```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Dense
import matplotlib.pyplot as plt
from tensorflow.keras.datasets import mnist
(x_train, _), (x_test, _) = mnist.load_data()
x_train = x_train.astype('float32') / 255.0
x_{\text{test}} = x_{\text{test.astype}}('float32') / 255.0
x_train = x_train.reshape((len(x_train), np.prod(x_train.shape[1:])))
x_test = x_test.reshape((len(x_test), np.prod(x_test.shape[1:])))
input_dim = x_train.shape[1]
encoding dim = 64
input_img = Input(shape=(input_dim,))
encoded = Dense(encoding_dim, activation='relu')(input_img)
decoded = Dense(input_dim, activation='sigmoid')(encoded)
autoencoder = Model(input_img, decoded)
encoder = Model(input_img, encoded)
encoded_input = Input(shape=(encoding_dim,))
decoder layer = autoencoder.layers[-1]
decoder = Model(encoded_input, decoder_layer(encoded_input))
autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
autoencoder.fit(x_train, x_train,
                epochs=50,
                batch_size=256,
                shuffle=True,
                validation_data=(x_test, x_test))
encoded_imgs = encoder.predict(x_test)
decoded_imgs = decoder.predict(encoded_imgs)
n = 10
plt.figure(figsize=(20, 4))
for i in range(n):
   ax = plt.subplot(2, n, i + 1)
   plt.imshow(x_test[i].reshape(28, 28))
   plt.gray()
   ax.get_xaxis().set_visible(False)
   ax.get_yaxis().set_visible(False)
   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()
```

11490434/11490434	Os Ous/step
Epoch 1/50	F
235/235	- 11s 31ms/step - loss: 0.3441 - val_loss: 0.1606
Epoch 2/50	
	- 6s 14ms/step - loss: 0.1505 - val_loss: 0.1242
Epoch 3/50	- 2c Ome/cton locar 0 1202 val locar 0 1066
235/235 ——————— Epoch 4/50	- 2s 9ms/step - loss: 0.1202 - val_loss: 0.1066
•	- 2s 9ms/step - loss: 0.1048 - val_loss: 0.0964
Epoch 5/50	·
235/235 ————	- 3s 10ms/step - loss: 0.0955 - val_loss: 0.0895
Epoch 6/50	
	- 3s 11ms/step - loss: 0.0892 - val_loss: 0.0847
Epoch 7/50 235/235 ————————————————————————————————————	- 3s 13ms/step - loss: 0.0845 - val_loss: 0.0813
Epoch 8/50	33 13m3/3ccp 1033. 0.0043 Var_1033. 0.0013
•	- 2s 9ms/step - loss: 0.0814 - val_loss: 0.0790
Epoch 9/50	
235/235	- 3s 9ms/step - loss: 0.0793 - val_loss: 0.0773
Epoch 10/50	2- 0/
	- 2s 9ms/step - loss: 0.0778 - val_loss: 0.0762
Epoch 11/50 235/235 ————————————————————————————————————	- 2s 9ms/step - loss: 0.0768 - val_loss: 0.0756
Epoch 12/50	. ,
	- 4s 15ms/step - loss: 0.0760 - val_loss: 0.0750
Epoch 13/50	
	- 4s 9ms/step - loss: 0.0753 - val_loss: 0.0745
Epoch 14/50 235/235 ————————————————————————————————————	- 2s 9ms/step - loss: 0.0750 - val_loss: 0.0742
Epoch 15/50	
	- 2s 9ms/step - loss: 0.0745 - val_loss: 0.0738
Epoch 16/50	
	- 3s 12ms/step - loss: 0.0744 - val_loss: 0.0736
Epoch 17/50	- Fr 10ms/ston loss, 0 0742 wellloss, 0 0725
235/235 ————————————————————————————————————	- 5s 10ms/step - loss: 0.0743 - val_loss: 0.0735
235/235	- 3s 10ms/step - loss: 0.0740 - val_loss: 0.0733
Epoch 19/50	· -
235/235 —	- 3s 10ms/step - loss: 0.0739 - val_loss: 0.0732
Epoch 20/50	2. 42/
235/235 ————————————————————————————————————	- 3s 13ms/step - loss: 0.0737 - val_loss: 0.0731
	- 4s 10ms/step - loss: 0.0735 - val_loss: 0.0730
Epoch 22/50	
235/235 —————	- 2s 10ms/step - loss: 0.0736 - val_loss: 0.0730
Epoch 23/50	
235/235 ————————————————————————————————————	- 3s 10ms/step - loss: 0.0737 - val_loss: 0.0729
· ·	- 3s 12ms/step - loss: 0.0735 - val loss: 0.0728
Epoch 25/50	33 12ms/seep 1055. 0.0/55 var_1055. 0.0/20
235/235	- 3s 13ms/step - loss: 0.0733 - val_loss: 0.0727
Epoch 26/50	
235/235 ————————————————————————————————————	- 2s 10ms/step - loss: 0.0731 - val_loss: 0.0727
Epoch 27/50 235/235 ————————————————————————————————————	- 3s 10ms/step - loss: 0.0731 - val loss: 0.0727
Epoch 28/50	33 τομο/οιερ - 1000. 0.0/01 - Val_1055: 0.0/2/
235/235	- 2s 10ms/step - loss: 0.0731 - val_loss: 0.0726
Epoch 29/50	-
	- 3s 12ms/step - loss: 0.0731 - val_loss: 0.0726
Epoch 30/50	1e 12ms/ston loss 0.0724
235/235 ————————————————————————————————————	- 3s 13ms/step - loss: 0.0731 - val_loss: 0.0726
•	- 4s 10ms/step - loss: 0.0730 - val_loss: 0.0726
Epoch 32/50	
	- 2s 10ms/step - loss: 0.0729 - val_loss: 0.0725
Epoch 33/50	2.44 / 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
	- 3s 11ms/step - loss: 0.0731 - val_loss: 0.0725
Epoch 34/50 235/235 ————————————————————————————————————	- 5s 10ms/step - loss: 0.0731 - val_loss: 0.0724
Epoch 35/50	
•	- 2s 10ms/step - loss: 0.0730 - val_loss: 0.0724
Epoch 36/50	
	- 2s 10ms/step - loss: 0.0729 - val_loss: 0.0724
Epoch 37/50	3c 11mc/cton - loccy 0 0720 - val loccy 0 0724
235/235 ————————————————————————————————————	- 3s 11ms/step - loss: 0.0729 - val_loss: 0.0724
235/235	- 5s 11ms/step - loss: 0.0729 - val loss: 0.0724
	,
Epoch 39/50	
Epoch 39/50 235/235 ————————————————————————————————————	- 2s 10ms/step - loss: 0.0729 - val_loss: 0.0725
Epoch 39/50 235/235 Epoch 40/50	-
Epoch 39/50 235/235 Epoch 40/50	- 2s 10ms/step - loss: 0.0729 - val_loss: 0.0725 - 3s 10ms/step - loss: 0.0729 - val_loss: 0.0723