

# Tensors Introduction

May 22, 2021

## 1 Introduction to Tensor Notes :

A tensor is a container which can house data in N dimensions, along with its linear operations, though there is nuance in what tensors technically are and what we refer to as tensors in practice.

### Scalar Vector Matrix Tensor

1

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$

$\begin{bmatrix} \begin{bmatrix} 1 & 2 \end{bmatrix} & \begin{bmatrix} 3 & 2 \end{bmatrix} \\ \begin{bmatrix} 1 & 7 \end{bmatrix} & \begin{bmatrix} 5 & 4 \end{bmatrix} \end{bmatrix}$

```
[4]: #Import Tensorflow
import tensorflow as tf
print(tf.__version__)
import numpy
```

2.3.0

```
[3]: #Create tensors with tf.constant()
scalar = tf.constant(7)
scalar
```

```
[3]: <tf.Tensor: shape=(), dtype=int32, numpy=7>
```

```
[6]: #Check the number of dimensions of tensor using ndim
scalar.ndim
```

```
[6]: 0
```

```
[7]: #Create a vector
vector = tf.constant([10,10])
vector
```

```
[7]: <tf.Tensor: shape=(2,), dtype=int32, numpy=array([10, 10])>
```

```
[8]: vector.ndim
```

```
[8]: 1
```

```
[10]: #Create a matrix  
matrix = tf.constant([[10,7],  
                      [7,10]]  
                      )  
  
matrix
```

```
[10]: <tf.Tensor: shape=(2, 2), dtype=int32, numpy=  
array([[10,  7],  
       [ 7, 10]])>
```

```
[11]: matrix.ndim
```

```
[11]: 2
```

At this point, we can relate that ndim represents the number of elements in the shape tuple

```
[13]: #Create another matrix  
matrix_2 = tf.constant([[10.,7.],  
                       [3.,2.],  
                       [1.,2.]],dtype=tf.float16) #Here we use float16 since  
↪our numbers are small  
  
matrix_2
```

```
[13]: <tf.Tensor: shape=(3, 2), dtype=float16, numpy=  
array([[10.,  7.],  
       [ 3.,  2.],  
       [ 1.,  2.]], dtype=float16)>
```

```
[14]: matrix_2.ndim
```

```
[14]: 2
```

```
[18]: #Create a tensor  
tensor = tf.constant([[[1,2,3],  
                      [4,5,6]],  
                      [[7,8,9],  
                      [10,11,12]],  
                      [[13,14,15],  
                      [16,17,18]]])  
  
tensor
```

```
[18]: <tf.Tensor: shape=(3, 2, 3), dtype=int32, numpy=  
array([[[ 1,  2,  3],  
        [ 4,  5,  6]],
```

```

[[ 7,  8,  9],
 [10, 11, 12]],

[[13, 14, 15],
 [16, 17, 18]]])>

```

```

[19]: rank_3_tensor = tf.constant([
    [[0, 1, 2, 3, 4],
     [5, 6, 7, 8, 9]],
    [[10, 11, 12, 13, 14],
     [15, 16, 17, 18, 19]],
    [[20, 21, 22, 23, 24],
     [25, 26, 27, 28, 29]],])

print(rank_3_tensor)

```

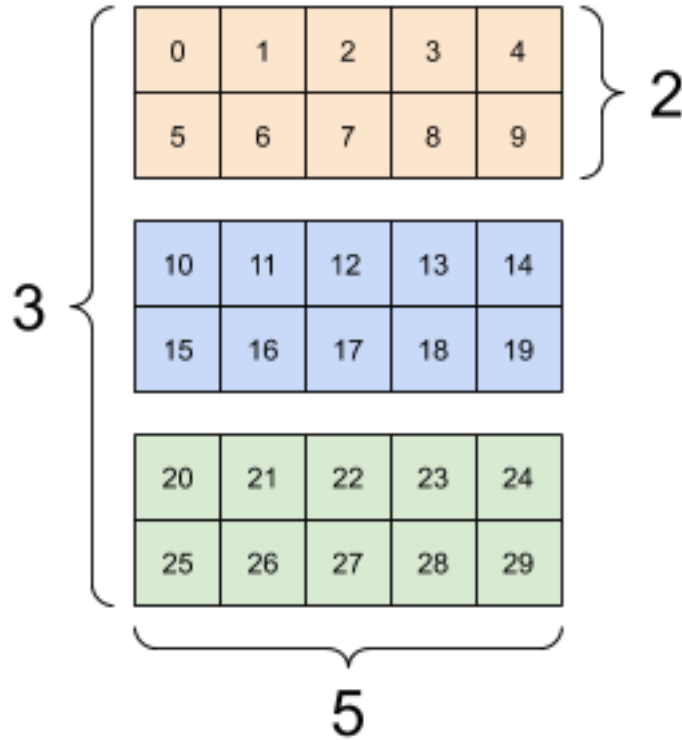
```

tf.Tensor(
[[[ 0  1  2  3  4]
  [ 5  6  7  8  9]]

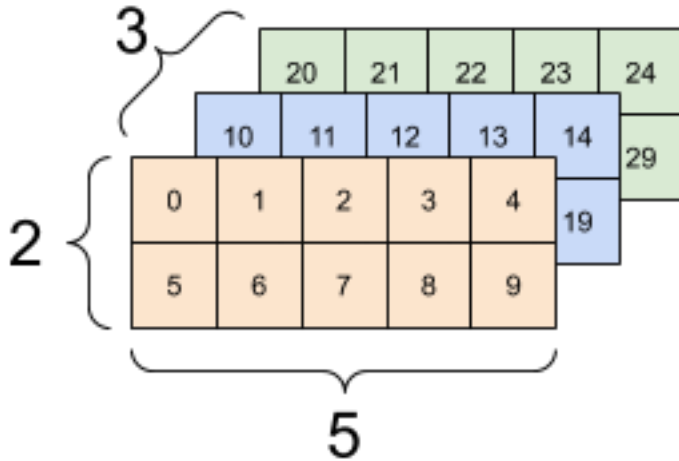
[[10 11 12 13 14]
 [15 16 17 18 19]]

[[20 21 22 23 24]
 [25 26 27 28 29]]], shape=(3, 2, 5), dtype=int32)

```



The above example for a 3 dimensional tensor represents : Number of matrices , Number of rows in a matrices and Number of columns in a matrices i.e. Shape = (3,2,5). You can also visualise this as matrices stacked on top of each other to produce a 3D structure as shown below



## 2 Summary so far :

1. Scalar : Single number
2. Vector : A number with both direction and magnitude
3. Matrix : A 2 dimensional array of numbers
4. Tensor : A n-dimensional array of numbers which can constitute all of the above as well.