Assignment is below at the bottom

Video 13.1 https://www.youtube.com/watch?v=kIGHE7Cfe1s

Video 13.2 https://www.youtube.com/watch?v=Rm9bJcDd1KU

Video 13.3 https://youtu.be/6HjZk-3LsjE

```
In [77]:
         from keras.callbacks import TensorBoard
         from keras.layers import Input, Dense
         from keras.models import Model
          from keras.datasets import mnist
          import numpy as np
          (xtrain, ytrain), (xtest, ytest) = mnist.load_data()
         xtrain = xtrain.astvpe('float32') / 255.
         xtest = xtest.astype('float32') / 255.
         xtrain = xtrain.reshape((len(xtrain), np.prod(xtrain.shape[1:])))
          xtest = xtest.reshape((len(xtest), np.prod(xtest.shape[1:])))
         xtrain.shape, xtest.shape
         ((60000, 784), (10000, 784))
Out[77]:
In [78]: # this is the size of our encoded representations
         encoding dim = 4 # 32 floats -> compression of factor 24.5, assuming the input is 784
         # this is our input placeholder
         x = input_img = Input(shape=(784,))
         # "encoded" is the encoded representation of the input
          x = Dense(256, activation='relu')(x)
          x = Dense(128, activation='relu')(x)
          encoded = Dense(encoding dim, activation='relu')(x)
         # "decoded" is the lossy reconstruction of the input
         x = Dense(128, activation='relu')(encoded)
         x = Dense(256, activation='relu')(x)
         decoded = Dense(784, activation='sigmoid')(x)
         # this model maps an input to its reconstruction
          autoencoder = Model(input_img, decoded)
         encoder = Model(input img, encoded)
         # create a placeholder for an encoded (32-dimensional) input
          encoded input = Input(shape=(encoding dim,))
          # retrieve the last layer of the autoencoder model
          dcd1 = autoencoder.layers[-1]
          dcd2 = autoencoder.layers[-2]
          dcd3 = autoencoder.layers[-3]
         # create the decoder model
          decoder = Model(encoded input, dcd1(dcd2(dcd3(encoded input))))
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Epoch 1/100
888
Epoch 2/100
Epoch 3/100
645
Epoch 4/100
1609
Epoch 5/100
1583
Epoch 6/100
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Epoch 7/100
1553
Epoch 8/100
1537
Epoch 9/100
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Epoch 10/100
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Epoch 11/100
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Epoch 12/100
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Epoch 13/100
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Epoch 20/100
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Epoch 21/100
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Epoch 22/100
Epoch 23/100
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Epoch 24/100
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Epoch 27/100
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Epoch 28/100
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Epoch 29/100
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Epoch 30/100
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Epoch 31/100
1435
Epoch 32/100
1434
Epoch 33/100
235/235 [============= ] - 3s 11ms/step - loss: 0.1401 - val loss: 0.
Epoch 34/100
1433
Epoch 35/100
1431
Epoch 36/100
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Epoch 37/100
Epoch 38/100
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Epoch 39/100
1425
Epoch 40/100
1425
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Epoch 41/100
1422
Epoch 42/100
Epoch 43/100
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Epoch 44/100
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Epoch 45/100
Epoch 46/100
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Epoch 47/100
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Epoch 48/100
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Epoch 49/100
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Epoch 50/100
1420
Epoch 51/100
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Epoch 52/100
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Epoch 53/100
Epoch 54/100
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Epoch 55/100
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Epoch 56/100
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Epoch 57/100
Epoch 58/100
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Epoch 59/100
1417
Epoch 60/100
1413
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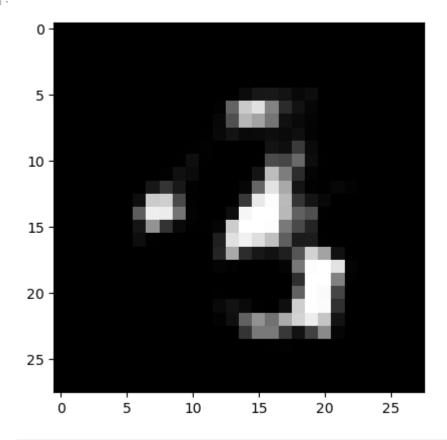
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Epoch 61/100
1414
Epoch 62/100
Epoch 63/100
1414
Epoch 64/100
1413
Epoch 65/100
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Epoch 66/100
1412
Epoch 67/100
1411
Epoch 68/100
1410
Epoch 69/100
Epoch 70/100
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Epoch 71/100
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Epoch 72/100
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Epoch 73/100
Epoch 74/100
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Epoch 75/100
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Epoch 76/100
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Epoch 77/100
Epoch 78/100
1412
Epoch 79/100
1410
Epoch 80/100
1408
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Epoch 81/100
1410
Epoch 82/100
Epoch 83/100
1409
Epoch 84/100
1408
Epoch 85/100
1407
Epoch 86/100
1406
Epoch 87/100
1410
Epoch 88/100
1408
Epoch 89/100
1408
Epoch 90/100
1411
Epoch 91/100
1406
Epoch 92/100
1409
Epoch 93/100
235/235 [============= ] - 3s 12ms/step - loss: 0.1334 - val loss: 0.
Epoch 94/100
1407
Epoch 95/100
1408
Epoch 96/100
1408
Epoch 97/100
Epoch 98/100
1408
Epoch 99/100
1406
Epoch 100/100
1410
```

<keras.src.callbacks.History at 0x2202830f040>

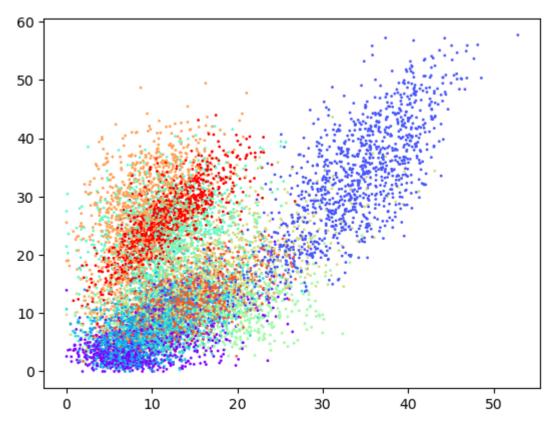
```
Out[80]:
In [11]:
         encoded_imgs
         array([[38.304043 ,
                                         9.823541 , 22.699774 ],
Out[11]:
                [ 4.0672283,
                             0.
                                         6.457829 , 2.599104 ],
                [94.89234 ,
                             0.
                                      , 48.237885 , 22.512798 ],
                . . . ,
                [20.050169 ,
                             0.
                                         8.473201 , 8.569888 ],
                [19.308092 ,
                                      , 15.271025 , 8.056674 ],
                                         2.7675493, 3.3705957]], dtype=float32)
                [ 3.769403 ,
         noise = np.random.normal(20,4, (4,4))
In [12]:
         noise preds = decoder.predict(noise)
         1/1 [======= ] - 0s 18ms/step
         plt.imshow(noise_preds[1].reshape(28,28))
In [81]:
```

<matplotlib.image.AxesImage at 0x22028529e40> Out[81]:



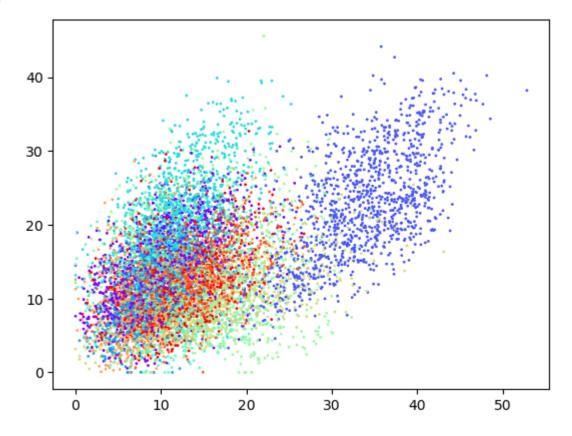
```
np.max(encoded imgs)
In [14]:
         98.423355
Out[14]:
         encoded imgs = encoder.predict(xtest)
In [82]:
          decoded imgs = decoder.predict(encoded imgs)
          import matplotlib.pyplot as plt
          n = 20 # how many digits we will display
          plt.figure(figsize=(40, 4))
          for i in range(n):
              # display original
```

```
ax = plt.subplot(2, n, i + 1)
            plt.imshow(xtest[i].reshape(28, 28))
            plt.gray()
            ax.get_xaxis().set_visible(False)
            ax.get yaxis().set visible(False)
            # display reconstruction
            ax = plt.subplot(2, n, i + 1 + n)
            plt.imshow(decoded_imgs[i].reshape(28, 28))
            plt.gray()
            ax.get_xaxis().set_visible(False)
            ax.get_yaxis().set_visible(False)
        plt.show()
        313/313 [========== ] - 1s 2ms/step
        313/313 [========== ] - 0s 1ms/step
                                   9590690159734
                1041998906901597
In [15]: encoded_imgs
                                     9.823541 , 22.699774 ],
        array([[38.304043 , 0.
Out[15]:
              [ 4.0672283, 0. , 6.457829 , 2.599104 ],
              [94.89234 , 0.
                                 , 48.237885 , 22.512798 ],
              [19.308092 , 0. , 8.473201 , 8.569888 ],
[3.760403
              ...,
              [ 3.769403 , 0.
                                     2.7675493, 3.3705957]], dtype=float32)
        %matplotlib inline
In [16]:
        plt.scatter(encoded_imgs[:,1], encoded_imgs[:,0], s=1, c=ytest, cmap='rainbow')
In [83]:
        # plt.show()
        <matplotlib.collections.PathCollection at 0x22028502ce0>
Out[83]:
```



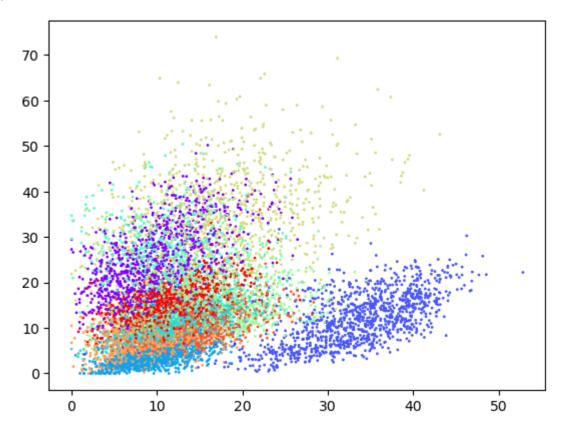
In [84]: plt.scatter(encoded_imgs[:,1], encoded_imgs[:,3], s=1, c=ytest, cmap='rainbow')
plt.show()

Out[84]: <matplotlib.collections.PathCollection at 0x220281d6710>



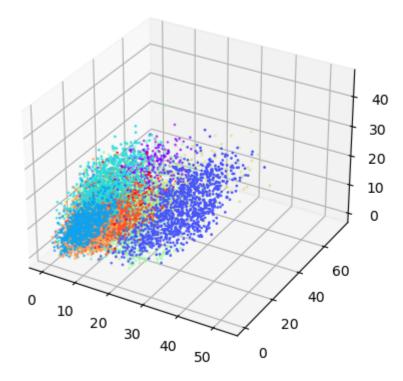
In [85]: plt.scatter(encoded_imgs[:,1], encoded_imgs[:,2], s=1, c=ytest, cmap='rainbow')
plt.show()

Out[85]: <matplotlib.collections.PathCollection at 0x22028109ff0>



```
In [86]: from mpl_toolkits.mplot3d import Axes3D
fig = plt.figure()
    ax = fig.add_subplot(111, projection='3d')
    ax.scatter(encoded_imgs[:,1], encoded_imgs[:,2], encoded_imgs[:,3], c=ytest, cmap='rai
```

Out[86]: <mpl_toolkits.mplot3d.art3d.Path3DCollection at 0x2202802b9a0>



```
In [ ]:
```

Assignment

1. change the encoding_dim through various values (range(2,18,2) and save the loss you can get. Plot the 8 pairs of dimensions vs loss on a scatter plot

```
(xtrain, ytrain), (xtest, ytest) = mnist.load_data()
In [21]:
         xtrain = xtrain.astype('float32') / 255.
          xtest = xtest.astype('float32') / 255.
         xtrain = xtrain.reshape((len(xtrain), np.prod(xtrain.shape[1:])))
          xtest = xtest.reshape((len(xtest), np.prod(xtest.shape[1:])))
         xtrain.shape, xtest.shape
         ((60000, 784), (10000, 784))
Out[21]:
In [23]:
         losses = []
         dimensions = [2, 4, 6, 8, 10, 12, 14, 16]
         for encoding dim in dimensions:
             x = input_img = Input(shape=(784,))
             x = Dense(256, activation='relu')(x)
             x = Dense(128, activation='relu')(x)
             encoded = Dense(encoding dim, activation='relu')(x)
             x = Dense(128, activation='relu')(encoded)
             x = Dense(256, activation='relu')(x)
             decoded = Dense(784, activation='sigmoid')(x)
             autoencoder = Model(input img, decoded)
             encoder = Model(input img, encoded)
             encoded input = Input(shape=(encoding dim,))
             dcd1 = autoencoder.layers[-1]
             dcd2 = autoencoder.layers[-2]
             dcd3 = autoencoder.layers[-3]
             decoder = Model(encoded input, dcd1(dcd2(dcd3(encoded input))))
             autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
             autoencoder.fit(xtrain, xtrain,
                          epochs=50,
                          batch_size=256,
                          shuffle=True,
                          validation data=(xtest, xtest),
                          callbacks=[TensorBoard(log dir='/tmp/autoencoder')])
             loss = autoencoder.evaluate(xtrain, xtrain, verbose = 0)
              losses.append(loss)
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Epoch 1/50
498
Epoch 2/50
Epoch 3/50
2040
Epoch 4/50
1981
Epoch 5/50
1931
Epoch 6/50
1899
Epoch 7/50
1873
Epoch 8/50
1860
Epoch 9/50
1843
Epoch 10/50
1832
Epoch 11/50
1824
Epoch 12/50
1819
Epoch 13/50
1811
Epoch 14/50
1801
Epoch 15/50
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Epoch 16/50
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Epoch 17/50
1786
Epoch 18/50
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Epoch 19/50
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Epoch 20/50
1774
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Epoch 21/50
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Epoch 22/50
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Epoch 23/50
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Epoch 24/50
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Epoch 25/50
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Epoch 26/50
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Epoch 27/50
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Epoch 28/50
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Epoch 29/50
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Epoch 30/50
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Epoch 31/50
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Epoch 32/50
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Epoch 33/50
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Epoch 34/50
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Epoch 35/50
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Epoch 36/50
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Epoch 37/50
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Epoch 38/50
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Epoch 39/50
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Epoch 40/50
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Epoch 41/50
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Epoch 42/50
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Epoch 43/50
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Epoch 44/50
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Epoch 45/50
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Epoch 46/50
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Epoch 47/50
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Epoch 48/50
1730
Epoch 49/50
1728
Epoch 50/50
1727
Epoch 1/50
2240
Epoch 2/50
2093
Epoch 3/50
2010
Epoch 4/50
1955
Epoch 5/50
1933
Epoch 6/50
1902
Epoch 7/50
1879
Epoch 8/50
1864
Epoch 9/50
1851
Epoch 10/50
1839
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Epoch 11/50
1827
Epoch 12/50
1823
Epoch 13/50
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Epoch 14/50
1802
Epoch 15/50
1804
Epoch 16/50
1793
Epoch 17/50
1787
Epoch 18/50
1784
Epoch 19/50
1779
Epoch 20/50
1772
Epoch 21/50
1769
Epoch 22/50
1766
Epoch 23/50
235/235 [============= ] - 3s 11ms/step - loss: 0.1747 - val loss: 0.
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Epoch 24/50
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Epoch 25/50
1759
Epoch 26/50
1756
Epoch 27/50
1756
Epoch 28/50
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Epoch 29/50
1749
Epoch 30/50
1747
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Epoch 31/50
1746
Epoch 32/50
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Epoch 33/50
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Epoch 34/50
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Epoch 35/50
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Epoch 36/50
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Epoch 37/50
1739
Epoch 38/50
1733
Epoch 39/50
1733
Epoch 40/50
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Epoch 41/50
1731
Epoch 42/50
1731
Epoch 43/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1690 - val loss: 0.
Epoch 44/50
1730
Epoch 45/50
1733
Epoch 46/50
1725
Epoch 47/50
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Epoch 48/50
1727
Epoch 49/50
1730
Epoch 50/50
1731
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Epoch 1/50
1823
Epoch 2/50
1577
Epoch 3/50
1480
Epoch 4/50
1439
Epoch 5/50
1412
Epoch 6/50
1395
Epoch 7/50
1375
Epoch 8/50
1360
Epoch 9/50
1350
Epoch 10/50
1341
Epoch 11/50
1331
Epoch 12/50
1324
Epoch 13/50
1318
Epoch 14/50
1311
Epoch 15/50
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Epoch 16/50
1303
Epoch 17/50
1297
Epoch 18/50
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Epoch 19/50
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Epoch 20/50
1282
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Epoch 21/50
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Epoch 22/50
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Epoch 23/50
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Epoch 24/50
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Epoch 25/50
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Epoch 26/50
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Epoch 27/50
1262
Epoch 28/50
1261
Epoch 29/50
1258
Epoch 30/50
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Epoch 31/50
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Epoch 32/50
1252
Epoch 33/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1237 - val loss: 0.
1252
Epoch 34/50
1247
Epoch 35/50
1247
Epoch 36/50
1246
Epoch 37/50
1246
Epoch 38/50
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Epoch 39/50
1243
Epoch 40/50
1242
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Epoch 41/50
1241
Epoch 42/50
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Epoch 43/50
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Epoch 44/50
1239
Epoch 45/50
1233
Epoch 46/50
1235
Epoch 47/50
1232
Epoch 48/50
1234
Epoch 49/50
1231
Epoch 50/50
1231
Epoch 1/50
1857
Epoch 2/50
1572
Epoch 3/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1517 - val loss: 0.
1469
Epoch 4/50
1430
Epoch 5/50
1407
Epoch 6/50
1388
Epoch 7/50
1372
Epoch 8/50
1363
Epoch 9/50
1348
Epoch 10/50
1340
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Epoch 11/50
1331
Epoch 12/50
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Epoch 13/50
1316
Epoch 14/50
1313
Epoch 15/50
1305
Epoch 16/50
1300
Epoch 17/50
1295
Epoch 18/50
1289
Epoch 19/50
1287
Epoch 20/50
1283
Epoch 21/50
1283
Epoch 22/50
1273
Epoch 23/50
235/235 [============== ] - 3s 11ms/step - loss: 0.1265 - val loss: 0.
1270
Epoch 24/50
1267
Epoch 25/50
1268
Epoch 26/50
1261
Epoch 27/50
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Epoch 28/50
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Epoch 29/50
1258
Epoch 30/50
1250
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Epoch 31/50
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Epoch 32/50
1253
Epoch 33/50
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Epoch 34/50
1248
Epoch 35/50
1247
Epoch 36/50
1244
Epoch 37/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1224 - val loss: 0.
1240
Epoch 38/50
1239
Epoch 39/50
1240
Epoch 40/50
1237
Epoch 41/50
1236
Epoch 42/50
1238
Epoch 43/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1213 - val loss: 0.
1236
Epoch 44/50
1233
Epoch 45/50
1233
Epoch 46/50
1230
Epoch 47/50
1234
Epoch 48/50
1230
Epoch 49/50
1229
Epoch 50/50
1229
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Epoch 1/50
1637
Epoch 2/50
1398
Epoch 3/50
1335
Epoch 4/50
1300
Epoch 5/50
1275
Epoch 6/50
1255
Epoch 7/50
1242
Epoch 8/50
1225
Epoch 9/50
1212
Epoch 10/50
1204
Epoch 11/50
1197
Epoch 12/50
1188
Epoch 13/50
235/235 [============= ] - 3s 13ms/step - loss: 0.1185 - val loss: 0.
1179
Epoch 14/50
1173
Epoch 15/50
1166
Epoch 16/50
1164
Epoch 17/50
1158
Epoch 18/50
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Epoch 19/50
1149
Epoch 20/50
1148
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Epoch 21/50
1146
Epoch 22/50
1141
Epoch 23/50
1138
Epoch 24/50
1136
Epoch 25/50
1133
Epoch 26/50
1134
Epoch 27/50
1129
Epoch 28/50
1127
Epoch 29/50
1128
Epoch 30/50
1125
Epoch 31/50
1123
Epoch 32/50
1121
Epoch 33/50
235/235 [============= ] - 3s 11ms/step - loss: 0.1108 - val loss: 0.
1120
Epoch 34/50
1120
Epoch 35/50
1118
Epoch 36/50
1117
Epoch 37/50
1116
Epoch 38/50
1112
Epoch 39/50
1114
Epoch 40/50
1113
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Epoch 41/50
1110
Epoch 42/50
1113
Epoch 43/50
1111
Epoch 44/50
1109
Epoch 45/50
1106
Epoch 46/50
1105
Epoch 47/50
1104
Epoch 48/50
1103
Epoch 49/50
1103
Epoch 50/50
1103
Epoch 1/50
1632
Epoch 2/50
1360
Epoch 3/50
235/235 [============= ] - 3s 11ms/step - loss: 0.1331 - val loss: 0.
1282
Epoch 4/50
1237
Epoch 5/50
1210
Epoch 6/50
1189
Epoch 7/50
1177
Epoch 8/50
1154
Epoch 9/50
1145
Epoch 10/50
1132
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Epoch 11/50
1125
Epoch 12/50
1117
Epoch 13/50
1113
Epoch 14/50
1104
Epoch 15/50
1098
Epoch 16/50
1093
Epoch 17/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1090 - val loss: 0.
1092
Epoch 18/50
1083
Epoch 19/50
1078
Epoch 20/50
1076
Epoch 21/50
1074
Epoch 22/50
1071
Epoch 23/50
235/235 [============= ] - 3s 12ms/step - loss: 0.1064 - val loss: 0.
1069
Epoch 24/50
1064
Epoch 25/50
1062
Epoch 26/50
1061
Epoch 27/50
1057
Epoch 28/50
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Epoch 29/50
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Epoch 30/50
1052
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Epoch 31/50
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Epoch 32/50
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Epoch 33/50
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Epoch 34/50
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Epoch 35/50
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Epoch 36/50
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Epoch 37/50
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Epoch 38/50
1041
Epoch 39/50
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Epoch 40/50
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Epoch 41/50
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Epoch 42/50
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Epoch 43/50
235/235 [============= ] - 3s 11ms/step - loss: 0.1022 - val loss: 0.
1034
Epoch 44/50
1035
Epoch 45/50
1032
Epoch 46/50
1033
Epoch 47/50
1034
Epoch 48/50
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Epoch 49/50
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Epoch 50/50
1029
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Epoch 1/50
1515
Epoch 2/50
1286
Epoch 3/50
1209
Epoch 4/50
1156
Epoch 5/50
1124
Epoch 6/50
1104
Epoch 7/50
1085
Epoch 8/50
1066
Epoch 9/50
1057
Epoch 10/50
1052
Epoch 11/50
1042
Epoch 12/50
1033
Epoch 13/50
1026
Epoch 14/50
1021
Epoch 15/50
1009
Epoch 16/50
1010
Epoch 17/50
1001
Epoch 18/50
1001
Epoch 19/50
0995
Epoch 20/50
0997
```

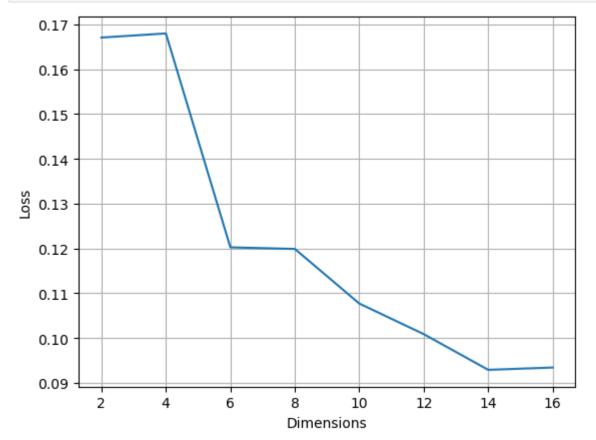
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Epoch 21/50
0988
Epoch 22/50
0984
Epoch 23/50
0987
Epoch 24/50
0979
Epoch 25/50
0978
Epoch 26/50
0975
Epoch 27/50
235/235 [============= ] - 3s 11ms/step - loss: 0.0969 - val loss: 0.
0970
Epoch 28/50
9972
Epoch 29/50
0968
Epoch 30/50
0968
Epoch 31/50
0966
Epoch 32/50
0962
Epoch 33/50
0962
Epoch 34/50
0963
Epoch 35/50
0960
Epoch 36/50
0959
Epoch 37/50
0956
Epoch 38/50
0955
Epoch 39/50
0955
Epoch 40/50
0953
```

```
Epoch 41/50
0951
Epoch 42/50
0951
Epoch 43/50
0949
Epoch 44/50
0952
Epoch 45/50
0948
Epoch 46/50
0951
Epoch 47/50
235/235 [============= ] - 3s 11ms/step - loss: 0.0935 - val loss: 0.
0947
Epoch 48/50
9946
Epoch 49/50
0947
Epoch 50/50
0945
Epoch 1/50
1554
Epoch 2/50
1294
Epoch 3/50
1197
Epoch 4/50
1149
Epoch 5/50
1120
Epoch 6/50
1100
Epoch 7/50
1084
Epoch 8/50
1067
Epoch 9/50
1058
Epoch 10/50
1047
```

```
Epoch 11/50
1038
Epoch 12/50
1030
Epoch 13/50
1026
Epoch 14/50
1019
Epoch 15/50
1018
Epoch 16/50
1011
Epoch 17/50
235/235 [============= ] - 3s 11ms/step - loss: 0.1009 - val loss: 0.
1007
Epoch 18/50
1001
Epoch 19/50
0998
Epoch 20/50
0994
Epoch 21/50
0992
Epoch 22/50
0987
Epoch 23/50
0984
Epoch 24/50
0980
Epoch 25/50
0981
Epoch 26/50
0984
Epoch 27/50
0974
Epoch 28/50
0974
Epoch 29/50
0968
Epoch 30/50
0969
```

```
Epoch 31/50
0968
Epoch 32/50
0966
Epoch 33/50
0965
Epoch 34/50
0963
Epoch 35/50
0960
Epoch 36/50
0962
Epoch 37/50
235/235 [============= ] - 3s 12ms/step - loss: 0.0953 - val loss: 0.
0962
Epoch 38/50
0957
Epoch 39/50
0958
Epoch 40/50
0956
Epoch 41/50
0958
Epoch 42/50
0955
Epoch 43/50
0954
Epoch 44/50
0951
Epoch 45/50
0952
Epoch 46/50
0953
Epoch 47/50
Epoch 48/50
0949
Epoch 49/50
0948
Epoch 50/50
0948
```

```
In [24]: plt.figure()
   plt.plot(dimensions, losses)
   plt.xlabel('Dimensions')
   plt.ylabel('Loss')
   plt.grid(True)
   plt.show()
```



1. **After** training an autoencoder with encoding_dim=8, apply noise (like the previous assignment) to *only* the input of the trained autoencoder (not the output). The output images should be without noise.

Print a few noisy images along with the output images to show they don't have noise.

```
In [60]:
         (xtrain, ytrain), (xtest, ytest) = mnist.load_data()
         xtrain = xtrain.astype('float32') / 255.
         xtest = xtest.astype('float32') / 255.
         xtrain = xtrain.reshape((len(xtrain), np.prod(xtrain.shape[1:])))
         xtest = xtest.reshape((len(xtest), np.prod(xtest.shape[1:])))
         xtrain.shape, xtest.shape
         ((60000, 784), (10000, 784))
Out[60]:
         encoding_dim = 8
In [61]:
         x = input_img = Input(shape=(784,))
         x = Dense(256, activation='relu')(x)
         x = Dense(128, activation='relu')(x)
         encoded = Dense(encoding dim, activation='relu')(x)
         x = Dense(128, activation='relu')(encoded)
```

```
x = Dense(256, activation='relu')(x)
         decoded = Dense(784, activation='sigmoid')(x)
         autoencoder = Model(input_img, decoded)
         encoder = Model(input_img, encoded)
         encoded input = Input(shape=(encoding dim,))
         dcd1 = autoencoder.layers[-1]
         dcd2 = autoencoder.layers[-2]
         dcd3 = autoencoder.layers[-3]
          decoder = Model(encoded_input, dcd1(dcd2(dcd3(encoded_input))))
In [62]:
         autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
In [58]:
         noise = np.random.normal(20, 4, (60000,784))
         noisy_train = noise + xtrain
In [63]:
         autoencoder.fit(noisy_train, xtrain,
                          epochs=100,
                          batch size=256,
                          shuffle=True,
                          validation_data=(xtest, xtest),
                          callbacks=[TensorBoard(log dir='/tmp/autoencoder')])
```

```
Epoch 1/100
685
Epoch 2/100
Epoch 3/100
572
Epoch 4/100
6444
Epoch 5/100
6206
Epoch 6/100
6024
Epoch 7/100
5792
Epoch 8/100
5618
Epoch 9/100
5504
Epoch 10/100
5438
Epoch 11/100
5287
Epoch 12/100
5190
Epoch 13/100
5072
Epoch 14/100
4985
Epoch 15/100
4877
Epoch 16/100
4738
Epoch 17/100
Epoch 18/100
4576
Epoch 19/100
4485
Epoch 20/100
4390
```

```
Epoch 21/100
4275
Epoch 22/100
4232
Epoch 23/100
4100
Epoch 24/100
3994
Epoch 25/100
3897
Epoch 26/100
3812
Epoch 27/100
3706
Epoch 28/100
3594
Epoch 29/100
3488
Epoch 30/100
3372
Epoch 31/100
3288
Epoch 32/100
3166
Epoch 33/100
Epoch 34/100
3000
Epoch 35/100
2982
Epoch 36/100
2935
Epoch 37/100
Epoch 38/100
2902
Epoch 39/100
2859
Epoch 40/100
2856
```

```
Epoch 41/100
2873
Epoch 42/100
2839
Epoch 43/100
2840
Epoch 44/100
2807
Epoch 45/100
2827
Epoch 46/100
2798
Epoch 47/100
2804
Epoch 48/100
2792
Epoch 49/100
2790
Epoch 50/100
2819
Epoch 51/100
2799
Epoch 52/100
2769
Epoch 53/100
Epoch 54/100
2803
Epoch 55/100
2763
Epoch 56/100
2788
Epoch 57/100
2783
Epoch 58/100
2755
Epoch 59/100
2769
Epoch 60/100
2754
```

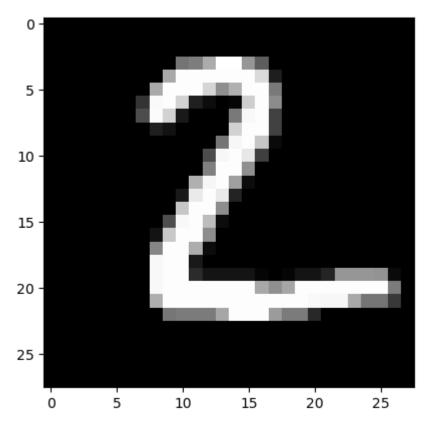
```
Epoch 61/100
2751
Epoch 62/100
2731
Epoch 63/100
2733
Epoch 64/100
2720
Epoch 65/100
2718
Epoch 66/100
2725
Epoch 67/100
2710
Epoch 68/100
2720
Epoch 69/100
2729
Epoch 70/100
2698
Epoch 71/100
2690
Epoch 72/100
2754
Epoch 73/100
Epoch 74/100
2685
Epoch 75/100
2701
Epoch 76/100
2691
Epoch 77/100
2735
Epoch 78/100
2713
Epoch 79/100
2703
Epoch 80/100
2696
```

```
Epoch 81/100
2728
Epoch 82/100
2701
Epoch 83/100
2714
Epoch 84/100
2714
Epoch 85/100
2717
Epoch 86/100
2711
Epoch 87/100
2706
Epoch 88/100
2701
Epoch 89/100
2706
Epoch 90/100
2687
Epoch 91/100
2691
Epoch 92/100
2733
Epoch 93/100
Epoch 94/100
2709
Epoch 95/100
2685
Epoch 96/100
2710
Epoch 97/100
Epoch 98/100
2682
Epoch 99/100
2698
Epoch 100/100
2686
```

```
<keras.src.callbacks.History at 0x2204524ed10>
Out[63]:
        encoded imgs = encoder.predict(xtest)
In [73]:
        decoded_imgs = decoder.predict(encoded_imgs)
        import matplotlib.pyplot as plt
        n = 20 # how many digits we will display
        plt.figure(figsize=(40, 4))
        for i in range(n):
            # display original
            ax = plt.subplot(2, n, i + 1)
            plt.imshow(xtest[i].reshape(28, 28))
            plt.gray()
            ax.get_xaxis().set_visible(False)
            ax.get_yaxis().set_visible(False)
            # display reconstruction
            ax = plt.subplot(2, n, i + 1 + n)
            plt.imshow(decoded_imgs[i].reshape(28, 28))
            plt.gray()
            ax.get_xaxis().set_visible(False)
            ax.get_yaxis().set_visible(False)
        plt.show()
        313/313 [========== ] - 0s 1ms/step
        313/313 [========== ] - 0s 1ms/step
                            149590690159734
In [71]: plt.imshow(xtest[1].reshape(28,28))
```

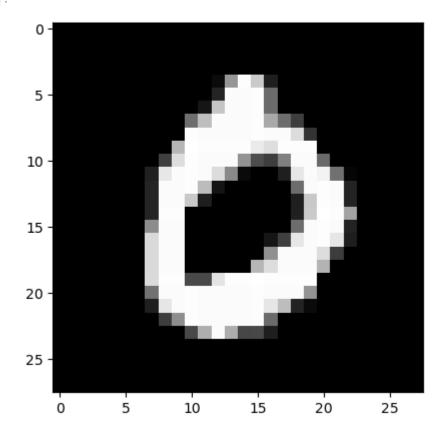
<matplotlib.image.AxesImage at 0x220270c0760> Out[71]:

file:///C:/Users/jorda/Downloads/MLNN Module12 (1).html



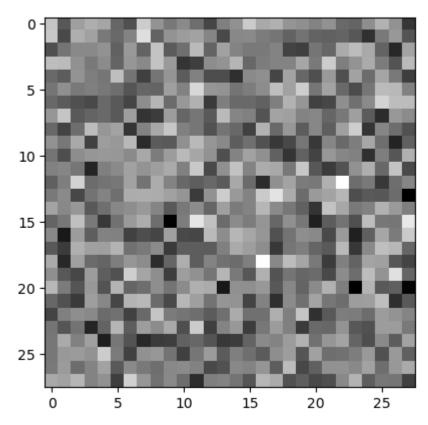
In [75]: plt.imshow(xtest[3].reshape(28,28))

Out[75]: <matplotlib.image.AxesImage at 0x220281f3520>



In [76]: plt.imshow(noisy_train[3].reshape(28,28))

Out[76]: <matplotlib.image.AxesImage at 0x2202826a620>



In []: