Importing dataset and preprocessing the data

```
In [2]:
```

```
from keras.datasets import mnist
import numpy as np

(x_train, _), (x_test, _) = mnist.load_data()
# Reshape
x_train = np.reshape(x_train, (len(x_train), 28, 28, 1))
x_test = np.reshape(x_test, (len(x_test), 28, 28, 1))
```

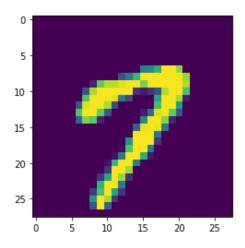
Visualising input image

In [3]:

```
from matplotlib import pyplot as plt
import numpy as np

first_image = x_train[15]
first_image = np.asarray(first_image)
print(first_image.shape)
pixels = first_image.reshape((28, 28))
plt.imshow(pixels)
plt.show()
```

```
(28, 28, 1)
```



In [4]:

```
# image_array size - w*h , d
img = first_image
print(first_image.shape)
w, h, d = first_image.shape
image_array = img.reshape(w*h,d) #Flatten
print('ReShaped'.center(20,'='))
print(image_array.shape)
```

```
(28, 28, 1)
=====ReShaped======
(784, 1)
```

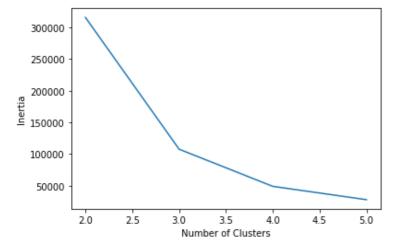
Analysing the relationship between the number of clusters and the inertia to find the optimum number of clusters

```
In [6]:
```

```
# for clustering image pixels
from sklearn.cluster import KMeans
x cluster = []
y inertia = []
def calculate metrics(model):
x cluster.append(model.n clusters)
y_inertia.append(model.inertia_)
from sklearn import metrics
cluster_number = [2,3,4,5]
kmeans number = []
for i in cluster number:
 #total clusters = len(np.unique(image array))
  # Initialize the K-Means model
 kmeans = KMeans(n clusters = i,random state=42,verbose=2,n jobs=-1).fit(image array)
 kmeans number.append(kmeans)
  # Calculating the metrics
 calculate metrics (kmeans)
```

In [7]:

```
# plotting the points
plt.plot(x_cluster, y_inertia)
# naming the x axis
plt.xlabel('Number of Clusters')
# naming the y axis
plt.ylabel('Inertia')
plt.show()
```



Clustering the pixel points

```
In [8]:
```

```
kmeans64 = KMeans(n_clusters = 5,random_state=42,verbose=2,n_jobs=-1).fit(image_array)
```

Constructing the Compressed image

```
In [9]:
```

```
def recreate_image(kmeans):
   compressed_image_array = kmeans.cluster_centers_[kmeans.labels_] #[0,3,2,1,.....]
   #Reshape the image to original dimension
   compressed_image = compressed_image_array.reshape(w,h)
   #Save and display output image
   return compressed_image
```

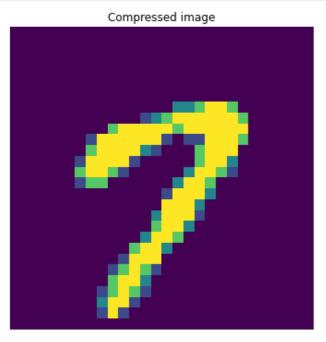
Compare the original and the reconstructed image

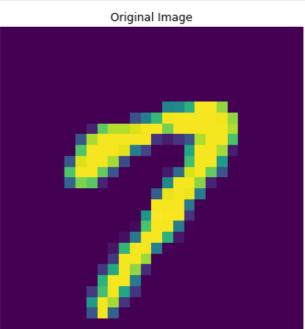
```
In [10]:

def nLabels(kmeans):
   return kmeans.n_clusters

In [11]:
```

```
def plot_img(kmeans):
    str_n = "No. of clusters: "+str(nLabels(kmeans))
    plt.figure(figsize=(20,10))
    plt.title(str_n)
    plt.subplot(132)
    plt.axis('off')
    plt.title('Original Image')
    plt.imshow(pixels)
    plt.subplot(131)
    plt.axis('off')
    plt.title('Compressed image')
    op_img = recreate_image(kmeans)
    plt.imshow(op_img)
```

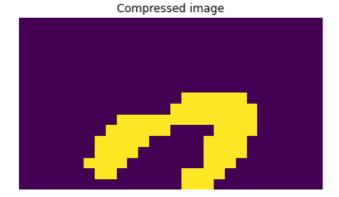




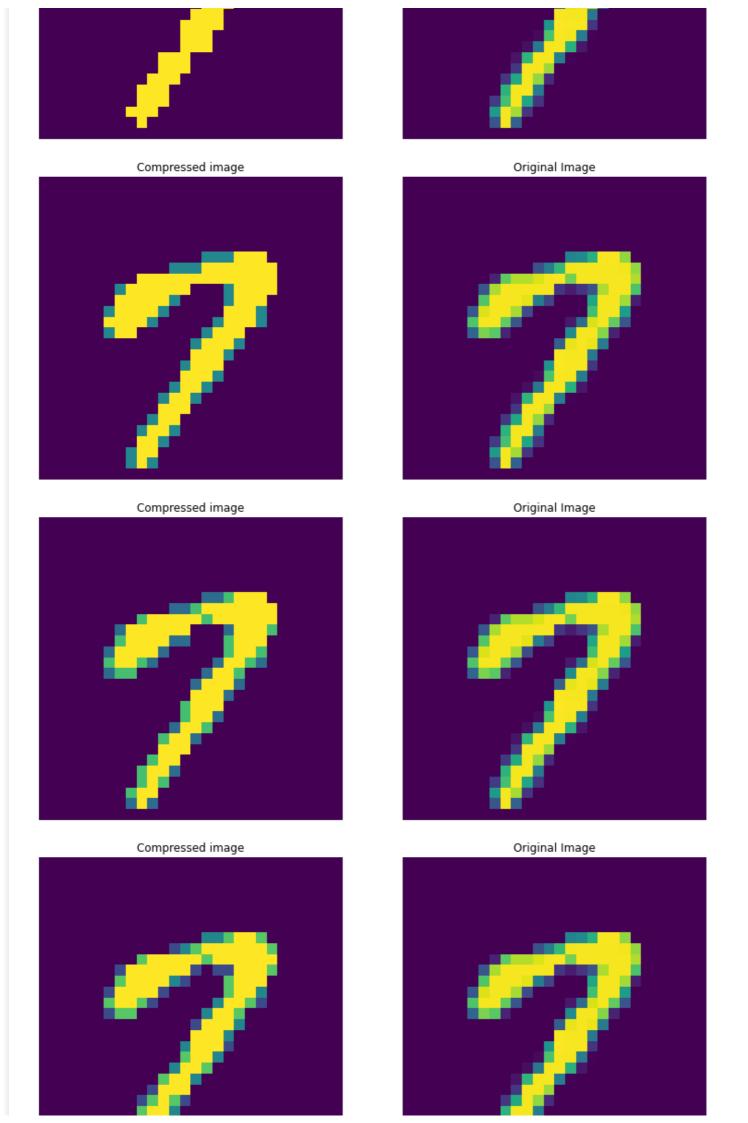
Visualising the compressed image for all the clusters

```
In [13]:
```

```
from sklearn import metrics
for i in range(len(cluster_number)):
    #total_clusters = len(np.unique(image_array))
    # Initialize the K-Means model
    plot_img(kmeans_number[i])
```







Analysing PSNR value for different cluster values

```
In [17]:
```

In [18]:

```
# plotting the points
plt.plot(x_cluster, y_psnr)
# naming the x axis
plt.xlabel('Number of Clusters')
# naming the y axis
plt.ylabel('PSNR')
plt.show()
```

