

```
import numpy as np
```

```
from google.colab import files
uploaded=files.upload()
```

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Saving sudoku\_solution.npy to sudoku\_solution.npy

Saving sudoku\_game.npy to sudoku\_game.npy

```
sudoku_game=np.load('sudoku_game.npy')
sudoku_game
```

```
array([[0, 0, 5, 0, 0, 9, 0, 0, 1],
       [0, 7, 0, 0, 6, 0, 0, 4, 3],
       [0, 0, 6, 0, 0, 2, 0, 8, 7],
       [1, 9, 0, 0, 0, 7, 4, 0, 0],
       [0, 5, 0, 0, 8, 3, 0, 0, 0],
       [6, 0, 0, 0, 0, 0, 1, 0, 5],
       [0, 0, 3, 5, 0, 8, 6, 9, 0],
       [0, 4, 2, 9, 1, 0, 3, 0, 0]])
```

+ Code

+ Text

## Flattening & Reshaping of Data

```
#Flatten Sudoku_game
flatten_sudoku=sudoku_game.flatten()
flatten_sudoku
```

```
array([0, 0, 5, 0, 0, 9, 0, 0, 1, 0, 7, 0, 0, 6, 0, 0, 4, 3, 0, 0, 6, 0,
       0, 2, 0, 8, 7, 1, 9, 0, 0, 0, 7, 4, 0, 0, 0, 5, 0, 0, 8, 3, 0, 0,
       0, 6, 0, 0, 0, 0, 1, 0, 5, 0, 0, 3, 5, 0, 8, 6, 9, 0, 0, 4, 2,
       9, 1, 0, 3, 0, 0])
```

```
sudoku_game
```

```
array([[0, 0, 5, 0, 0, 9, 0, 0, 1],
       [0, 7, 0, 0, 6, 0, 0, 4, 3],
       [0, 0, 6, 0, 0, 2, 0, 8, 7],
       [1, 9, 0, 0, 0, 7, 4, 0, 0],
       [0, 5, 0, 0, 8, 3, 0, 0, 0],
       [6, 0, 0, 0, 0, 0, 1, 0, 5],
       [0, 0, 3, 5, 0, 8, 6, 9, 0],
       [0, 4, 2, 9, 1, 0, 3, 0, 0]])
```

```
#Reshaping Concept
reshaped_sudoku=flatten_sudoku.reshape((9,8))
reshaped_sudoku
```

```
array([[0, 0, 5, 0, 0, 9, 0, 0],
       [1, 0, 7, 0, 0, 6, 0, 0],
       [4, 3, 0, 0, 6, 0, 0, 2],
       [0, 8, 7, 1, 9, 0, 0, 0],
       [7, 4, 0, 0, 0, 5, 0, 0],
       [8, 3, 0, 0, 0, 6, 0, 0],
       [0, 0, 0, 1, 0, 5, 0, 0],
       [3, 5, 0, 8, 6, 9, 0, 0],
       [4, 2, 9, 1, 0, 3, 0, 0]])
```

## Numpy Data Types Numpy Vs Python Data Types Sample Python Data Types

1. int
2. Float

### Sample Numpy Datatypes

1. np.int64
2. np.int32
3. np.float64
4. np.float32

### The .dtype Attribute

```
np.array([1.32,5.78,175.175]).dtype
```

```
dtype('float64')
```

### String Data

```
np.array(['Introduction','to','Numpy']).dtype
dtype('<U12')
```

### Dtype as an arguments

```
float_32_array=np.array([1.32,5.78,175.175],dtype=np.float32)
float_32_array
float_32_array.dtype
dtype('float32')
```

### Type Conversion

```
boolean_arr=np.array([[True,False],[False,False]],dtype=np.bool_)
boolean_arr.astype(np.int32)
#boolean_arr

<ipython-input-5-10f12a7faf92>:1: DeprecationWarning: `np.bool` is a deprecated alias for the builtin `bool`. To silence
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecation
boolean_arr=np.array([[True,False],[False,False]],dtype=np.bool)
array([[1, 0],
       [0, 0]], dtype=int32)
```

### Type Coercion

```
np.array([True,"BillGates",42,42.42])
array(['True', 'BillGates', '42', '42.42'], dtype='<U32')
```

### Type Coercion Hierarchy: Adding a Float to an array of integers will change all integers into float?

```
np.array([0,42,42.42]).dtype
dtype('float64')
```

### Adding an integer to an array of booleans will change all booleans in to int

```
np.array([True,False,99]).dtype
dtype('int64')
```

```
# Create an Array of Zeros with Three Rows and Two Cols
zero_arr=np.zeros((3,2))
zero_arr

array([[0., 0.],
       [0., 0.],
       [0., 0.]])
```

### Assignment: np.array([34.62,70.13,True,"TOM"]).astype(np.int64) ==>?

```
sudoku_game.dtype # Trying to Find datatype of Sudoku Game – Thridparty Numpy Array
dtype('int64')
```

```
sudoku_game

array([[0, 0, 5, 0, 0, 9, 0, 0, 1],
       [0, 7, 0, 0, 6, 0, 0, 4, 3],
       [0, 0, 6, 0, 0, 2, 0, 8, 7],
       [1, 9, 0, 0, 0, 7, 4, 0, 0],
       [0, 5, 0, 0, 8, 3, 0, 0, 0],
       [6, 0, 0, 0, 0, 0, 1, 0, 5],
       [0, 0, 3, 5, 0, 8, 6, 9, 0],
       [0, 4, 2, 9, 1, 0, 3, 0, 0]])
```

## Indexing 1D Array

```
arr_1=np.array([2,4,6,8,10])
arr_1
```

```
array([ 2,  4,  6,  8, 10])
```

```
arr_1[3]
```

```
8
```

```
sudoku_game
```

```
array([[0, 0, 5, 0, 0, 9, 0, 0, 1],
       [0, 7, 0, 0, 6, 0, 0, 4, 3],
       [0, 0, 6, 0, 0, 2, 0, 8, 7],
       [1, 9, 0, 0, 0, 7, 4, 0, 0],
       [0, 5, 0, 0, 8, 3, 0, 0, 0],
       [6, 0, 0, 0, 0, 0, 1, 0, 5],
       [0, 0, 3, 5, 0, 8, 6, 9, 0],
       [0, 4, 2, 9, 1, 0, 3, 0, 0]])
```

```
sudoku_game[2,4]
```

```
0
```

```
arr_1
```

```
array([ 2,  4,  6,  8, 10])
```

```
arr_1[3]
```

```
8
```

```
arr_1[:3]
```

```
array([2, 4, 6])
```

```
arr_1[3:]
```

```
array([ 8, 10])
```

```
sudoku_game[:,3]
```

```
array([0, 0, 0, 0, 0, 0, 5, 9])
```

## #Slicing 1D Array

```
arr_1[2:4]
```

```
arr_1
```

```
array([ 2,  4,  6,  8, 10])
```

```
arr_1[2:4]
```

```
array([6, 8])
```

## #Slicing 2D Array

```
sudoku_game[3:6,3:6]
```

```
array([[0, 0, 7],
       [0, 8, 3],
       [0, 0, 0]])
```

```
sudoku_game[3:6:2, 3:6:2] # What Expected?
```

```
array([[0, 7],
       [0, 0]])
```

```
sudoku_game
```

```
array([[0, 0, 5, 0, 0, 9, 0, 0, 1],
       [0, 7, 0, 0, 6, 0, 0, 4, 3],
       [0, 0, 6, 0, 0, 2, 0, 8, 7],
       [1, 9, 0, 0, 0, 7, 4, 0, 0],
       [0, 5, 0, 0, 8, 3, 0, 0, 0],
```

```
[6, 0, 0, 0, 0, 0, 1, 0, 5],
[0, 0, 3, 5, 0, 8, 6, 9, 0],
[0, 4, 2, 9, 1, 0, 3, 0, 0]])
```

```
#Sorting an array?
np.sort(sudoku_game)
```

```
array([[0, 0, 0, 0, 0, 0, 1, 5, 9],
       [0, 0, 0, 0, 0, 3, 4, 6, 7],
       [0, 0, 0, 0, 0, 2, 6, 7, 8],
       [0, 0, 0, 0, 0, 1, 4, 7, 9],
       [0, 0, 0, 0, 0, 0, 3, 5, 8],
       [0, 0, 0, 0, 0, 0, 1, 5, 6],
       [0, 0, 0, 0, 3, 5, 6, 8, 9],
       [0, 0, 0, 0, 1, 2, 3, 4, 9]])
```

Keep in Mind in a Matrix Rows Considered as Axis=1 Cols Considers as Axis=0

```
np.sort(sudoku_game,axis=1) # You can make axis=0 too
```

```
array([[0, 0, 0, 0, 0, 0, 1, 5, 9],
       [0, 0, 0, 0, 0, 3, 4, 6, 7],
       [0, 0, 0, 0, 0, 2, 6, 7, 8],
       [0, 0, 0, 0, 0, 1, 4, 7, 9],
       [0, 0, 0, 0, 0, 0, 3, 5, 8],
       [0, 0, 0, 0, 0, 0, 1, 5, 6],
       [0, 0, 0, 0, 3, 5, 6, 8, 9],
       [0, 0, 0, 0, 1, 2, 3, 4, 9]])
```

```
from google.colab import files
uplaoded=files.upload()
```

Choose files No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.  
Saving tree\_census.nov to tree\_census.nov

Imagine You are a researcher working with data from New York City Tree census. Each Row of the tree\_census . 2D Array List - Capstone Project

Select All rows of block id data from the second columns

```
tree_census=np.load('tree_census.npy')
tree_census
```

```
array([[ 3, 501451, 24, 0],
       [ 4, 501451, 20, 0],
       [ 7, 501911, 3, 0],
       ...,
       [1198, 227387, 11, 0],
       [1199, 227387, 11, 0],
       [1210, 227386, 6, 0]])
```

```
block_ids=tree_census[:,1]
block_ids
```

```
array([[501451, 501451, 501911, 501911, 501911, 501911, 501911, 501911],
       [501911, 501911, 501911, 501911, 501909, 501909, 501909, 501909],
       [501909, 501909, 501909, 501967, 501967, 501967, 501134, 501134],
       [501134, 501134, 501134, 501134, 501134, 501134, 501966, 501966],
       [501966, 501966, 501966, 501966, 500846, 500846, 500846, 500846],
       [500846, 500846, 500836, 500836, 500836, 500836, 500836, 500836],
       [500836, 500836, 500836, 500836, 500836, 501882, 501882, 501882],
       [501882, 501882, 501882, 501882, 501882, 501882, 501882, 501113],
       [501113, 501113, 501113, 501113, 500917, 500917, 500917, 501323],
       [501323, 501323, 501323, 501323, 501323, 501323, 501451, 501451],
       [516296, 516296, 516296, 516296, 516296, 516296, 516296, 516296],
       [516296, 516296, 516296, 516296, 516296, 516296, 503178, 503178],
       [503178, 503178, 503178, 503178, 501451, 501451, 501451, 501451],
       [501451, 501451, 501451, 501451, 501451, 501874, 501874, 501874],
       [501874, 501874, 501874, 501874, 501874, 501874, 501874, 501874],
       [501874, 501874, 501874, 501874, 501874, 501874, 501874, 501874],
       [501874, 501874, 501874, 501874, 501874, 501874, 501874, 501874],
       [501874, 501874, 501874, 501874, 501874, 501874, 501874, 501874],
       [501874, 501874, 501874, 501874, 501942, 501942, 501942, 501942],
       [314009, 314009, 314009, 314009, 314009, 314009, 314009, 314009],
       [314009, 314009, 314009, 314009, 314009, 314009, 314009, 314009],
       [314009, 314009, 314009, 314009, 314009, 314009, 314009, 501867],
       [501867, 501841, 501841, 501841, 501892, 501892, 501892, 501892],
       [501892, 501892, 501892, 501892, 501892, 501892, 501892, 501892],
       [501892, 501892, 501892, 501892, 501892, 501892, 501897, 501897],
       [501897, 501897, 501897, 501897, 501897, 501897, 501897, 501897]]
```

501897	501897	501897	501897	501897	501897	501897	501897
501897	501897	501897	501893	501893	501893	501893	501893
501893	501893	501893	501893	501893	501893	501893	107255
107255	107255	107255	413051	413051	413051	413051	413051
413051	413051	413051	413051	413051	413051	413051	413051
413051	413051	413051	413051	413051	413051	413051	413051
413051	413051	413051	413051	413051	413052	413052	413052
413052	413052	413052	413052	413052	107288	107288	107288
413086	413086	413086	413086	413086	413086	413086	413086
413086	413086	207803	207803	207803	207803	207803	207803
207803	207803	207803	107278	107278	107278	107188	107188
107188	107188	107188	207088	207088	207088	207088	107256
107256	107256	207945	207945	207945	207945	207945	207112
207112	207112	507568	507568	507568	107277	107277	107277
107277	107277	107277	107277	107277	107285	107285	107285
107285	207802	207802	207802	207041	207041	207041	207041
207041	516271	516271	516271	516271	516271	516271	207199
207199	207199	207199	207199	207199	207199	207199	207199
207199	507598	507598	207201	207201	507613	507613	507613
507613	507599	507599	507599	507599	507599	507599	507599
507599	507599	507599	207200	207200	207200	207200	207200
207200	207200	207200	207200	207200	207200	207200	207200
507614	507614	507539	507539	507539	507539	507539	507539
507615	507615	507615	507615	507615	507615	343368	343368
349684	349684	349684	349687	349687	349687	349687	342452
342452	342452	342452	342453	342453	342453	342453	344023
344023	344023	344023	344023	344023	344023	342454	342454
342454	342454	342454	342454	349688	349688	349688	349688
349688	349688	349688	349688	349688	349688	349688	349688
349688	349688	349688	349688	342455	342455	342455	342455
342455	342456	342456	342457	342457	342457	342457	342457

```
# Print the First Five block_ids
block_ids[:5]
```

```
array([501451, 501451, 501911, 501911, 501911])
```

#Steeping into 2D - Now Assume that your research requires you to take an admittedly unrepresentative sample of trunk diameter

```
# Create an array of the first 100 trunk diameters from tree census
hundred_dia=tree_census[:100,2]
hundred_dia
```

```
array([24, 20, 3, 3, 4, 4, 4, 4, 4, 3, 3, 4, 2, 2, 3, 4, 4,
       4, 0, 14, 3, 3, 4, 7, 8, 7, 8, 7, 5, 6, 5, 5, 17, 0, 19,
       21, 18, 4, 5, 3, 4, 3, 4, 4, 13, 13, 13, 5, 4, 4, 4, 11, 5,
       4, 5, 8, 51, 7, 4, 15, 3, 8, 6, 6, 3, 4, 3, 2, 3, 3,
       6, 5, 5, 5, 5, 9, 4, 4, 7, 7, 6, 5, 4, 4, 5, 5, 5,
       7, 3, 5, 3, 3, 6, 6, 8, 7, 4, 5, 4, 4, 4, 4])
```

**Extract trunk diameter information from smallest to largest**

```
sorted_trunk_dia=np.sort(tree_census[:,2])
sorted_trunk_dia
```

[illegible]

```

8, 8, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9,
9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9, 9,
9, 9, 9, 9, 9, 9, 9, 9, 9, 10, 10, 10, 10, 10, 10, 10,
10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10,
10, 10, 10, 10, 10, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11,
11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11,
11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11, 11,
11, 11, 11, 11, 11, 11, 11, 11, 11, 12, 12, 12, 12, 12, 12, 12,
12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12,
12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12, 12,
12, 12, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13,
13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13, 13,
13, 13, 13, 13, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14,
14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14, 14,
14, 14, 14, 14, 14, 14, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15,
15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15, 15,
15, 15, 15, 15, 15, 15, 15, 16, 16, 16, 16, 16, 16, 16, 16, 16,
16, 16, 16, 16, 16, 16, 16, 16, 16, 17, 17, 17, 17, 17, 17, 17,
17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17, 17,
17, 17, 17, 17, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18,
18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 18, 19, 19, 19,
19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19,
19, 19, 19, 19, 19, 19, 19, 20, 20, 20, 20, 20, 20, 20, 20, 20,
20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 21, 21, 21, 21,
21, 21, 21, 21, 21, 21, 21, 21, 21, 21, 21, 21, 21, 21, 22, 22,
22, 22, 22, 22, 22, 22, 22, 22, 22, 22, 22, 22, 22, 23, 23,
23, 23, 23, 23, 23, 23, 23, 23, 24, 24, 24, 24, 24, 24, 24, 24,
25. 25. 25. 25. 25. 25. 25. 25. 25. 25. 25. 26. 26. 26. 26. 26. 26.

```

## Concept of Filtering

1. Masks and Fancy Indexing
2. np.where()

### #Boolean Masking

```

one_to_five=np.arange(1,6)
one_to_five

```

```

array([1, 2, 3, 4, 5])

```

```

mask=one_to_five % 2==0
mask

```

```

array([False,  True, False,  True, False])

```

```

one_to_five[mask]

```

```

array([2, 4])

```

### # Implement the same with 2D Ones 2D Fancy Indexing

```

classroom_ids_and_sizes=np.array([[1,22],[2,21],[3,27],[4,26]])
classroom_ids_and_sizes

```

```

array([[ 1, 22],
       [ 2, 21],
       [ 3, 27],
       [ 4, 26]])

```

```

classroom_ids_and_sizes[:,0][classroom_ids_and_sizes[:,1]%2==0]

```

```

array([1, 4])

```

### #Filtering with np.where?

```

np.where(classroom_ids_and_sizes[:,1]%2==0)

```

```

(array([0, 3]),)

```

