```
Loading MNIST
      from keras.datasets import mnist
 In [10]:
      (x_train,_),(x_test,_)=mnist.load_data()
     Checking Dimensions
 In [11]:
      print(x_train.shape)
      (60000, 28, 28)
     Image Pre-processing
      def pre_process(X):
 In [12]:
        X = X/255.0
        X = X.reshape((len(X), 784))
        return X
      x_train = pre_process(x_train)
      x_test = pre_process(x_test)
      print("X_train", x_train.shape)
      print("X_test", x_test.shape)
      X_train (60000, 784)
      X_test (10000, 784)
     Visualized the training images
 In [13]:
      import matplotlib.pyplot as plt
      def show_data(X, n=10, height=28, width=28, title=""):
        plt.figure(figsize=(20, 5))
        for i in range(n):
          ax = plt.subplot(2, n, i+1)
          plt.imshow(X[i].reshape((height,width)))
          plt.gray()
          ax.get_xaxis().set_visible(False)
          ax.get_yaxis().set_visible(False)
        plt.suptitle(title, fontsize = 20)
 In [14]:
      show_data(x_train, title="Training images")
                             Training images
      5041921314
     Building the Stacked Autoencoder

    Input layer = 784 neurons

      • Hidden layer_1 = 256 neurons

    Code layer = 48 neurons

    Hidden layer_2 = 256 neurons

    Output layer = 784 neurons

      from keras.layers import Input, Dense
 In [15]:
      from keras.models import Model
 In [16]:
      # encoder
      input_layer = Input(shape=(784,), name="INPUT")
      hidden_layer_1 = Dense(256, activation='relu', name="HIDDEN_1")(input_layer)
      code_layer = Dense(100, activation='relu', name="CODE")(hidden_layer_1)
      # decoder
      hidden_layer_2 = Dense(256, activation='relu', name="HIDDEN_4")(code_layer)
      output_layer = Dense(784, activation='sigmoid', name="OUTPUT")(hidden_layer_2)
     Compiling the layers
      stacked_autoencoder=Model(input_layer, output_layer)
 In [17]:
      stacked_autoencoder.compile(optimizer="Adam",loss="binary_crossentropy")
      stacked_autoencoder.summary()
      Model: "model_1"
      Layer (type)
                    Output Shape
                                  Param #
      INPUT (InputLayer)
                    [(None, 784)]
                                  0
      HIDDEN_1 (Dense)
                                  200960
                    (None, 256)
      CODE (Dense)
                     (None, 100)
                                  25700
      HIDDEN_4 (Dense)
                     (None, 256)
                                  25856
      OUTPUT (Dense)
                                  201488
                     (None, 784)
      Total params: 454,004
      Trainable params: 454,004
      Non-trainable params: 0
     Fit into the neural network
      stacked_autoencoder.fit(x_train, x_train, epochs=100, batch_size=256, shuffle=True,
 In [18]:
                  validation_data=(x_test, x_test))
      Epoch 1/100
      235/235 [==
                     Epoch 2/100
                      ======] - 7s 28ms/step - loss: 0.1096 - val_loss: 0.0946
      235/235 [===
      Epoch 3/100
                      235/235 [==
      Epoch 4/100
                    ========] - 7s 30ms/step - loss: 0.0870 - val_loss: 0.0828
      235/235 [===
      Epoch 5/100
                   235/235 [==
      Epoch 6/100
                     =======] - 7s 29ms/step - loss: 0.0805 - val_loss: 0.0781
      235/235 [==
      Epoch 7/100
                      ======] - 7s 29ms/step - loss: 0.0786 - val_loss: 0.0769
      235/235 [==
      Epoch 8/100
                      ======] - 6s 27ms/step - loss: 0.0773 - val_loss: 0.0758
      235/235 [===
      Epoch 9/100
                      ======] - 6s 26ms/step - loss: 0.0761 - val_loss: 0.0749
      235/235 [==
      Epoch 10/100
                      ======] - 7s 29ms/step - loss: 0.0753 - val_loss: 0.0742
      235/235 [===
      Epoch 11/100
                      ======] - 8s 32ms/step - loss: 0.0744 - val_loss: 0.0733
      235/235 [==
      Epoch 12/100
                    =======] - 7s 30ms/step - loss: 0.0737 - val_loss: 0.0728
      235/235 [====
      Epoch 13/100
      235/235 [===
                    =======] - 7s 30ms/step - loss: 0.0731 - val_loss: 0.0723
      Epoch 14/100
                     =======] - 7s 30ms/step - loss: 0.0726 - val_loss: 0.0718
      235/235 [===
      Epoch 15/100
                       =====] - 7s 29ms/step - loss: 0.0722 - val_loss: 0.0713
      235/235 [===
      Epoch 16/100
      235/235 [===
                      ======] - 7s 30ms/step - loss: 0.0718 - val_loss: 0.0712
      Epoch 17/100
                      ======] - 7s 28ms/step - loss: 0.0714 - val_loss: 0.0709
      235/235 [==
      Epoch 18/100
                     =======] - 7s 28ms/step - loss: 0.0710 - val_loss: 0.0705
      235/235 [====
      Epoch 19/100
                         235/235 [==
      Epoch 20/100
      Epoch 21/100
             235/235 [====
      Epoch 22/100
                235/235 [====
      Epoch 23/100
              235/235 [====
      Epoch 24/100
                   :========] - 6s 26ms/step - loss: 0.0696 - val_loss: 0.0699
      235/235 [====
      Epoch 25/100
                      ======] - 6s 26ms/step - loss: 0.0694 - val_loss: 0.0691
      235/235 [====
      Epoch 26/100
                    =======] - 6s 25ms/step - loss: 0.0693 - val_loss: 0.0690
      235/235 [====
      Epoch 27/100
                  =========] - 6s 25ms/step - loss: 0.0693 - val_loss: 0.0689
      235/235 [====
      Epoch 28/100
      Epoch 29/100
      235/235 [====
             Epoch 30/100
                235/235 [====
      Epoch 31/100
      235/235 [====
                Epoch 32/100
              235/235 [====
      Epoch 33/100
                    235/235 [====
      Epoch 34/100
      235/235 [====
                  =========] - 6s 26ms/step - loss: 0.0683 - val_loss: 0.0682
      Epoch 35/100
                   235/235 [====
      Epoch 36/100
                235/235 [========
      Epoch 37/100
               235/235 [====
      Epoch 38/100
                 235/235 [====
      Epoch 39/100
                 235/235 [=====
      Epoch 40/100
      235/235 [======
                   =========] - 6s 25ms/step - loss: 0.0679 - val_loss: 0.0677
      Epoch 41/100
      235/235 [======
                    Epoch 42/100
                    ========] - 6s 27ms/step - loss: 0.0677 - val_loss: 0.0678
      235/235 [======
      Epoch 43/100
                   =========] - 6s 27ms/step - loss: 0.0676 - val_loss: 0.0676
      235/235 [=======
      Epoch 44/100
      Epoch 45/100
      Epoch 46/100
                 235/235 [=====
      Epoch 47/100
      235/235 [=====
                  ==========] - 6s 27ms/step - loss: 0.0676 - val_loss: 0.0675
      Epoch 48/100
                 235/235 [======
      Epoch 49/100
      235/235 [=====
                    Epoch 50/100
                    235/235 [=====
      Epoch 51/100
      235/235 [========
                   Epoch 52/100
                 235/235 [========
      Epoch 53/100
      235/235 [====
                  Epoch 54/100
                 235/235 [====
      Epoch 55/100
      235/235 [=====
                   =========] - 6s 27ms/step - loss: 0.0671 - val_loss: 0.0673
      Epoch 56/100
                   =========] - 6s 27ms/step - loss: 0.0672 - val_loss: 0.0673
      235/235 [====
      Epoch 57/100
      235/235 [====
                   Epoch 58/100
                   235/235 [====
      Epoch 59/100
      235/235 [==
                           - 6s 26ms/step - loss: 0.0671 - val loss: 0.0671
      Epoch 60/100
      Epoch 61/100
      Epoch 62/100
      Epoch 63/100
      Epoch 64/100
      Epoch 65/100
      Epoch 66/100
      Epoch 67/100
      Epoch 68/100
      Epoch 69/100
      Epoch 70/100
      Epoch 71/100
      Epoch 72/100
      Epoch 73/100
      Epoch 74/100
      Epoch 75/100
      Epoch 76/100
      Epoch 77/100
      Epoch 78/100
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      Epoch 80/100
      Epoch 81/100
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      Epoch 96/100
      Epoch 97/100
      Epoch 98/100
      Epoch 99/100
      235/235 [==
                    =======] - 8s 33ms/step - loss: 0.0665 - val_loss: 0.0667
      Epoch 100/100
      Out[18]: <tensorflow.python.keras.callbacks.History at 0x7fe40df064f0>
     Predicting
      decoded_data = stacked_autoencoder.predict(x_test)
 In [19]:
     Visualization
      show_data(x_test, title="original data")
 In [20]:
      show_data(decoded_data, title="decoded data")
                             original data
            2 / 0 4 /
                                      495
                             decoded data
            21041495
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