [47]:

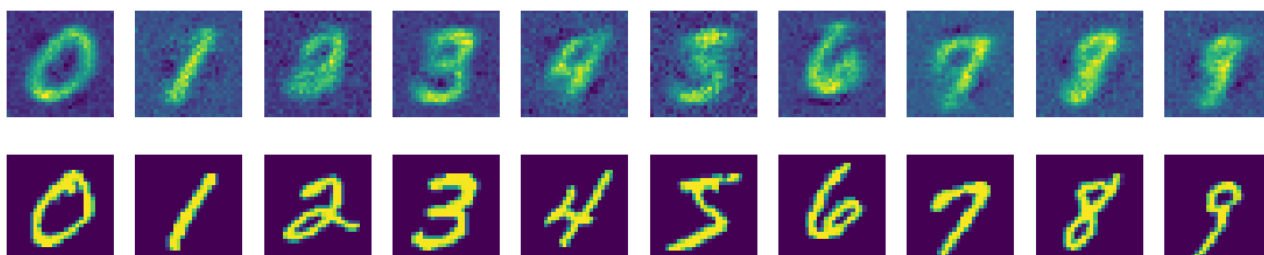
```
1 h=64
2
3 class linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x = self.decoder(x)
12        return x
13
14 net = linear_Autoencoder()
15 loss_function = nn.MSELoss()
16 optimizer = optim.Adam(net.parameters(), lr=0.001)
17
18 for epoch in range(epoch_n):
19     for data in trainset:
20         X, y = data
21         net.zero_grad()
22         output = net(X.view(-1,784))
23         loss = loss_function(output, X.view(-1,784))
24         loss.backward()
25         optimizer.step()
26     print(loss)
```

```
tensor(0.0900, grad_fn=<MseLossBackward>)
tensor(0.0656, grad_fn=<MseLossBackward>)
tensor(0.0592, grad_fn=<MseLossBackward>)
tensor(0.0547, grad_fn=<MseLossBackward>)
tensor(0.0505, grad_fn=<MseLossBackward>)
tensor(0.0466, grad_fn=<MseLossBackward>)
tensor(0.0430, grad_fn=<MseLossBackward>)
tensor(0.0400, grad_fn=<MseLossBackward>)
tensor(0.0374, grad_fn=<MseLossBackward>)
tensor(0.0351, grad_fn=<MseLossBackward>)
```

In [48]:

```
1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=x_train[index[i]].detach().numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 64



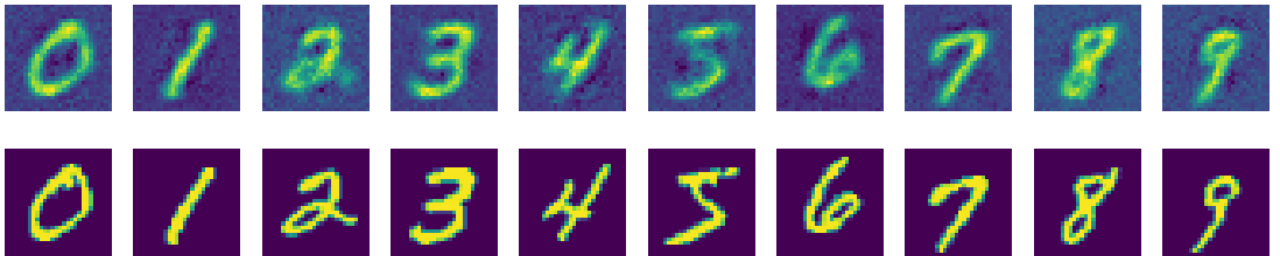
h=128

```
In [78]: 1 h=128
2
3 class linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x = self.decoder(x)
12        return x
13
14 net = linear_Autoencoder()
15 loss_function = nn.MSELoss()
16 optimizer = optim.Adam(net.parameters(), lr=0.001)
17
18 for epoch in range(epoch_n):
19     for data in trainset:
20         X, y = data
21         net.zero_grad()
22         output = net(X.view(-1,784))
23         loss = loss_function(output, X.view(-1,784))
24         loss.backward()
25         optimizer.step()
26     print(loss)
```

```
tensor(0.0681, grad_fn=<MseLossBackward>)
tensor(0.0556, grad_fn=<MseLossBackward>)
tensor(0.0460, grad_fn=<MseLossBackward>)
tensor(0.0388, grad_fn=<MseLossBackward>)
tensor(0.0339, grad_fn=<MseLossBackward>)
tensor(0.0303, grad_fn=<MseLossBackward>)
tensor(0.0274, grad_fn=<MseLossBackward>)
tensor(0.0249, grad_fn=<MseLossBackward>)
tensor(0.0227, grad_fn=<MseLossBackward>)
tensor(0.0208, grad_fn=<MseLossBackward>)
```

```
In [79]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 128



c) $F_1(x)$ (use $h = 128$ here) on the test set

```
In [80]: 1 # I run this code after training with F_1 with h=128
2
3 LOSS=0
4 for data in testset:
5     X, y = data
6     net.zero_grad()
7     output = net(X.view(-1,784))
8     LOSS+= loss_function(output, X.view(-1,784))
9     #loss.backward()
10    #optimizer.step()
11    print("F_1(x) with h = 128 loss on the test set is", LOSS.detach().numpy())
```

F_1(x) with h = 128 loss on the test set is 0.04061942

b) Non_linear Autoencoder

```
In [51]: 1 class non_linear_Autoencoder(nn.Module):
2         def __init__(self):
3             super().__init__()
4             self.encoder = nn.Linear(28*28, h)
5             self.decoder = nn.Linear(h, 28*28)
6
7         def forward(self, x):
8             x = self.encoder(x)
9             x=F.relu(x)
10            x = self.decoder(x)
11            x=F.relu(x)
12            return x
13 net = non_linear_Autoencoder()
14 loss_function = nn.MSELoss()
15 optimizer = optim.Adam(net.parameters(), lr=0.001)
```

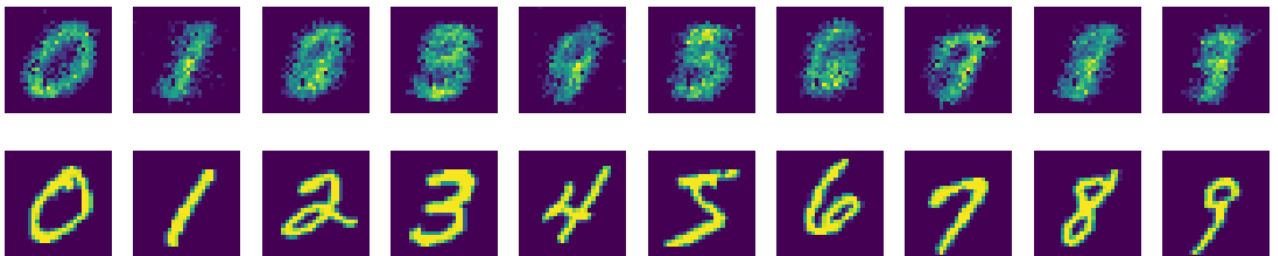
h=32

```
In [52]: 1 h=32
2
3 class non_linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x=F.relu(x)
12        x = self.decoder(x)
13        x=F.relu(x)
14        return x
15
16 net = non_linear_Autoencoder()
17 loss_function = nn.MSELoss()
18 optimizer = optim.Adam(net.parameters(), lr=0.001)
19
20 for epoch in range(epoch_n):
21     for data in trainset:
22         X, y = data
23         net.zero_grad()
24         output = net(X.view(-1,784))
25         loss = loss_function(output, X.view(-1,784))
26         loss.backward()
27         optimizer.step()
28     print(loss)
```

```
tensor(0.0949, grad_fn=<MseLossBackward>)
tensor(0.0842, grad_fn=<MseLossBackward>)
tensor(0.0745, grad_fn=<MseLossBackward>)
tensor(0.0676, grad_fn=<MseLossBackward>)
tensor(0.0631, grad_fn=<MseLossBackward>)
tensor(0.0595, grad_fn=<MseLossBackward>)
tensor(0.0561, grad_fn=<MseLossBackward>)
tensor(0.0529, grad_fn=<MseLossBackward>)
tensor(0.0500, grad_fn=<MseLossBackward>)
tensor(0.0470, grad_fn=<MseLossBackward>)
```

```
In [53]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 32



h=64

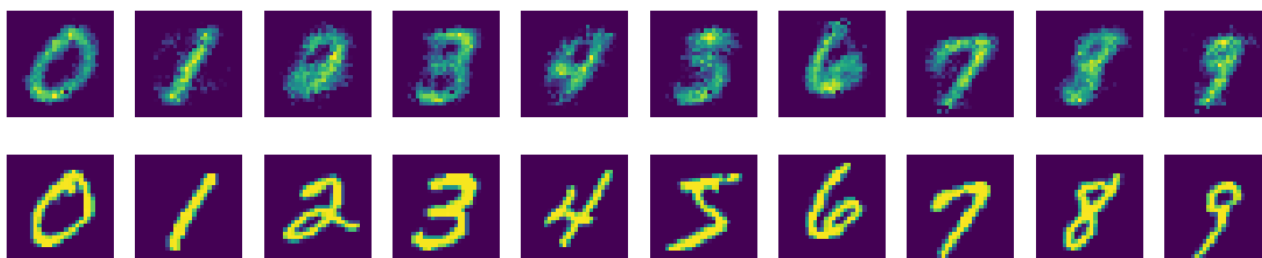
In [54]:

```
1 h=64
2
3 class non_linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x=F.relu(x)
12        x = self.decoder(x)
13        x=F.relu(x)
14        return x
15
16 net = non_linear_Autoencoder()
17 loss_function = nn.MSELoss()
18 optimizer = optim.Adam(net.parameters(), lr=0.001)
19
20 for epoch in range(epoch_n):
21     for data in trainset:
22         x, y = data
23         net.zero_grad()
24         output = net(X.view(-1,784))
25         loss = loss_function(output, X.view(-1,784))
26         loss.backward()
27         optimizer.step()
28     print(loss)
```

```
tensor(0.0852, grad_fn=<MseLossBackward>)
tensor(0.0688, grad_fn=<MseLossBackward>)
tensor(0.0598, grad_fn=<MseLossBackward>)
tensor(0.0539, grad_fn=<MseLossBackward>)
tensor(0.0482, grad_fn=<MseLossBackward>)
tensor(0.0431, grad_fn=<MseLossBackward>)
tensor(0.0387, grad_fn=<MseLossBackward>)
tensor(0.0350, grad_fn=<MseLossBackward>)
tensor(0.0317, grad_fn=<MseLossBackward>)
tensor(0.0291, grad_fn=<MseLossBackward>)
```

```
In [55]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 64



h=128

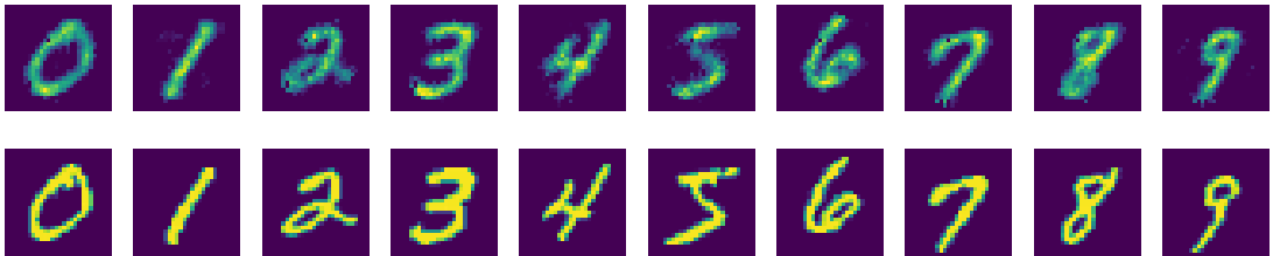
In [81]:

```
1 h=128
2
3 class non_linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x=F.relu(x)
12        x = self.decoder(x)
13        x=F.relu(x)
14        return x
15
16 net = non_linear_Autoencoder()
17 loss_function = nn.MSELoss()
18 optimizer = optim.Adam(net.parameters(), lr=0.001)
19
20 for epoch in range(epoch_n):
21     for data in trainset:
22         x, y = data
23         net.zero_grad()
24         output = net(X.view(-1,784))
25         loss = loss_function(output, X.view(-1,784))
26         loss.backward()
27         optimizer.step()
28     print(loss)
```

```
tensor(0.0707, grad_fn=<MseLossBackward>)
tensor(0.0561, grad_fn=<MseLossBackward>)
tensor(0.0456, grad_fn=<MseLossBackward>)
tensor(0.0375, grad_fn=<MseLossBackward>)
tensor(0.0313, grad_fn=<MseLossBackward>)
tensor(0.0268, grad_fn=<MseLossBackward>)
tensor(0.0235, grad_fn=<MseLossBackward>)
tensor(0.0207, grad_fn=<MseLossBackward>)
tensor(0.0186, grad_fn=<MseLossBackward>)
tensor(0.0169, grad_fn=<MseLossBackward>)
```

```
In [82]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 128



c) $F_2(x)$ (use $h = 128$ here) on the test set

```
In [83]: 1 # I run this code after training with F_2 with h=128
2
3 LOSS=0
4 for data in testset:
5     X, y = data
6     net.zero_grad()
7     output = net(X.view(-1,784))
8     LOSS+= loss_function(output, X.view(-1,784))
9     #loss.backward()
10    #optimizer.step()
11    print("F_2(x) with h = 128 loss on the test set is", LOSS.detach().numpy())
```

F_2(x) with h = 128 loss on the test set is 0.033017926

```
In [ ]: 1
```

out of curiosity part

h=10

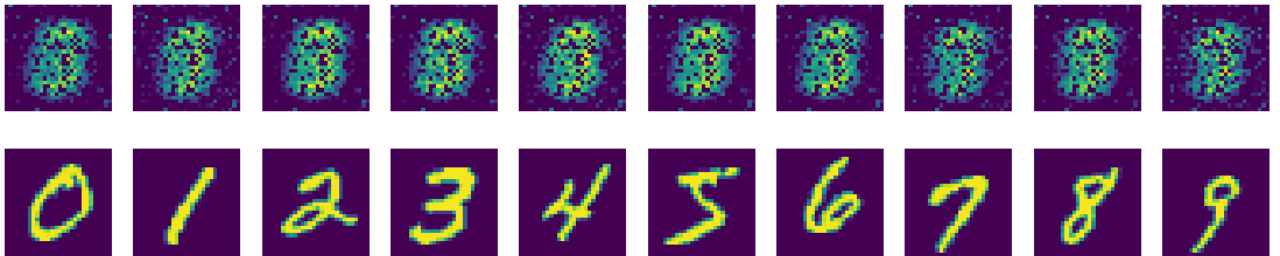
In [77]:

```
1 h=10
2
3 class non_linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x=F.relu(x)
12        x = self.decoder(x)
13        x=F.relu(x)
14        return x
15
16 net = non_linear_Autoencoder()
17 loss_function = nn.MSELoss()
18 optimizer = optim.Adam(net.parameters(), lr=0.001)
19
20 for epoch in range(epoch_n):
21     for data in trainset:
22         x, y = data
23         net.zero_grad()
24         output = net(X.view(-1,784))
25         loss = loss_function(output, X.view(-1,784))
26         loss.backward()
27         optimizer.step()
28     print(loss)
```

```
tensor(0.1042, grad_fn=<MseLossBackward>)
tensor(0.1011, grad_fn=<MseLossBackward>)
tensor(0.0974, grad_fn=<MseLossBackward>)
tensor(0.0933, grad_fn=<MseLossBackward>)
tensor(0.0889, grad_fn=<MseLossBackward>)
tensor(0.0848, grad_fn=<MseLossBackward>)
tensor(0.0812, grad_fn=<MseLossBackward>)
tensor(0.0784, grad_fn=<MseLossBackward>)
tensor(0.0763, grad_fn=<MseLossBackward>)
tensor(0.0748, grad_fn=<MseLossBackward>)
```

```
In [59]: 1 i=-1
2 N=10
3 index=[1,3,5,7,9,0,13,15,17,19]
4 plt.figure(figsize=(14,3),dpi=100)
5 print('h=',h)
6 for j in range(N):
7     i+=1
8     j+=1
9     plt.subplot(2,N,j)
10    ww=net(x_train[index[i]].detach()).numpy()
11    x_shape=np.reshape(ww,(28,28))
12    plt.axis('off')
13    plt.imshow(x_shape)
14    plt.subplot(2,N,N+j)
15    ww=x_train[index[i]].detach().numpy()
16    x_shape=np.reshape(ww,(28,28))
17    plt.axis('off')
18    plt.imshow(x_shape)
19
20 plt.show()
```

h= 10



h=28*28

In [60]:

```
1 h=28*28
2
3 class non_linear_Autoencoder(nn.Module):
4     def __init__(self):
5         super().__init__()
6         self.encoder = nn.Linear(28*28, h)
7         self.decoder = nn.Linear(h, 28*28)
8
9     def forward(self, x):
10        x = self.encoder(x)
11        x=F.relu(x)
12        x = self.decoder(x)
13        x=F.relu(x)
14        return x
15
16 net = non_linear_Autoencoder()
17 loss_function = nn.MSELoss()
18 optimizer = optim.Adam(net.parameters(), lr=0.001)
19
20 for epoch in range(epoch_n):
21     for data in trainset:
22         x, y = data
23         net.zero_grad()
24         output = net(X.view(-1,784))
25         loss = loss_function(output, X.view(-1,784))
26         loss.backward()
27         optimizer.step()
28     print(loss)
```

```
tensor(0.0514, grad_fn=<MseLossBackward>)
tensor(0.0317, grad_fn=<MseLossBackward>)
tensor(0.0223, grad_fn=<MseLossBackward>)
tensor(0.0163, grad_fn=<MseLossBackward>)
tensor(0.0126, grad_fn=<MseLossBackward>)
tensor(0.0104, grad_fn=<MseLossBackward>)
tensor(0.0089, grad_fn=<MseLossBackward>)
tensor(0.0079, grad_fn=<MseLossBackward>)
tensor(0.0072, grad_fn=<MseLossBackward>)
tensor(0.0066, grad_fn=<MseLossBackward>)
```

```

In [61]: 1 i=-1
          2 N=10
          3 index=[1,3,5,7,9,0,13,15,17,19]
          4 plt.figure(figsize=(14,3),dpi=100)
          5 print('h=',h)
          6 for j in range(N):
          7     i+=1
          8     j+=1
          9     plt.subplot(2,N,j)
         10     ww=net(x_train[index[i]]).detach().numpy()
         11     x_shape=np.reshape(ww,(28,28))
         12     plt.axis('off')
         13     plt.imshow(x_shape)
         14     plt.subplot(2,N,N+j)
         15     ww=x_train[index[i]].detach().numpy()
         16     x_shape=np.reshape(ww,(28,28))
         17     plt.axis('off')
         18     plt.imshow(x_shape)
         19
         20 plt.show()

```

h= 784

