AI Tutorial

IMP Links

* <https://colab.research.google.com/github/tensorflow/docs/blob/master/site/en/tutorials/quickstart/beginner.ipynb#scrollTo=he5u_okAYS4a>
* <https://www.tensorflow.org/lite/guide/hosted_models> (All Tensor Flow Models)
* <https://www.tensorflow.org/lite/examples> (Examples of Tensorflow Models)
* <https://www.tensorflow.org/lite/guide> (Starting guide for Tensorflow Model Development)
* https://www.tensorflow.org/lite/guide/android (Guide for Tensorflow Model Development on Android)
* <https://www.tensorflow.org/lite>
* [Intro to TensorFlow for Deep Learning - Udacity](https://classroom.udacity.com/courses/ud187/lessons/7b590cdb-0acf-4118-848c-8728ced19bc6/concepts/4e1b5f7e-5094-481e-ae29-284bc2fb847f) (Crash Course for beginners to learn Deep Learning) – added on 04-05-2022
* TensorFlow Lite is a set of tools that enables on-device machine learning by helping developers run their models on mobile, embedded, and edge devices.

**Development workflow**

### **1. Generate a TensorFlow Lite model**

A TensorFlow Lite model is represented in a special efficient portable format known as [FlatBuffers](https://google.github.io/flatbuffers/). This provides several advantages over TensorFlow's protocol buffer model format such as reduced size (small code footprint) and faster inference (data is directly accessed without an extra parsing/unpacking step) that enables TensorFlow Lite to execute efficiently on devices with limited compute and memory resources.

You can generate a TensorFlow Lite model in the following ways:

* **Use an existing TensorFlow Lite model:** Refer to [TensorFlow Lite Examples](https://www.tensorflow.org/lite/examples) to pick an existing model. Models may or may not contain metadata.
* **Create a TensorFlow Lite model:** Use the [TensorFlow Lite Model Maker](https://www.tensorflow.org/lite/guide/model_maker) to create a model with your own custom dataset. By default, all models contain metadata.
* **Convert a TensorFlow model into a TensorFlow Lite model:** Use the [TensorFlow Lite Converter](https://www.tensorflow.org/lite/convert/index) to convert a TensorFlow model into a TensorFlow Lite model. During conversion, you can apply [optimizations](https://www.tensorflow.org/lite/performance/model_optimization) such as [quantization](https://www.tensorflow.org/lite/performance/post_training_quantization) to reduce model size and latency with minimal or no loss in accuracy. By default, all models don't contain metadata.

**Simple ML Code to create a model and to predict Temperature from Celsius to Fahrenheit**

Link for the code : https://colab.research.google.com/drive/1MUg0qYcnmmGV6Iw3926WTSjL9HTDXVe6#scrollTo=mp7NnuK5jzAO

import tensorflow as tf

import numpy as np

#input data which is in Celsius format

x = np.array([0, 10, 20, 30, 38, 32], dtype=float)

y = np.array([32, 50, 68, 86, 100, 90], dtype=float)

#Creating Layers to process the input data and to predict the output

l1 = tf.keras.layers.Dense(units=4, name='l1', input\_shape=[1])

l2 = tf.keras.layers.Dense(units=4, name='l2')

l3 = tf.keras.layers.Dense(units=1, name='l3')

#Creating Model

model = tf.keras.Sequential([l1, l2, l3])

#Compile the model

model.compile(

    loss='mean\_squared\_error',

    optimizer=tf.keras.optimizers.Adam(0.1),

)

#Train the model

history = model.fit(x, y, epochs=500,verbose=False)

#Visualize the loss rate

import matplotlib.pyplot as plt

plt.xlabel("Epochs")

plt.ylabel("Loss")

plt.plot(history.history['loss'])

#predict the model

print(model.predict([100.0]))

Table

Description automatically generated