

✓ Matrix multiplication in Python vs. NumPy

```
# In Python, matrices can be implemented as lists of lists.
A = [[1, -1, 2],
      [0, 3, 1]]
B = [[1, 2],
      [3, 0],
      [2, 1]]

# Python function for multiplying two matrices.
def mat_mul(A, B):
    rowsA = len(A)
    colsA = len(A[0])
    rowsB = len(B)
    colsB = len(B[0])

    if colsA != rowsB:
        print("Cannot multiply these matrices, go learn some linear algebra!")
        return None

    C = []
    for i in range(rowsA):
        C.append([]) # create the next (empty as of now) row i
        for j in range(colsB):
            # Compute C[i][j]
            cij = 0
            for k in range(colsA):
                cij = cij + A[i][k] * B[k][j]
            C[i].append(cij) # this sets C[i][j] = cij

    return C
```

```
mat_mul(A, B)
```

```
↪ [[2, 4], [11, 1]]
```

✓ NumPy version

```
import numpy as np

# NumPy can multiply Python matrices represented as lists of lists.
np.dot(A, B)
```

```
↪ array([[ 2,  4],
         [11,  1]])
```

```
np.matmul(A, B)
```

```
↪ array([[ 2,  4],
         [11,  1]])
```

```
# Let's create matrices as NumPy arrays.
npA = np.array(A)
npB = np.array(B)
print(npA)
print(npB)
print(type(npA))
```

```
↪ [[ 1 -1  2]
   [ 0  3  1]]
   [[1 2]
   [3 0]
   [2 1]]
   <class 'numpy.ndarray'>
```

```
# Use the @ operator to multiply two matrices (Arrays) in NumPy.
npA @ npB
```

```
↻ array([[ 2,  4],  
        [11,  1]])
```