	0, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0
	0, 0], (0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 80, 156, 107, 253, 253, 205, 11, 0, 43, 154, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
	0, 45, 186, 253, 253, 150, 27, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
	[ 0, 0, 0, 0, 0, 0, 18, 171, 219, 253, 253, 253, 253, 195, 80, 9, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
	tlib.image.AxesImage at 0x2a85f019100> 5 10 15 20 25
0 0 -	tshow(X_train[1])  tlib.image.AxesImage at 0x2a85f0ed130>  5 10 15 20 25
	n[0]  n = X_train / 255 = X_test / 255
X_train	
	0.
	0.65098039, 1. , 0.96862745, 0.49803922, 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0
	0.
	[0.
	[0.
	0.50980392, 0.71764706, 0.99215686, 0.99215686, 0.81176471, 0.00784314, 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0.
	0.77647059, 0.31764706, 0.00784314, 0. , 0. , 0. , 0. , 0. , 0. , 0. , 0.
	0.         , 0.         ,
X_test  X_train  (60000)	0. , 0. , 0. ]])  n_flattened = X_train.reshape(len(X_train), 28*28)  flattened = X_test.reshape(len(X_test), 28*28)  n_flattened.shape  784)  n_flattened[0]
	0.
	0.
	0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.99215686, 0.98431373, 0.36470588, 0.32156863, 0.32156863, 0.21960784, 0.15294118, 0. , , , , , , , , , , , , , , , , , ,
	0.
	0.
	0.
	0. , 0. , 0. , 0. , 0. , 0. , 0. , 0. ,
	0. 83137255, 0.52941176, 0.51764706, 0.0627451 , 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
model :  model :  Epoch : 1875/18 Epoch : 1875/18	75 [====================================
Epoch 4 1875/18 Epoch 9 1875/18 : <tensor< td=""><td>75 [====================================</td></tensor<>	75 [====================================
y_pred array(	9.1807551e-06, 1.1243588e-10, 4.0124105e-05, 8.6890757e-03, 1.3472301e-06, 8.3354549e-05, 1.4762296e-09, 7.1893895e-01, 7.2274859e-05, 1.2923479e-03], dtype=float32)  tshow(X_test[0])  tlib.image.AxesImage at 0x2a85ebf7a00>  5 10 15 20 25
15 - 20 - 25 - np.argn	ax finds a maximum element from an array and returns the index of it  max(y_predicted[0])
y_pred: [7, 2, cm = t cm	f.math.confusion_matrix(labels=y_test,predictions=y_predicted_labels) sor: shape=(10, 10), dtype=int32, numpy=
Collection Down Require Require Require Require Require	[ 1, 1, 6, 0, 912, 0, 9, 3, 9, 41], [ 8, 2, 4, 32, 11, 776, 13, 3, 34, 9], [ 6, 3, 8, 1, 8, 14, 914, 2, 2, 0], [ 1, 7, 26, 4, 8, 1, 0, 930, 2, 49], [ 4, 5, 7, 14, 10, 20, 9, 9, 879, 17], [ 7, 6, 2, 10, 19, 4, 0, 8, 8, 945]])>  Install seaborn  ing seaborn  oading seaborn-0.11.1-py3-none-any.whl (285 kB)  ment already satisfied: matplotlib>=2.2 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from seaborn) (3.3.4)  ing pandas>=0.23  oading pandas>1.2.2-cp38-cp38-win_amd64.whl (9.3 MB)  ment already satisfied: numpy>=1.15 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from seaborn) (1.19.2)  ment already satisfied: scipy>=1.0 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from seaborn) (1.6.0)  ment already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0, 3 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from matplotlib>=2.2->seaborn) (2.4.7)  ment already satisfied: python-dateutil>=2.1 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from matplotlib>=2.2->seaborn) (2.8.1)
Require Require Collect Down Install Success  import plt.fi sn.hea plt.xl plt.yl	ment already satisfied: kiwisolver>=1.0.1 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from matplotlib=2.2->seaborn) (1.3.1) ment already satisfied: pillow>=6.2.0 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from matplotlib>=2.2->seaborn) (8.1.0) ment already satisfied: cycler>=0.10 in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from matplotlib>=2.2->seaborn) (0.10.0) ment already satisfied: six in d:\programdata\anaconda3\envs\tensor\lib\site-packages (from cycler>=0.10->matplotlib>=2.2->seaborn) (1.15.0) ing pytz>=2017.3 oading pytz-2021.1-py2.py3-none-any.whl (510 kB) ing collected packages: pytz, pandas, seaborn fully installed pandas-1.2.2 pytz-2021.1 seaborn-0.11.1  seaborn as sn gure(figsize = (10,7)) tmap(cm, annot=True, fmt='d') abel('Predicted') abel('Predicted') abel('Truth')  .0, 0.5, 'Truth')
	57 0 2 2 2 0 5 7 3 3 1 -1000  10 1106 4 2 0 1 4 2 16 0  4 7 930 12 8 4 11 9 42 5  1 0 21 924 1 19 2 8 19 15  1 1 6 0 912 0 9 3 9 41 -600  3 2 4 32 11 776 13 3 34 9  5 3 8 1 8 14 914 2 2 0 -400
Using h  model ke	4 5 7 14 10 20 9 9 879 17 7 6 2 10 19 4 0 8 8 945 0 1 2 3 4 5 6 7 8 9  dden layer  = keras. Sequential([ ras.layers. Dense(100, input_shape=(784,), activation='relu'), ras.layers. Dense(10, activation='sigmoid')  compile(optimizer='adam',
model :  Epoch : 1875/18 Epoch : 1875/18 Epoch : 1875/18 Epoch : 1875/18	loss='sparse_categorical_crossentropy', metrics=['accuracy'])  fit(X_train_flattened, y_train, epochs=5)  /5  75 [====================================
313/313 : [0.0900 : y_pred y_pred cm = t plt.fi sn.hea plt.xl plt.yl	<pre>evaluate(X_test_flattened,y_test)  [===================================</pre>
Futh 6 5 4 3 2 1 0	1127
Using  model  ke	1
model :  model :  model :  Epoch : 1875/18	Compile(optimizer='adam',
Epoch ( 1875/18 Epoch ( 1875/18 Epoch ( 1875/18 Epoch ( 1875/18	/10 75 [====================================