

# TensorFlow Developer Certificate Notes

## Contents

## Introduction:

- `tf.constant()` is not mutable, but `tf.Variable()` is by using the `.assign()` method on the var object.
- You must set both the global `tf.random.set_seed()` and function `seed=` parameter to get reproducible results for shuffle function.
- We can *add dimensions* to a tensor whilst keeping the same information (*newaxis* and *expand\_dims* have same output).

```
1 rank_3_tensor = rank_2_tensor[..., tf.newaxis] # "..." means "all dims prior to"
2 rank_2_tensor, rank_3_tensor # shape (2, 2), shape (2, 2, 1)
3 tf.expand_dims(rank_2_tensor, axis=-1) # "-1" means last axis (2, 2, 1)
```

- `tf.reshape()` will change the shape in the order they appear (top left to bottom right) and `tf.transpose()` simply flips the matrix.
- We can reduce tensor sizes in memory by changing the datatype (i.e. float32 cast to float16).
- We can perform aggregation on tensors by using `reduce()[action]` and using min, max, mean, sum, etc. We can also find positional arguments using `tf.argmin()` or `tf.argmax()`.

## Neural Network Classification:

- We can create a **learning rate callback** to update our learning rate during training.

```
1 # Create a learning rate scheduler callback
2 lr_scheduler = tf.keras.callbacks.LearningRateScheduler(lambda epoch:
3 1e-4 * 10**(epoch/20))
```

- Traverse a set of learning rate values starting from 1e-4, increasing by 10\*\*(epoch/20) every epoch.
- Note that learning rate exponentially increases as epochs increases.
- We can use a plot to determine the **ideal learning rate**, which we want to take the value where loss is still decreasing but not quite flattened out. It is the value around 10x smaller than the lowest point (refer to notebook for graph and point selection).

```
1 lrs = 1e-4 * (10 ** (np.arange(100)/20))
2 plt.figure(figsize=(10, 7))
3 plt.semilogx(lrs, history.history["loss"]) # x-axis (lr) to be log scale
```

## Computer Vision & CNNs:

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