

## What is observing the metrics?

A metric is any value you can measure over time. It can be blocks used on a filesystem, the number of nodes in a cluster, or a temperature reading. Observe reports Metrics in the form of a time series: a set of values in time order.

## Evaluate the Handwritten Digit Recognition Model on Test Data

```
#evaluate the model  
test_loss, test_acc = model_1.evaluate(X_test, Y_test)  
print('Test accuracy:', test_acc)
```

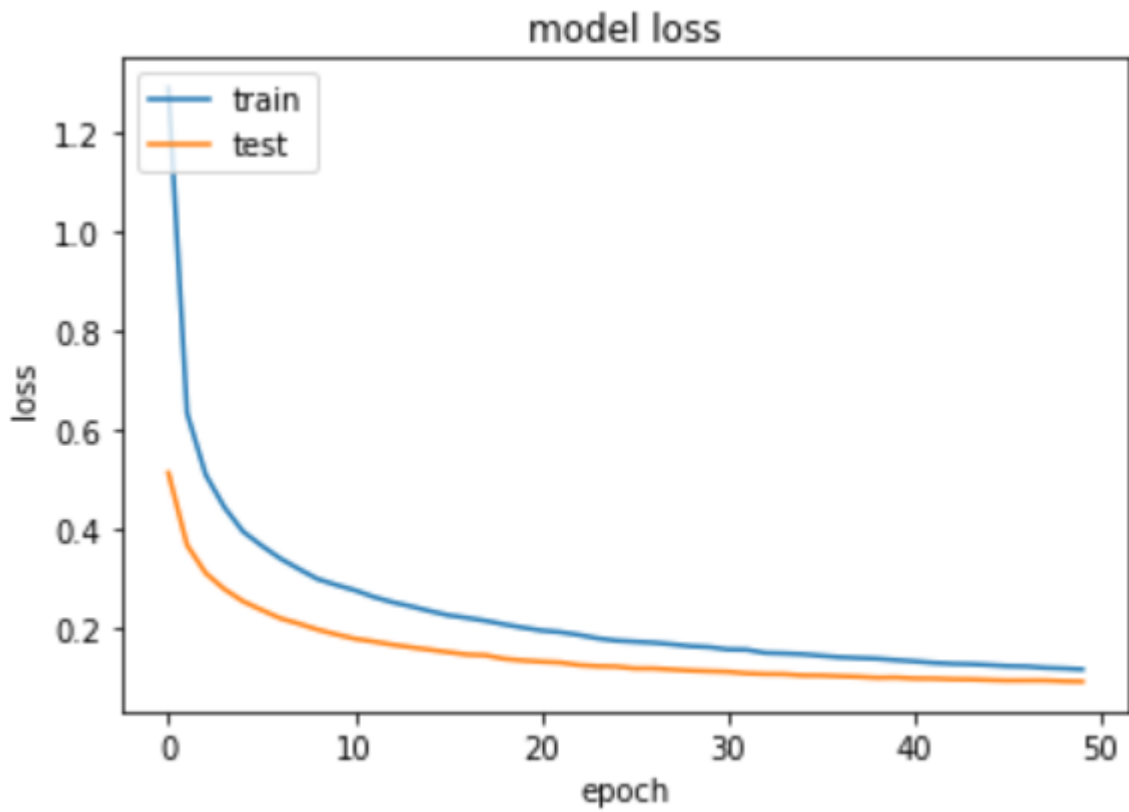
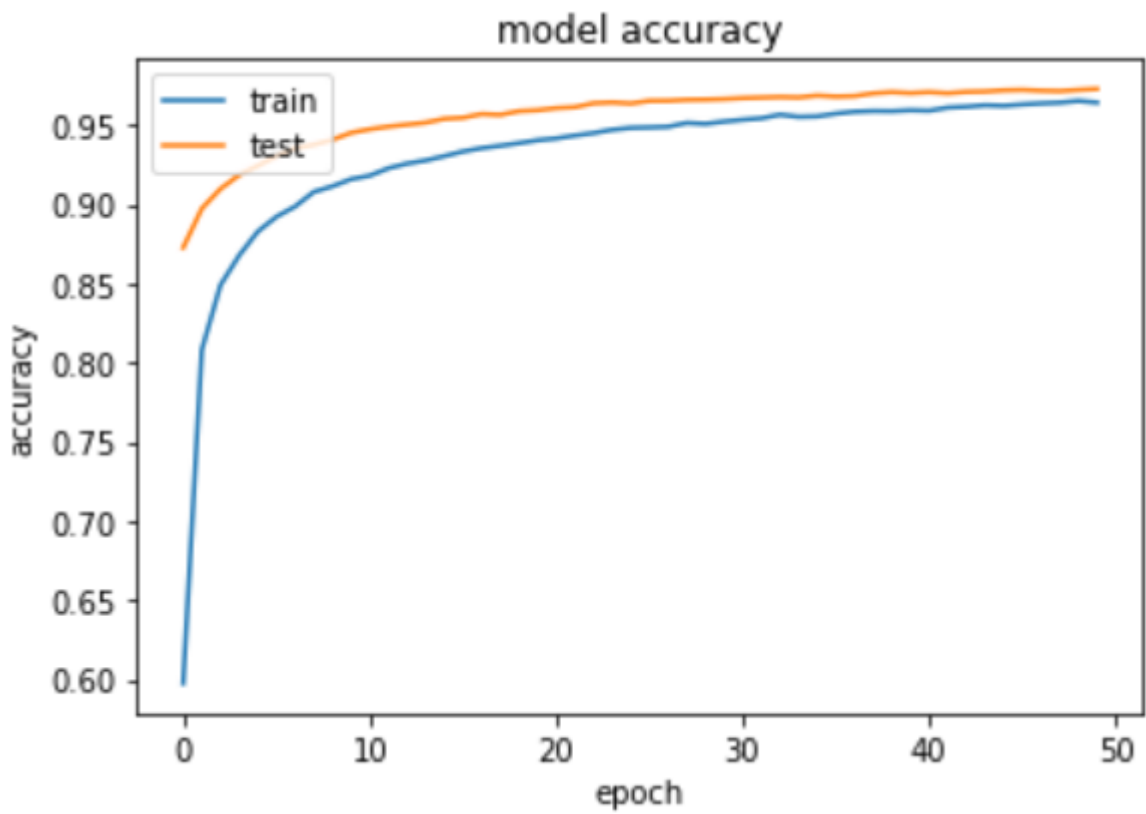
### Output:

313/313 [=====] – 0s 1ms/stop  
– loss: 0.0935 – Accuracy: 0.9737 Test Accuracy:  
0.9736999869346619The second model is giving an output of 97%.  
Further, we can improve the model by adding a dropout to avoid  
overfitting.

### Plot the change in metrics per epoch:

Model will be now trained on the on the training data. For this we will be defining the epochs, batchsize, and validation size

- epoch:** Number of times that the model will run through the training dataset
- batch\_size:** Number of training instances to be shown to the model before a weight is updated
- validation\_split:** Defines the fraction of data to be used for validation purpose



```
import matplotlib.pyplot as plt
%matplotlib inline
# list all data in training
print(training.history.keys())
# summarize training for accuracy
plt.plot(training.history['accuracy'])
plt.plot(training.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
# summarize training for loss
plt.plot(training.history['loss'])
plt.plot(training.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```