TensorFlow 2 quickstart for experts

Run in

<u>Google</u> (https://colab.research.google.com/github/tensorflow/docs/blob/master/site/en/tutorials/quic Colab

This is a <u>Google Colaboratory</u> (https://colab.research.google.com/notebooks/welcome.ipynb) notebook file. Python programs are run directly in the browser—a great way to learn and use TensorFlow. To follow this tutorial, run the notebook in Google Colab by clicking the button at the top of this page.

- 1. In Colab, connect to a Python runtime: At the top-right of the menu bar, select *CONNECT*.
- 2. Run all the notebook code cells: Select Runtime > Run all.

Download and install TensorFlow 2. Import TensorFlow into your program:

Jpgrade **pip** to install the TensorFlow 2 package. See the <u>install guide</u> (https://www.tensorflow.org/installs.

Import TensorFlow into your program:

```
t tensorflow as tf
tensorflow.keras.layers import Dense, Flatten, Conv2D
tensorflow.keras import Model
```

Load and prepare the MNIST dataset (http://yann.lecun.com/exdb/mnist/).

```
:= tf.keras.datasets.mnist

rain, y_train), (x_test, y_test) = mnist.load_data()

rin, x_testie uses to the from Soogke testiver as selvices and to analyze traffic.

I a channels dimension

rin = x_train[..., tf.newaxis]

rit = x_test[..., tf.newaxis]
```

Use <u>tf.data</u> (/api_docs/python/tf/data) to batch and shuffle the dataset:

```
i_ds = tf.data.Dataset.from_tensor_slices(
  x_train, y_train)).shuffle(10000).batch(32)

ds = tf.data.Dataset.from_tensor_slices((x_test, y_test)).batch(32)
```

Build the <u>tf.keras</u> (/api_docs/python/tf/keras) model using the Keras <u>model subclassing API</u> (https://www.tensorflow.org/guide/keras#model_subclassing):

```
MyModel(Model):
   __init__(self):
   uper(MyModel, self).__init__()
   elf.conv1 = Conv2D(32, 3, activation='relu')
   elf.flatten = Flatten()
   elf.d1 = Dense(128, activation='relu')
   elf.d2 = Dense(10)

call(self, x):
   = self.conv1(x)
   = self.flatten(x)
   = self.flatten(x)
   eturn self.d2(x)

ate an instance of the model
   = MyModel()
```

Choose an optimizer and loss function for training:

```
object = tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True)
sizer = tf.keras.optimizers.Adam()
```

Select metrics to measure the loss and the accuracy of the model. These metrics accumulate the values over epochs and then print the overall result.

This site uses cookies from Google to deliver its services and to analyze traffic.

```
loss = tf.keras.metrics.Mean(name='test_loss')
_accuracy = tf.keras.metrics.SparseCategoricalAccuracy(name='test_accuracy')
Use <u>tf.GradientTape</u> (/api_docs/python/tf/GradientTape) to train the model:
unction
rain_step(images, labels):
:h tf.GradientTape() as tape:
training=True is only needed if there are layers with different
behavior during training versus inference (e.g. Dropout).
redictions = model(images, training=True)
.oss = loss_object(labels, predictions)
idients = tape.gradient(loss, model.trainable_variables)
:imizer.apply_gradients(zip(gradients, model.trainable_variables))
in_loss(loss)
in_accuracy(labels, predictions)
Test the model:
unction
:est_step(images, labels):
raining=False is only needed if there are layers with different
ehavior during training versus inference (e.g. Dropout).
dictions = model(images, training=False)
.oss = loss_object(labels, predictions)
t_loss(t_loss)
it_accuracy(labels, predictions)
IS = 5
poch in range(EPOCHS):
leset the metrics at the start of the next epoch
in_loss.reset_states()
in_accuracy.reset_states()
st_loss.reset_states()
t_accumas yite eset estate shom Google to deliver its services and to analyze traffic.
images, labels in train_ds:
                                                   More details
                                                                 OK
:rain_step(images, labels)
```

NG:tensorflow:Layer my_model is casting an input tensor from dtype float64 to u intended to run this layer in float32, you can safely ignore this warning.

lange all layers to have dtype float64 by default, call `tf.keras.backend.set_

1, Loss: 0.13576626777648926, Accuracy: 95.89666748046875, Test Loss: 0.0658

2, Loss: 0.04346510395407677, Accuracy: 98.66667175292969, Test Loss: 0.0584

3, Loss: 0.024708667770028114, Accuracy: 99.1883316040039, Test Loss: 0.0513

4, Loss: 0.014630611054599285, Accuracy: 99.49500274658203, Test Loss: 0.056

5, Loss: 0.010477297939360142, Accuracy: 99.61333465576172, Test Loss: 0.064

The image classifier is now trained to ~98% accuracy on this dataset. To learn more, read the <u>TensorFlow tutorials</u> (https://www.tensorflow.org/tutorials).

Except as otherwise noted, the content of this page is licensed under the <u>Creative Commons Attribution 4.0</u>
<u>License</u> (https://creativecommons.org/licenses/by/4.0/), and code samples are licensed under the <u>Apache 2.0 License</u> (https://www.apache.org/licenses/LICENSE-2.0). For details, see the <u>Google Developers Site Policies</u> (https://developers.google.com/site-policies). Java is a registered trademark of Oracle and/or its affiliates.

Last updated 2020-06-12.

This site uses cookies from Google to deliver its services and to analyze traffic.

More details OK