

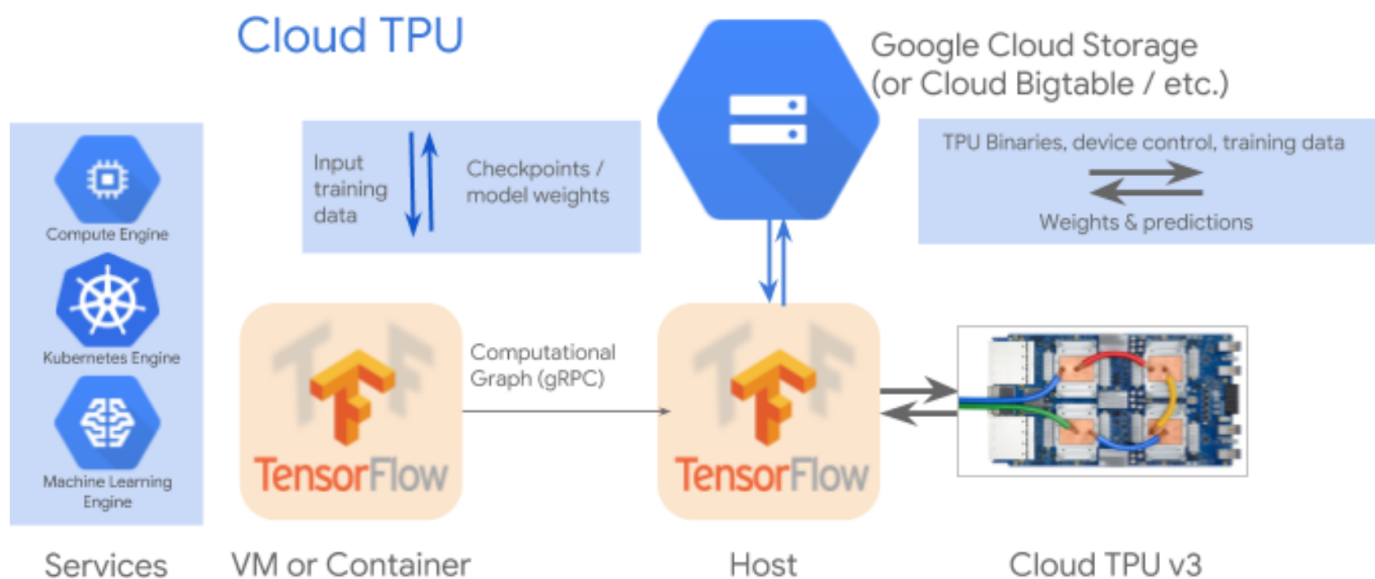
▼ 1. Configuration

1.1 Define some constants

```
GCS_PATH = 'gs://tf-everywhere-col'
DATASET_NAME = 'rock_paper_scissors'
BATCH_SIZE = 128
NUM_CLASSES = 3
DENSE_UNITS = 128
DROPOUT = 0.4
```

▼ 1.2 Authenticate

Authenticates the Colab machine and the TPU using your credentials so you can access the private GCS bucket.



```
from google.colab import auth
```

```
auth.authenticate_user()
```

▼ 1.3 Setup TPU

```
!pip install tensorflow-addons
```

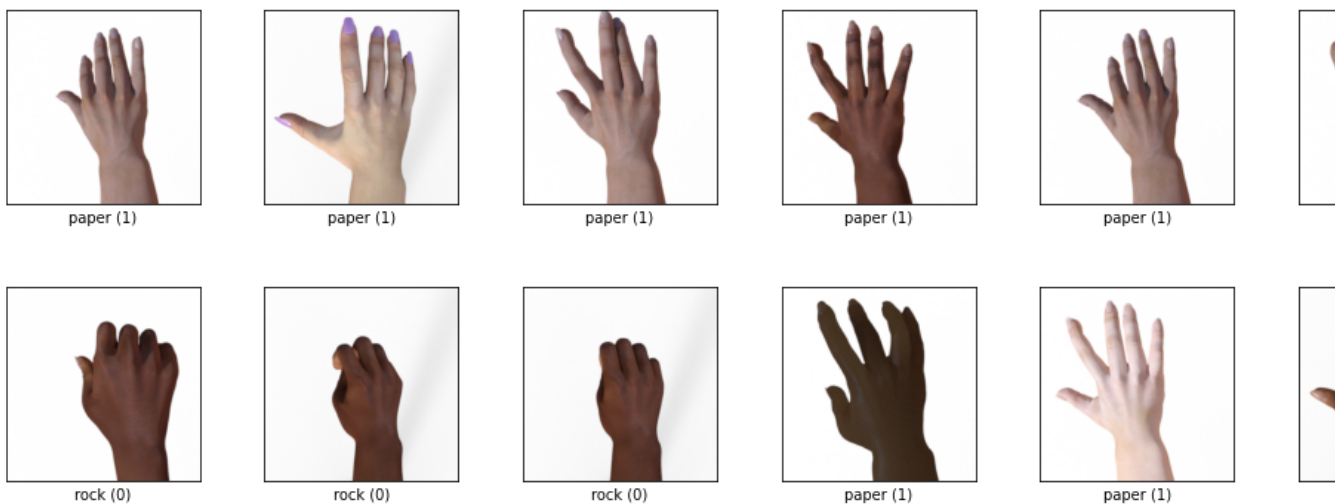


```
INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
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INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
INFO:tensorflow:*** Available Device: _DeviceAttributes(/job:worker/replica:0/task:0/dev
```

▼ 2. Dataset definition

```
train_ds, train_ds_info = tfds.load(DATASET_NAME,
                                    split="train",
                                    with_info=True,
                                    shuffle_files=True,
                                    data_dir=f'{GCS_PATH}/{DATASET_NAME}/train')
val_ds, val_ds_info = tfds.load(DATASET_NAME,
                                 split="test",
                                 with_info=True,
                                 shuffle_files=False,
                                 data_dir=f'{GCS_PATH}/{DATASET_NAME}/val')
```

```
tfds.visualization.show_examples(train_ds, train_ds_info, rows=2, cols=6);
```



▼ 2.1 Data Preprocessing

```
from tensorflow.data.experimental import AUTOTUNE
```

```
from tensorflow.data.experimental import AUTOTUNE
```

```
@tf.function
def preprocess(example, height=224, width=224):
    image = example['image']
    label = example['label']
    image = tf.image.resize_with_crop_or_pad(image, height, width)

    image = tf.keras.applications.resnet.preprocess_input(image)
    label = tf.one_hot(label, depth=NUM_CLASSES)
    return image, label

def get_datasets(train_ds, val_ds, batch_size):
    norm_train_ds = train_ds.map(preprocess, num_parallel_calls=AUTOTUNE)\
        .shuffle(500).batch(batch_size).prefetch(AUTOTUNE)
    norm_val_ds = val_ds.map(preprocess, num_parallel_calls=AUTOTUNE)\
        .batch(batch_size).cache()
    return norm_train_ds, norm_val_ds
```

▼ 2.3 Create train/val datasets

Note the number of replicas is **8** then we need to multiply the original batch size with the number of replicas, to optimize the usage of the TPU.

```
norm_train_ds, norm_val_ds = get_datasets(train_ds, val_ds,
                                          batch_size=strategy.num_replicas_in_sync*BATCH_SIZE)
```

```
strategy.num_replicas_in_sync
```

```
8
```

▼ 3. Model definition

```
def create_model(num_classes, units, dropout):

    input_layer = layers.Input(shape=(224,224,3))

    feature_extractor = tf.keras.applications.ResNet50(include_top=False, weights='imagenet',
    feature_extractor.trainable = False

    features = layers.GlobalAveragePooling2D()(feature_extractor.output)
    x = layers.Dense(units, 'relu')(features)
    x = layers.Dropout(dropout)(x)
    x = layers.Dense(num_classes, 'softmax', dtype='float32')(x)

    model = Model(inputs=[input_layer], outputs=[x], name='feature_extractor')
    model.summary()
```

```
model.summary()
return model
```

▼ 3.1 Create & Compile

Using the TPU distribution strategy create and compile the model

```
with strategy.scope():
    model = create_model(NUM_CLASSES, DENSE_UNITS, DROPOUT)
    model.compile(optimizer='adam',
                  loss=tf.keras.losses.CategoricalCrossentropy(label_smoothing=0.2),
                  metrics=[tf.keras.metrics.CategoricalAccuracy(),
                           tfa.metrics.F1Score(num_classes=NUM_CLASSES, average='macro')])
```

Downloading data from <https://storage.googleapis.com/tensorflow/keras-applications/r94773248/94765736> [=====] - 1s 0us/step

Model: "feature_extractor"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3)	0	input_1[0][0]
conv1_conv (Conv2D)	(None, 112, 112, 64)	9472	conv1_pad[0][0]
conv1_bn (BatchNormalization)	(None, 112, 112, 64)	256	conv1_conv[0][0]
conv1_relu (Activation)	(None, 112, 112, 64)	0	conv1_bn[0][0]
pool1_pad (ZeroPadding2D)	(None, 114, 114, 64)	0	conv1_relu[0][0]
pool1_pool (MaxPooling2D)	(None, 56, 56, 64)	0	pool1_pad[0][0]
conv2_block1_1_conv (Conv2D)	(None, 56, 56, 64)	4160	pool1_pool[0][0]
conv2_block1_1_bn (BatchNormalization)	(None, 56, 56, 64)	256	conv2_block1_1_conv
conv2_block1_1_relu (Activation)	(None, 56, 56, 64)	0	conv2_block1_1_bn[0]
conv2_block1_2_conv (Conv2D)	(None, 56, 56, 64)	36928	conv2_block1_1_relu
conv2_block1_2_bn (BatchNormalization)	(None, 56, 56, 64)	256	conv2_block1_2_conv
conv2_block1_2_relu (Activation)	(None, 56, 56, 64)	0	conv2_block1_2_bn[0]
conv2_block1_0_conv (Conv2D)	(None, 56, 56, 256)	16640	pool1_pool[0][0]
conv2_block1_3_conv (Conv2D)	(None, 56, 56, 256)	16640	conv2_block1_2_relu
conv2_block1_0_bn (BatchNormalization)	(None, 56, 56, 256)	1024	conv2_block1_0_conv
conv2_block1_3_bn (BatchNormalization)	(None, 56, 56, 256)	1024	conv2_block1_3_conv

conv2_block1_add (Add)	(None, 56, 56, 256)	0	conv2_block1_0_bn[0] conv2_block1_3_bn[0]
conv2_block1_out (Activation)	(None, 56, 56, 256)	0	conv2_block1_add[0]
conv2_block2_1_conv (Conv2D)	(None, 56, 56, 64)	16448	conv2_block1_out[0]
conv2_block2_1_bn (BatchNormali	(None, 56, 56, 64)	256	conv2_block2_1_conv
conv2_block2_1_relu (Activation	(None, 56, 56, 64)	0	conv2_block2_1_bn[0]
conv2_block2_2_conv (Conv2D)	(None, 56, 56, 64)	36928	conv2_block2_1_relu
conv2_block2_2_bn (BatchNormali	(None, 56, 56, 64)	256	conv2_block2_2_conv
conv2_block2_2_relu (Activation	(None, 56, 56, 64)	0	conv2_block2_2_bn[0]
conv2_block2_3_conv (Conv2D)	(None, 56, 56, 256)	16640	conv2_block2_2_relu

▼ 3.2 Train & Validate model

```
%%time
```

```
history = model.fit(norm_train_ds, validation_data=norm_val_ds, epochs=20)
```

```
Epoch 1/20
```

```
3/3 [=====] - 25s 6s/step - loss: 2.0653 - categorical_accuracy
```

```
Epoch 2/20
```

```
3/3 [=====] - 2s 705ms/step - loss: 1.0916 - categorical_accuracy
```

```
Epoch 3/20
```

```
3/3 [=====] - 2s 693ms/step - loss: 0.8546 - categorical_accuracy
```

```
Epoch 4/20
```

```
3/3 [=====] - 2s 670ms/step - loss: 0.7435 - categorical_accuracy
```

```
Epoch 5/20
```

```
3/3 [=====] - 2s 683ms/step - loss: 0.6876 - categorical_accuracy
```

```
Epoch 6/20
```

```
3/3 [=====] - 2s 653ms/step - loss: 0.6498 - categorical_accuracy
```

```
Epoch 7/20
```

```
3/3 [=====] - 2s 849ms/step - loss: 0.6254 - categorical_accuracy
```

```
Epoch 8/20
```

```
3/3 [=====] - 2s 662ms/step - loss: 0.6039 - categorical_accuracy
```

```
Epoch 9/20
```

```
3/3 [=====] - 2s 708ms/step - loss: 0.5957 - categorical_accuracy
```

```
Epoch 10/20
```

```
3/3 [=====] - 2s 676ms/step - loss: 0.5924 - categorical_accuracy
```

```
Epoch 11/20
```

```
3/3 [=====] - 2s 684ms/step - loss: 0.5794 - categorical_accuracy
```

```
Epoch 12/20
```

```
3/3 [=====] - 2s 672ms/step - loss: 0.5722 - categorical_accuracy
```

```
Epoch 13/20
```

```
3/3 [=====] - 2s 847ms/step - loss: 0.5663 - categorical_accuracy
```

```
Epoch 14/20
```

```
3/3 [=====] - 2s 666ms/step - loss: 0.5603 - categorical_accuracy
```

```
Epoch 15/20
```

```
3/3 [=====] - 2s 676ms/step - loss: 0.5569 - categorical_accu
Epoch 16/20
3/3 [=====] - 2s 649ms/step - loss: 0.5505 - categorical_accu
Epoch 17/20
3/3 [=====] - 2s 650ms/step - loss: 0.5502 - categorical_accu
Epoch 18/20
3/3 [=====] - 2s 649ms/step - loss: 0.5441 - categorical_accu
Epoch 19/20
3/3 [=====] - 2s 827ms/step - loss: 0.5446 - categorical_accu
Epoch 20/20
3/3 [=====] - 2s 653ms/step - loss: 0.5411 - categorical_accu
CPU times: user 35.7 s, sys: 3.01 s, total: 38.7 s
Wall time: 1min 9s
```