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Link to the notebook:

[https://github.com/vikpy/AISem3/blob/master/HW/HomeWork2\\_Tensor\\_Vector\\_Binary\\_Tree.ipynb](https://github.com/vikpy/AISem3/blob/master/HW/HomeWork2_Tensor_Vector_Binary_Tree.ipynb)

# Operations:



```
[26] # In tensorflow

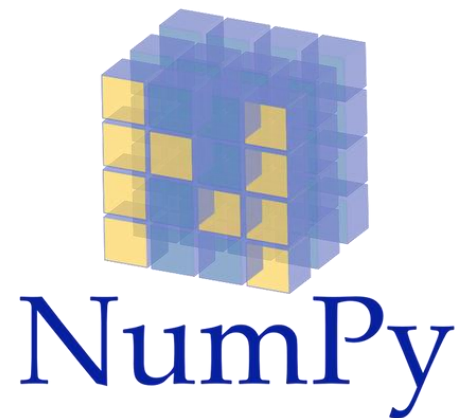
a_tf = tf.constant([[1, 2],
                    [3, 4]])
b_tf = tf.constant([[2, -1],
                    [1, 7]]) # Could have also said `tf.ones([2,2])`

print(tf.add(a_tf, b_tf), "\n")
print(tf.multiply(a_tf, b_tf), "\n")
print(tf.matmul(a_tf, b_tf), "\n")

↳ tf.Tensor(
[[ 3  1]
 [ 4 11]], shape=(2, 2), dtype=int32)

tf.Tensor(
[[ 2 -2]
 [ 3 28]], shape=(2, 2), dtype=int32)

tf.Tensor(
[[ 4 13]
 [10 25]], shape=(2, 2), dtype=int32)
```



```
▶ from numpy import np
a_np = np.array([[1, 2],
                 [3, 4]])
b_np = np.array([
    [[2, -1],
     [1, 7]] |
])

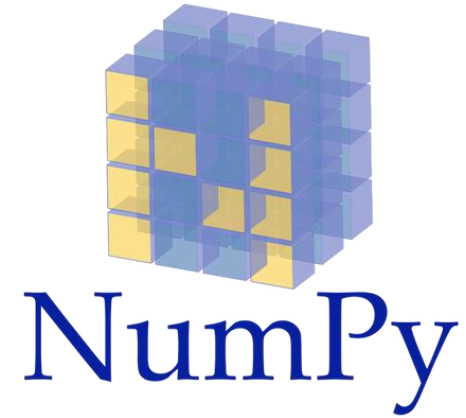
# Tensor operations similar to tf still works for numpy
print(a_np + b_np, type(a_np + b_np), "\n") # element-wise addition
print(a_np * b_np, type(a_np + b_np), "\n") # element-wise multiplication
print(a_np @ b_np, type(a_np + b_np), "\n") # matrix multiplication

↳ [[ 3  1]
   [ 4 11]] <class 'numpy.ndarray'>

[[ 2 -2]
 [ 3 28]] <class 'numpy.ndarray'>

[[ 4 13]
 [10 25]] <class 'numpy.ndarray'>
```

# Tensor division operation:



```
[29] print(a_np/b_np,type(a_np/b_np), "\n") # In numpy
      print(a_tf/b_tf,type(a_tf/b_tf), "\n") # In tensorflow
```

```
↳ [[ 0.5      -2.       ]
    [ 3.       0.57142857]] <class 'numpy.ndarray'>
```

```
tf.Tensor(
[[ 0.5      -2.       ]
 [ 3.       0.57142857]], shape=(2, 2), dtype=float64) <class 'tensorflow.python.framework.ops.EagerTensor'>
```

# Tensor dot Operation:



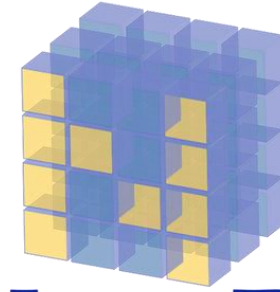
```
[36] #tensor dot is similar in tensorflow as well
c_tf = tf.tensordot(a_tf, b_tf, axes=0)
print(c_tf, type(c_tf))

tf.Tensor(
[[[ [ 2 -1]
      [ 1  7]]

  [[ 4 -2]
      [ 2 14]]

  [[ 6 -3]
      [ 3 21]]

  [[ 8 -4]
      [ 4 28]]]], shape=(2, 2, 2, 2), dtype=int32) <class 'tf.Tensor'>
```



## NumPy

```
#Tensor Dot Product using np array
c_np = np.tensordot(a_np, b_np, axes=0)
print(c_np, type(c_np))
```

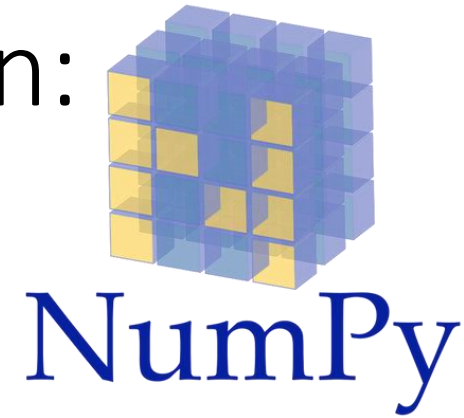
```
[[[ [ 2 -1]
      [ 1  7]]

  [[ 4 -2]
      [ 2 14]]

  [[ 6 -3]
      [ 3 21]]

  [[ 8 -4]
      [ 4 28]]]] <class 'numpy.ndarray'>
```

# Speed comparison Operation:



```
[57] import tensorflow as tf

A = tf.constant(A)
B = tf.constant(B)

timer = timeit.Timer("tf.matmul(A, B)", setup="import tensorflow as tf; from __main__ import A, B")
tensorflow_times_list = timer.repeat()
min(tensorflow_times_list)
```

16.763378783000007

```
[44] import numpy
import timeit
A = numpy.random.rand(10, 10).astype(numpy.float32)
B = numpy.random.rand(10, 10).astype(numpy.float32)

timer = timeit.Timer("numpy.dot(A, B)", "import numpy; from __main__ import A, B")
numpy_times_list = timer.repeat()
min(numpy_times_list)
```

1.456392951999078

# Thank You!

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