```
In [3]: | import tensorflow as tf
        from tensorflow.keras import layers, models
        from tensorflow.keras.datasets import mnist
        import matplotlib.pyplot as plt
        (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(-1, 28*28)}
        x_{\text{test}} = x_{\text{test.reshape}}(-1, 28*28)
        input_img = layers.Input(shape=(28 * 28,))
        encoded = layers.Dense(128, activation='relu')(input_img)
        decoded = layers.Dense(28 * 28, activation='sigmoid')(encoded)
        autoencoder = models.Model(input_img, decoded)
        autoencoder.compile(optimizer='adam', loss='binary crossentropy')
        autoencoder.fit(x train, x train, epochs=10, batch size=256, validation data=(x 1
        reconstruction loss = autoencoder.evaluate(x test, x test)
        print(f"Reconstruction Loss: {reconstruction loss}5")
        decoded_imgs = autoencoder.predict(x_test)
        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), cmap="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), cmap="gray")
            ax.get xaxis().set visible(False)
            ax.get_yaxis().set_visible(False)
        plt.show()
        Epoch 1/10
                                      - 2s 5ms/step - loss: 0.3137 - val loss: 0.1347
        235/235 -
        Epoch 2/10
                                     - 1s 5ms/step - loss: 0.1259 - val_loss: 0.1028
        235/235 -
```

```
Epoch 3/10
235/235 -
                             • 1s 5ms/step - loss: 0.0993 - val loss: 0.0885
Epoch 4/10
235/235 -
                             • 1s 5ms/step - loss: 0.0875 - val loss: 0.0812
Epoch 5/10
235/235 -
                            - 1s 5ms/step - loss: 0.0808 - val_loss: 0.0769
Epoch 6/10
235/235 -
                            - 1s 5ms/step - loss: 0.0770 - val_loss: 0.0741
Epoch 7/10
235/235
                             1s 5ms/step - loss: 0.0741 - val_loss: 0.0723
Epoch 8/10
235/235 -
                             • 1s 5ms/step - loss: 0.0724 - val_loss: 0.0710
Epoch 9/10
235/235
                            - 1s 5ms/step - loss: 0.0712 - val_loss: 0.0701
Epoch 10/10
235/235
                             • 1s 5ms/step - loss: 0.0703 - val_loss: 0.0694
                            - 0s 865us/step - loss: 0.0691
313/313 -
Reconstruction Loss: 0.0693804919719696
313/313 •
                            - 0s 536us/step
```



```
In [5]: import numpy as np
        import tensorflow as tf
        from tensorflow.keras import layers, models
        import matplotlib.pyplot as plt
        (x_train, _), (x_test, _) = tf.keras.datasets.mnist.load_data()
        x_train = x_train.astype('float32') / 255.0
        x_{\text{test}} = x_{\text{test.astype}}('float32') / 255.0
        x train flat = x train.reshape((x train.shape[0], 28 * 28))
        x_test_flat = x_test.reshape((x_test.shape[0], 28 * 28))
        def add_noise(images, noise_factor=0.3):
            noise = np.random.normal(loc=0.0, scale=1.0, size=images.shape)
            noisy images = images + noise factor * noise
            noisy_images = np.clip(noisy_images, 0.0, 1.0) # Ensure values are in [0, 1]
            return noisy images
        x train noisy = add noise(x train flat)
        x test noisy = add noise(x test flat)
        encoder input = layers.Input(shape=(784,))
        encoded = layers.Dense(128, activation='relu')(encoder input)
        encoded = layers.Dense(64, activation='relu')(encoded)
        encoded = layers.Dense(32, activation='relu')(encoded)
        decoded = layers.Dense(64, activation='relu')(encoded)
        decoded = layers.Dense(128, activation='relu')(decoded)
        decoded = layers.Dense(784, activation='sigmoid')(decoded)
        autoencoder = models.Model(encoder_input, decoded)
        autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
        autoencoder.fit(x train noisy, x train flat, epochs=10, batch size=256, validatid
        decoded imgs = autoencoder.predict(x test noisy)
        decoded_imgs = decoded_imgs.reshape((x_test.shape[0], 28, 28))
        n = 10
        plt.figure(figsize=(20, 6))
        for i in range(n):
            ax = plt.subplot(3, n, i + 1)
            plt.imshow(x test noisy[i].reshape(28, 28), cmap='gray')
            ax.axis('off')
            ax = plt.subplot(3, n, i + 1 + n)
            plt.imshow(x_test[i], cmap='gray')
            ax.axis('off')
            ax = plt.subplot(3, n, i + 1 + 2 * n)
            plt.imshow(decoded imgs[i], cmap='gray')
            ax.axis('off')
```

```
Epoch 1/10
                             - 2s 6ms/step - loss: 0.3351 - val_loss: 0.1901
235/235 •
Epoch 2/10
235/235
                              1s 6ms/step - loss: 0.1755 - val_loss: 0.1491
Epoch 3/10
235/235 -
                             • 1s 6ms/step - loss: 0.1469 - val_loss: 0.1387
Epoch 4/10
                             • 1s 6ms/step - loss: 0.1378 - val_loss: 0.1323
235/235 -
Epoch 5/10
                             1s 6ms/step - loss: 0.1320 - val loss: 0.1276
235/235 -
Epoch 6/10
235/235 -
                             • 1s 6ms/step - loss: 0.1272 - val_loss: 0.1227
Epoch 7/10
235/235 -
                             1s 6ms/step - loss: 0.1226 - val_loss: 0.1200
Epoch 8/10
                             1s 6ms/step - loss: 0.1201 - val loss: 0.1174
235/235 -
Epoch 9/10
235/235 -
                             - 1s 6ms/step - loss: 0.1180 - val_loss: 0.1162
Epoch 10/10
235/235 -
                             • 1s 6ms/step - loss: 0.1165 - val_loss: 0.1146
313/313 -
                             0s 684us/step
```

In []: