In [81]:

```
from tensorflow.keras import datasets
from tensorflow.keras.layers import Dense
from tensorflow.keras.models import Sequential
import matplotlib.pyplot as plt
import numpy as np
%matplotlib inline
```

In [82]:

```
(X_train, Y_train), (X_test, Y_test) = datasets.mnist.load_data()
```

In [83]:

```
len(X_train), X_train.shape
```

Out[83]:

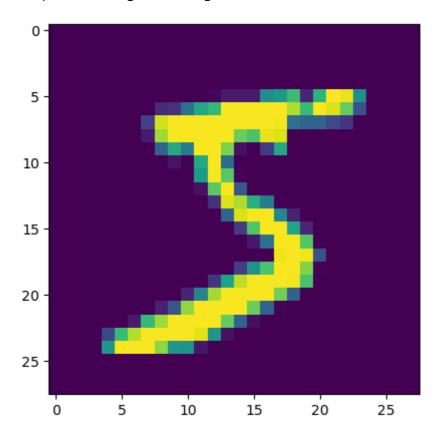
```
(60000, (60000, 28, 28))
```

In [84]:

```
plt.imshow(X_train[0])
```

Out[84]:

<matplotlib.image.AxesImage at 0x1f9c9ba88b0>



```
In [85]:
Y_train[0]
Out[85]:
5
In [86]:
X_train = X_train/255
X_{\text{test}} = X_{\text{test}/255}
In [87]:
X_train_flatten = X_train.reshape(len(X_train), 28 * 28)
X_test_flatten = X_test.reshape(len(X_test), 28 * 28)
In [88]:
X_train_flatten.shape, X_test_flatten.shape
Out[88]:
((60000, 784), (10000, 784))
In [89]:
model1 = Sequential([
    Dense(784, activation='relu', input_shape = (784, )),
    Dense(10, activation='sigmoid')
])
In [90]:
model1.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy
```

```
In [91]:
```

```
model1.fit(X_train_flatten, Y_train, epochs=10, validation_data=(X_test_flatten, Y_test))
Epoch 1/10
accuracy: 0.9447 - val_loss: 0.0935 - val_accuracy: 0.9700
Epoch 2/10
accuracy: 0.9771 - val_loss: 0.0672 - val_accuracy: 0.9790
Epoch 3/10
accuracy: 0.9848 - val_loss: 0.0854 - val_accuracy: 0.9733
Epoch 4/10
accuracy: 0.9887 - val_loss: 0.0706 - val_accuracy: 0.9776
Epoch 5/10
accuracy: 0.9922 - val_loss: 0.0689 - val_accuracy: 0.9821
Epoch 6/10
accuracy: 0.9931 - val_loss: 0.0867 - val_accuracy: 0.9780
Epoch 7/10
accuracy: 0.9946 - val_loss: 0.0682 - val_accuracy: 0.9815
accuracy: 0.9948 - val_loss: 0.0996 - val_accuracy: 0.9768
Epoch 9/10
accuracy: 0.9956 - val_loss: 0.0986 - val_accuracy: 0.9776
Epoch 10/10
accuracy: 0.9962 - val_loss: 0.0967 - val_accuracy: 0.9778
Out[91]:
<keras.callbacks.History at 0x1f9c69070d0>
In [92]:
model1.evaluate(X_test_flatten, Y_test)
313/313 [=============== ] - 2s 5ms/step - loss: 0.0967 - accu
racy: 0.9778
Out[92]:
[0.09668231755495071, 0.9778000116348267]
In [93]:
Y_pred = model1.predict(X_test_flatten)
313/313 [=========== ] - 2s 5ms/step
```

In [94]:

```
Y_pred[0]
```

Out[94]:

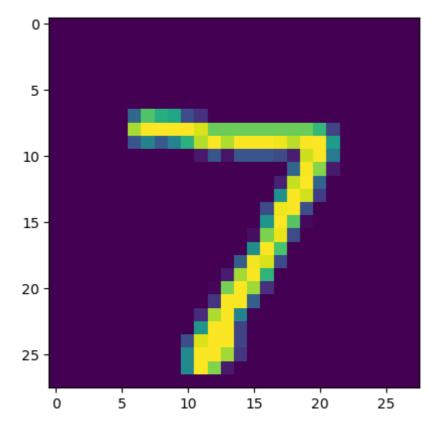
```
array([2.3075788e-07, 1.0286807e-06, 1.3131865e-06, 1.4375013e-01, 8.2776286e-10, 6.6616099e-06, 3.4993720e-13, 9.9999976e-01, 1.0598345e-06, 7.7417299e-02], dtype=float32)
```

In [95]:

```
plt.imshow(X_test[0])
```

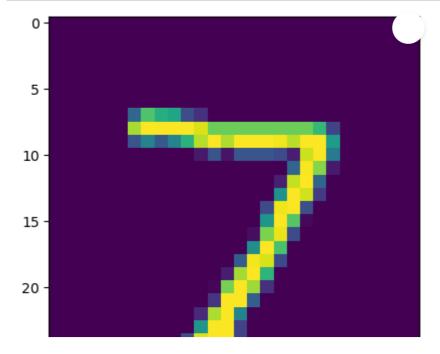
Out[95]:

<matplotlib.image.AxesImage at 0x1f9c8b1b3a0>



In [96]:

```
for i in range(12):
   plt.imshow(X_test[i])
   plt.show()
   print(np.argmax(Y_pred[i]))
```



In [97]:

```
Y_pred_int = [np.argmax(i) for i in Y_pred]
```

In [98]:

from sklearn.metrics import confusion_matrix, classification_report

In [99]:

```
mt = confusion_matrix(Y_test, Y_pred_int)
```

In [100]:

print(classification_report(Y_test, Y_pred_int))

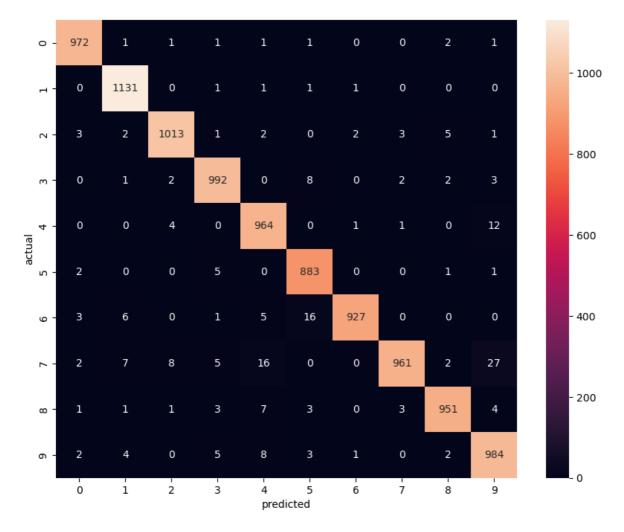
	precision	recall	f1-score	support
0	0.99	0.99	0.99	980
1	0.98	1.00	0.99	1135
2	0.98	0.98	0.98	1032
3	0.98	0.98	0.98	1010
4	0.96	0.98	0.97	982
5	0.97	0.99	0.98	892
6	0.99	0.97	0.98	958
7	0.99	0.93	0.96	1028
8	0.99	0.98	0.98	974
9	0.95	0.98	0.96	1009
accuracy			0.98	10000
macro avg	0.98	0.98	0.98	10000
weighted avg	0.98	0.98	0.98	10000

In [101]:

```
import seaborn as sns
plt.figure(figsize=(10, 8))
sns.heatmap(mt, annot=True, fmt='d')
plt.ylabel("actual")
plt.xlabel("predicted")
```

Out[101]:

Text(0.5, 58.72222222222, 'predicted')



In [102]:

```
import pandas as pd
data = pd.DataFrame({"Predictions": Y_pred_int, "True": Y_test})
```

In [104]:

```
np.savetxt("Predictions.csv", data, fmt="%d")
```