Numpy

December 16, 2020

0.0.1 Working with Numpy (Used For Mathematical Operation)

```
[1]: import numpy as np
[2]: a = np.array([1,2,3,4])
[3]: print(a)
     print(type(a))
     print(a.shape)
    [1 2 3 4]
    <class 'numpy.ndarray'>
    (4,)
[4]: b = np.array([[1],[2],[3],[4],[5]])
[5]: print(b)
     print(b.shape)
    [[1]
     [2]
     [3]
     [4]
     [5]]
    (5, 1)
      c = np.array([[1,2,3],[4,5,6]])
[6]:
[7]: print(c)
     print(c.shape)
     print(c[1][1])
    [[1 2 3]
     [4 5 6]]
    (2, 3)
    5
```

```
[8]: #Create Zeroes , Ones Array
      a = np.zeros((3,3))
      print(a)
      a = np.ones((3,3))
      print(a)
     [[0. 0. 0.]
      [0. 0. 0.]
      [0. 0. 0.]]
     [[1. 1. 1.]
      [1. 1. 1.]
      [1. 1. 1.]]
 [9]: #Array of some constants
      #First value is a tuple giving shape of matrix
      c = np.full((3,2),5)
      print(c)
     [[5 5]
      [5 5]
      [5 5]]
[10]: #Identity matrix
      d = np.eye(4)
      print(d)
     [[1. 0. 0. 0.]
      [0. 1. 0. 0.]
      [0. 0. 1. 0.]
      [0. 0. 0. 1.]]
[11]: #random matrix
      randomMatrix=np.random.random((2,3))
      print(randomMatrix)
     [[0.36866702 0.86576119 0.74815607]
      [0.91859001 0.01864204 0.4557789 ]]
[12]: # Starting Point : Ending Point
      # Starting point is included but not ending point
      print(randomMatrix[:,1:2])
     [[0.86576119]
      [0.01864204]]
[13]: print(randomMatrix[:,2])
```

```
[14]: randomMatrix[0:,2:]=2
      print(randomMatrix)
     [[0.36866702 0.86576119 2.
      [0.91859001 0.01864204 2.
                                        11
[15]: #Selecting columns
      z = np.zeros((3,3))
      z[1,:]=1
      z[:,2]=1
      z[:,-1]=-1
      print(z)
     [[ 0. 0. -1.]
      [ 1. 1. -1.]
      [ 0. 0. -1.]]
[16]: #Datatype of Numpy Array
      x = np.ones((5,5))
[17]: print(x)
      print(type(x))
      print(x.dtype)
     [[1. 1. 1. 1. 1.]
      [1. 1. 1. 1. 1.]
      [1. 1. 1. 1. 1.]
      [1. 1. 1. 1. 1.]
      [1. 1. 1. 1. 1.]]
     <class 'numpy.ndarray'>
     float64
[43]: x = np.ones((5,5),dtype=np.int64)
[44]: print(x)
      print(type(x))
      print(x.dtype)
     [[1 1 1 1 1]
      [1 1 1 1 1]
      [1 1 1 1 1]
      [1 1 1 1 1]
      [1 1 1 1 1]]
     <class 'numpy.ndarray'>
     int64
```

[0.74815607 0.4557789]

```
[18]: # Data type conversion
      k = np.ones((5,5),dtype=np.int64)
      print(k.dtype)
     int64
     0.1 Mathematical Operation
[19]: x = np.array([[1,2],[5,6]])
      y= np.array([[3,4],[2,2]])
      print(x+y)
      print(np.add(x,y))
     [[4 6]
      [7 8]]
     [[4 6]
      [7 8]]
[20]: #Matrix Multiplaction (Dot Product)
      print(x.dot(y))
      print(np.dot(x,y))
     [[ 7 8]
      [27 32]]
     [[ 7 8]
      [27 32]]
[21]: #Multiplication Of (Dot Product) of Vectors - Scalar
      a = np.array([1,2,6])
      b= np.array([3,4,2])
      print(a.dot(b))
      # 1*3+2*4+6*2
     23
[22]: #Sum of array
      su = np.array([1,2,3,4,5])
      print(sum(su))
      print(np.sum(su))
     15
     15
[23]: #Sum Along X-Axis
      mat=np.array([[1,2,3],[4,5,6],[7,8,9]])
      print(np.sum(mat,axis=0)) #Along column
      print(np.sum(mat,axis=1)) #Along rows
```

```
[12 15 18]
[ 6 15 24]
```

0.2 Stacking of arrays

```
[24]: a=np.array([1,2,3])
      b=np.array([4,5,6])
      b=b**2
      print(b)
     [16 25 36]
[25]: np.stack((a,b),axis=0)
[25]: array([[ 1, 2, 3],
             [16, 25, 36]])
[26]: np.stack((a,b),axis=1)
[26]: array([[ 1, 16],
             [2, 25],
             [3, 36]])
     0.2.1 Reshape Numpy Array
      #First define rows and second define columns
```

```
[45]: re = np.array([[1,2,3,4],[5,6,7,8]])
      re = re.reshape((4,2))
      print(re)
      \#If we give -1 , it automatically decides according to row the
      #number of columns
      ree = re.reshape((4,-1))
      print(ree)
      #For automatic column
      pree = re.reshape((-1,8))
      print(pree)
     [[1 2]
```

```
[3 4]
 [5 6]
[7 8]]
[[1 2]
[3 4]
 [5 6]
 [7 8]]
[[1 2 3 4 5 6 7 8]]
```

0.3 Numpy Random Module

- rand: Random values in a given shape
- randn : Return a sample from the 'standard normal' distribution
- randint : Return random integers from low to high
- random: Return random folats in the half open interval
- choice: Generates a random sample from given 1-d Array
- shuffle : Shuffle the contents of a sequence

```
[28]: a = np.arange(10)+5 print(a)
```

[5 6 7 8 9 10 11 12 13 14]

```
[50]: np.random.randint(50,60,10)
```

```
[50]: array([59, 50, 50, 51, 57, 59, 52, 50, 53, 50])
```

```
[30]: #To random shuffle of array
np.random.shuffle(a)
print(a)
```

[651413107981211]

```
[31]: #Returns Values from a Standard Normal Distributions
a = np.random.randn(2,3)
print(a)
```

```
[[-1.19853043 -0.24379676 1.00680161]
[-0.67026228 -1.26712389 -0.87927845]]
```

```
[32]: #Randomly pick one element from a array
element = np.random.choice([1,4,3,2,11,27])
print(element)
```

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0.4 Numpy Functions :- Statistics

- min,max
- mean
- median
- average
- variance
- standard deviation

0.4.1 MIN

```
[33]: a = np.array([[1,2,3,4],[5,6,7,8]])
      print(a)
      print(np.min(a))
      #Along columns
      print(np.min(a,axis=0))
      #Along Rows
      print(np.min(a,axis=1))
     [[1 2 3 4]
      [5 6 7 8]]
     [1 2 3 4]
     [1 5]
     0.4.2 MAX
[51]: a = np.array([[1,2,3,4],[5,6,7,8]])
      print(a)
      print(np.max(a))
      #Along columns
      print(np.max(a,axis=0))
      #Along Rows
      print(np.max(a,axis=1))
     [[1 2 3 4]
      [5 6 7 8]]
     [5 6 7 8]
     [4 8]
     0.4.3 MEAN & MEDIAN
[52]: #Mean (Average of all elements)
      b = np.array([1,2,3,4,5])
      m = sum(b)/5
      print(m)
      print(np.mean(b))
      print(np.median(b))
      a = np.array([[1,2,3,4],[5,6,7,8]])
      #Along columns
      print(np.mean(a,axis=0))
```

3.0

#Along Rows

print(np.mean(a,axis=1))

3.0

```
3.0
[3. 4. 5. 6.]
[2.5 6.5]
```

0.4.4 STANDARD DEVIATION

```
[55]: #Standard deviation and Variance
    c = np.array([1,5,4,2,0])
    print(np.median(c))

u = np.mean(c)
    myStd = np.sqrt(np.mean(abs(c-u)**2))
    print(myStd)
    print(myStd**2)
    dev = np.std(c)
    var = np.var(c) # Square of standard deviation
    print(dev)
    print(var)
```

- 2.0
- 1.854723699099141
- 3.440000000000001
- 1.854723699099141
- 3.4400000000000004

0.5 Statistics :- Interview Question

Given a running stream of numbers, compute mean and variance at any given point

We will spend O(n) time computing the new mean and variance each time for each new number added.

```
[36]: import cv2
import matplotlib.pyplot as plt

[37]: img = cv2.imread('numpy.jfif')

[38]: print(img)

[[[175 182 177]
        [172 179 174]
        [170 177 172]
        ...
        [167 181 177]
        [162 176 170]
        [159 173 167]]

[[176 183 178]
        [174 181 176]
```

```
[171 178 173]
       [160 174 170]
       [155 169 163]
       [152 166 160]]
      [[177 184 179]
       [175 182 177]
       [173 180 175]
       [166 180 176]
       [161 175 169]
       [158 172 166]]
      [[151 161 155]
       [154 164 158]
       [157 167 161]
       [175 186 184]
       [175 186 184]
       [175 186 184]]
      [[127 137 131]
       [134 144 138]
       [142 152 146]
       [175 186 184]
       [175 186 184]
       [175 186 184]]
      [[ 63 73 67]
       [ 74 84 78]
       [88 98 92]
       [175 186 184]
       [175 186 184]
       [175 186 184]]]
[39]: print(img.shape)
     (1280, 860, 3)
[40]: plt.figure(figsize=(20,20))
      plt.imshow(img)
      plt.axis("off")
```

plt.show()

Date Page
rage
Criven a sunning stream of numbers compute mean & voxiance at any given point
X ₁ , X ₂ X _{n-1}
$E(x) = Mean = 1 \sum_{i=1}^{n} X_i$
Also known as Expected value of X
Vasiance = 02
5. Standard deviation = $\frac{1}{n} \sum_{i=1}^{n} (x_i - \mu)^2$
So, we try to derive other formula because this is
number than it will take O(n)
$\sigma^{2} = 1 \left[\sum_{i=1}^{n} (X_{i} - u)^{2} \right]$ $= 1 \left[\sum_{i=1}^{n} (X_{i}^{2} + \sum_{i=1}^{n} X_{i}^{2} - 2u \sum_{i=1}^{n} X_{i} \right]$
= $\frac{1}{2} \sum_{i=1}^{n} x_{i}^{2} + \sum_{i=1}^{n} x_{i}^{2} - 2u \sum_{i=1}^{n} x_{i}$
Stand Himph
$= 1 \left[\sum X_i^2 + n \tilde{u}^2 - 2u \sum_{i=1}^n X_i^2 \right] \approx 3$
$= \sum X_i^2 + \mu^2 - 2\mu / \sum_{i=1}^n X_i / \sum M \text{ or } E(X)$
$= \sum X_i^2 + \mu^2 - 2\mu \left(\sum_{i=1}^n X_i\right) \mathcal{I}_{i} M \text{ or } E(X)$
= \(\Si\)2 \(\rightarrow\) Mean: Sum
n N
$= \mathbf{E}(X^2) - \mathbf{E}(X)^2 \qquad \int \mathbf{E}(X^2) : \operatorname{Sum}^2$
Survivue can be done easily. We can get this
Values easily

[]:[