Numpy Cheat Sheet

```
In [1]: import numpy as np
In [2]: %precision 2
   ipython_plain = get_ipython().display_formatter.formatters['text/plain']
   ipython_plain.for_type(np.float64, ipython_plain.lookup_by_type(float));

Create
In [3]: x = np.arange(3) ; print(x, x.dtype)
```

```
x = np.arange(3.0); print(x, x.dtype)
        print( np.arange(1, 3, 0.5) )
        [0 1 2] int64
        [0. 1. 2.] float64
        [1. 1.5 2. 2.5]
In [4]: v3 = np.array([1, 2, 3])
        vЗ
        array([1, 2, 3])
Out[4]:
In [5]: m33 = np.array([[1, 2, 3],
                         [4, 5, 6],
                         [7, 8, 9]])
        m33
       array([[1, 2, 3],
Out[5]:
               [4, 5, 6],
               [7, 8, 9]])
In [6]: from scipy.sparse import csr matrix
        sparse = csr matrix(np.array([[0, 0, 1]]))
        (sparse + sparse)[0, 2]
Out[6]:
In [7]: v2 = np.array([1, 2])
        v2a = v2
        v2a[0] = 3
        # NOTE: Direct assignment results in a reference!
        array([3, 2])
Out[7]:
In [8]: v2b = v2 + 1
        v2b[0] = 5
        v2
        # NOTE: Arithmetic operations results in a copy!
        array([3, 2])
Out[8]:
```

```
In [9]: v2c = v2.copy()
        v2c[0] = 0
        v2
        array([3, 2])
```

Out[9]:

Stacking

```
In [10]: np.column_stack((v3, v3))
         array([[1, 1],
Out[10]:
                [2, 2],
                [3, 3]])
In [11]: np.hstack((v3, v3))
         array([1, 2, 3, 1, 2, 3])
Out[11]:
         np.hstack((m33, m33))
In [12]:
         array([[1, 2, 3, 1, 2, 3],
Out[12]:
                [4, 5, 6, 4, 5, 6],
                [7, 8, 9, 7, 8, 9]])
In [13]: np.vstack((v3, v3))
         array([[1, 2, 3],
Out[13]:
                [1, 2, 3]])
```

Describe

```
In [14]: m33.shape
         (3, 3)
Out[14]:
In [15]:
         m33.ndim
Out[15]:
In [16]: m33.size
Out[16]:
```

Select

```
In [17]: v3[2]
Out[17]:
In [18]: m33[2, 2]
Out[18]:
```

```
In [19]: print( m33[:, 2] )
         print( m33[2, :] )
         # NOTE: Both Slices Return 1-D *Vectors*
         [3 6 9]
         [7 8 9]
In [20]: print( m33[:, 2:3] )
         print( m33[2:3, :] )
         # NOTE: Both Slices Return 2-D *Matrixes*
         [[3]
          [6]
          [9]]
         [[7 8 9]]
In [21]: m33[1]
          # NOTE: Returns 1-D *Vector*
         array([4, 5, 6])
Out[21]:
In [22]: m33[1:3]
          # NOTE: Returns 2-D *Matrix*
         array([[4, 5, 6],
Out[22]:
                [7, 8, 9]])
In [23]: print( m33[1:2] )
         print( m33[[1]] )
          # NOTE: Returns 2-D *Matrix*
         [[4 5 6]]
         [[4 5 6]]
         np.vectorize()
```

```
In [24]: vect fn = np.vectorize(lambda x: -x if x < 3 else 0)
         vect fn(m33)
         array([[-1, -2, 0],
Out[24]:
                [ 0, 0, 0],
                [ 0, 0, 0]])
```

Aggregate

```
In [25]:
         m33
         array([[1, 2, 3],
Out[25]:
                [4, 5, 6],
                 [7, 8, 9]])
In [26]: print( m33.min() )
         print( m33.max() )
         1
         9
In [27]: print( m33.max(axis=0) )
```

```
print( m33.max(axis=1) )
# NOTE: Both Return 1-D *Vectors*

[7 8 9]
[3 6 9]

In [28]: m33.mean(), m33.std(), m33.var()

Out[28]: (5.00, 2.58, 6.67)
```

Reshape

```
In [29]: m23 = np.array([[1, 2, 3],
                          [4, 5, 6]])
         m23
         array([[1, 2, 3],
Out[29]:
                 [4, 5, 6]])
In [30]: m23.reshape(3, 2)
          # Scan Order: Left-to-Right, Top-to-Bottom
         array([[1, 2],
Out[30]:
                [3, 4],
                [5, 6]])
In [31]: m23.reshape(1, -1)
         array([[1, 2, 3, 4, 5, 6]])
Out[31]:
In [32]: print( m23.ravel() )
         print( m23.reshape(-1) )
         print( m23.flatten() )
          # NOTE: All Return 1-D *Vectors*
         [1 2 3 4 5 6]
         [1 2 3 4 5 6]
         [1 2 3 4 5 6]
In [33]: m23.T
         array([[1, 4],
Out[33]:
                 [2, 5],
                 [3, 6]])
In [34]: v3, v3.T
          # NOTE: Vectors *CANNOT* be Transposed!
          (array([1, 2, 3]), array([1, 2, 3]))
Out[34]:
In [35]: v3[None]
          # wrap/encapsulate inside an additional outer dimension
         array([[1, 2, 3]])
Out[35]:
In [36]: v3.reshape(-1, 1)
          # v3[None].T
          # convert vector to" column vector"
```

Linear Algebra

```
In [38]: np.triu(m33), np.tril(m33)
         (array([[1, 2, 3],
Out[38]:
                 [0, 5, 6],
                  [0, 0, 9]]),
          array([[1, 0, 0],
                 [4, 5, 0],
                 [7, 8, 9]]))
In [39]: v3
         array([1, 2, 3])
Out[39]:
In [40]:
          # vector dot product (1*1 + 2*2 + 3*3)
         14
Out[40]:
In [41]: m33 @ m33
          # matrix multiplication
         array([[ 30, 36, 42],
Out[41]:
                [ 66, 81, 96],
                [102, 126, 150]])
In [42]: m33 * m33
          # element-wise multiplication
         array([[ 1, 4, 9],
Out[42]:
                [16, 25, 36],
                [49, 64, 81]])
In [43]: np.linalg.inv(np.array([[1,0], [0,2]]))
          # matrix inverse
         array([[1. , 0. ],
Out[43]:
                [0., 0.5]])
```

Random

```
In [45]:
         np.random.seed(0)
In [46]:
         np.random.random(3)
          # uniform between 0.0 and 1.0
         array([0.55, 0.72, 0.6])
Out[46]:
In [47]:
         [np.random.randint(3) for _ in range(10)]
          # np.random.randint(<max-excl>)
          [1, 1, 2, 0, 2, 0, 0, 0, 2, 1]
Out[47]:
         [np.random.randint(1, 4) for in range(10)]
In [48]:
          # np.random.randint(<min-incl>, <max-excl>)
         [3, 3, 1, 2, 2, 2, 2, 1, 2, 1]
Out[48]:
         np.random.uniform(1, 5, 10)
In [49]:
          # np.random.uniform(<min>, <max>, <size>)
         array([1.08, 4.33, 4.11, 4.48, 4.91, 4.2, 2.85, 4.12, 1.47, 3.56])
Out[49]:
In [50]: np.random.normal(5, 1, (3, 4))
          # np.random.normal(<mean>, <stddev>, <size>)
         array([[2.45, 5.65, 5.86, 4.26],
Out [50]:
                [7.27, 3.55, 5.05, 4.81],
                [6.53, 6.47, 5.15, 5.38]])
In [51]: np.random.choice(v3, size=5, replace=True)
          # 'replace': with/without replacement
         array([3, 1, 2, 2, 2])
Out[51]:
In [52]: x = np.arange(5)
         np.random.shuffle(x)
          # shuffle *in-place*
         array([0, 4, 2, 3, 1])
Out[52]:
```

Filter

```
In [53]: m23
```

```
In [54]: m23 < 4
         array([[ True, True, True],
Out[54]:
                [False, False, False]])
          (m23 < 2) \mid (m23 > 4)
In [55]:
         array([[ True, False, False],
Out[55]:
                [False, True, True]])
         m23[(m23 < 4) & (m23 >= 1)]
In [56]:
          # NOTE: Returns 1-D *Vector*
         array([1, 2, 3])
Out [56]:
In [57]: np.where (m23 < 4, 1, -1)
          # np.where(<condition>, <val if true>, <val if false>)
         array([[ 1, 1, 1],
Out[57]:
                [-1, -1, -1]]
         Misc
In [58]: v10 = np.linspace(0, 5, 6)
         v10
          # np.linspace(<start-incl>, <stop-incl>, <num samples>)
         array([0., 1., 2., 3., 4., 5.])
Out [58]:
In [59]: np.percentile(v10, [25, 50, 100])
          # Returns value(s) corresponding to given percentile(s)
         array([1.25, 2.5, 5. ])
Out[59]:
In [60]: x = np.array([[1, 2, 1], [1, 3, 1], [1, 2, 1]])
         print('x:\n', x)
         print('Unique Elements:\n', np.unique(x))
         print('Unique Rows:\n', np.unique(x, axis=0))
         print('Unique Cols:\n', np.unique(x, axis=1))
         х:
          [[1 2 1]
          [1 3 1]
          [1 2 1]]
         Unique Elements:
          [1 2 3]
         Unique Rows:
          [[1 2 1]
          [1 3 1]]
         Unique Cols:
          [[1 2]
          [1 3]
          [1 2]]
 In [ ]:
```

Out[53]: array([[1, 2, 3],

[4, 5, 6]])