

53) `import numpy as np` } to create numpy
`a = np.array([0, 1, 2, 3])` } array using python
`print(a)` | a
→ `[0 1 2 3]` | `array([0 1 2 3])` list.
 `print(np.arange(10))` } to create numpy array
→ `[0 1 2 3]` from any integer
`[0 1 2 3 4 5 6 7 8 9]`

54) To calculate time to run a program -
Python list
`l = list(range(1000))`
% timeit `[i**2 for i in l]` } normally.
→ $237 \mu\text{s} \pm 1.36 \mu\text{s}$ per loop

`a = np.arange(1000)` } using numpy.
% time it. `a**2`
→ $1.03 \mu\text{s} \pm 12.3 \mu\text{s}$ per loop

55) Creating arrays in numpy -
1-D array (1-dimensional) → list (it is called e)
56) \oplus `a = np.array([0, 1, 2, 3])`
`a`
→ `array([0, 1, 2, 3])`

56) 1-D array → list
2-D array → matrix
n-D array → Tensor

57) # print dimension { find dimension }

a. ndim

→ 1

58) # shape { find shape }

a. shape

→ (1,)

59) len(a) { in tensor there is nothing like
4. length it's only shape }

60) #= 2-D, 3-D, ---

b = np.array ([[0, 1, 2], [3, 4, 5]])

b

→ array ([0, 1, 2],
[3, 4, 5])

61.) b.ndim

→ 2

62.) b.shape

→ (2, 3)

63.) len(b)

→ 2

(only return the
1st dim. 2nd dim. size of 1st dim.)

64.) in 3-D array

n. shape

→ (2, 2, 2)

) n = ([[1, 2], [2, 3], [3, 4]])

65.) function for creating array-

a = np.arange(2)

→ array ([0, 1]),

start ↓ end (exclusive) step ↓

66.) $b = \text{np.arange}(1, 10, 2)$

b

→ array([1, 3, 5, 7, 9])

67.) linspace →

start ↓ end (inclusive) no. of points ↓

a = np.linspace(0, 1, 6) (divide in 6 parts)

a

→ array([0. , 0.2 , 0.4 , 0.6 , 0.8 , 1.0])

1st part 2nd part 3rd part 4th part 5th part

68.) create array of ones

a = np.ones((2, 2))

a

→ array([[1., 1.],

[1., 1.]])

69.) create array of zeroes

b = np.zeros((2, 2))

b

→ array([[0., 0.],

[0., 0.]])

70.) diagonal matrix -

c = np.eye(2)

c

→ array([[1.0, 0.],

[0., 1.]])

71.) more than dim.

d = np.eye(3, 2)

d

→ array([[1.0, 0.],

[0., 1.],

[0., 0.]])

72.) create array using `diag` f"-

$a = np.diag([1, 2, 3, 4])$.

a

$\rightarrow \text{array}([[1, 0, 0, 0], [0, 2, 0, 0], [0, 0, 3, 0], [0, 0, 0, 4]])$

73.) $np.diag(a)$ # Extract diagonal

$\rightarrow \text{array}[1, 2, 3, 4]$)

74.) # create array using random. uniform distribution over $[0, 1]$

$a = np.random.rand(3)$

a

$\rightarrow \text{array}[0.85434586, 0.05106692, 0.37337949]$)

75.) $a = np.random.randn(4)$. (# std. normal

a

$\rightarrow \text{array}[1.9940753e+00, -1.33836224e+00, 3.6739503e-04]$)
in this the prob. of values near 0 is high)

76.) Basic data type-

$a = np.arange(10)$

a. `dtype`

$\rightarrow \text{dtype}'int64'$)

77.) to specify which data type we want -

`a = np.array([10, dtype='float64'])`
a

→ array [0., 1., 2., 3., 4., 5., 6., 7., 8., 9.]

78.) # The default data type is float for zeros & ones -

`a = np.zeros((2,2))`

`print(a)`

`a.dtype`

→ [[0., 0.]]

[0. 0.]

`dtype('float64')`

79.) Complex datatype.

`c = np.array([1+2j, 2+4j])`

`print(c.dtype)`

→ complex128

80.) `b = np.array([True, False])` # Boolean datatype.

`print(b.dtype)`

→ bool

81.) `s = np.array(['Ram', 'Rom'])` # string datatype

`s.dtype`

→ dtype('S6')

Q2.) Each built-in data type has a character code that uniquely identifies it.

'b' - boolean

'i' - (signed) integer [range -2^{149} to $+2^{149}$]

'u' - (unsigned) integer [range 0 to 4×10^9]

'f' - floating pt.

'c' - complex - floating pt

'm' - time delta [intime format or date format]

'M' - datetime

'O' - (Python) objects → if you don't know which datatype it is

'S', 'a' - (byte) string

'V' - raw data (void)

'U' - unicode

Q3.) samplearray = np.array([1, 2, 3, 4, 5])

samplearray * 5

→ array([5, 10, 15, 20, 25])

Q4.) samplearray / 2

→ array([0.5, 1.0, 1.5, 2., 2.5])

Q5.) print(type(samplearray))

→ <class 'numpy.ndarray'>

Q6.) arr = np.array([[1, 2], [3, 4]])

arr

→ array([[1, 2],
[3, 4]])

87.) arr. shape

 $\rightarrow (2, 2)$ 88.) Reshape the array - (to gain a particular used shape).~~arr = a.reshape arr = arr.reshape (1, 4).~~

arr. shape

 $\rightarrow (1, 4)$.89.) Random -

from numpy import

random

(from numpy import)

{import everything}

{(import only random) } {import random from 1st file}.

90.) $x = \text{random.randint}(100, \text{size} = (3, 5))$. $\Rightarrow x$ $\rightarrow \text{array}([[48, 27, 38, 57, 27],$
 $[26, 14, 26, 46, 61],$
 $[87, 80, 3, 20, 86]])$

Indexing and slicing -

91.) $a = \text{np.ones}((2, 2)) + 1$
 a $\rightarrow \text{array}([[2., 2.]$
 $[2., 2.]])$ 92.) to get
normal distribution
of $\mu = 100, \sigma = 15$
 \Rightarrow import numpy as np
import matplotlib.pyplot as plt
 $\mu, \sigma = 100, 15$
 $x = \mu + \sigma * \text{np.random. standard}(1000)$ \rightarrow

Indexing and slicing-

Indexing

93) $a = \text{np. arange}(10)$ // print element at 5th position.
 print ($a[5]$)

$\rightarrow 5$

$\begin{matrix} & (0, 1, 2, 3, 4, 5, 6, 7, 8, 9) \\ & \uparrow \uparrow 1 \\ \text{address} & 1 \quad 2 \quad - \quad - \quad - \quad 5 \end{matrix}$

94.) $a = \text{np. diag} [(1, 2, 3)]$

print (a)

print ($a[2, 2]$)

$\rightarrow a[2][1 \quad 0 \quad 0]$

$\begin{matrix} 1 & [0 \quad 2 \quad 0] \end{matrix}$

$\begin{matrix} 2 & [0 \quad 0 \quad 3] \end{matrix}$

$\begin{matrix} 3 & \overline{[0 \quad 1 \quad 2]} \end{matrix}$

95.) Assign any value -

$a[2, 1] = 5$

a

$\rightarrow \text{array} ([[1, 0, 0]]$

$[0, 2, 0]$

$[0, 5, 3]]$.

96.) slice out value from 1-D array.

$a[1:8:2]$

$\rightarrow \text{array} ([1, 3, 5, 7])$

97.) # we can also combine assignment & slicing:

$a = np.arange(10)$

$a[5:] = 10$

a

$\rightarrow \text{array}([0, 1, 2, 3, 4, 10, 10, 10, 10, 10])$

98.) $b = np.arange(5)$

$a[5:] = b[:-1]$ # assigning

a

$\rightarrow \text{array}([0, 1, 2, 3, 4, 4, 3, 2, 1, 0])$

Copies & views -

99.) $a = np.arange(10)$

$\rightarrow \text{array}([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])$

$b = a[::2] \rightarrow \underline{\text{view}}$

b

$\rightarrow \text{array}([0, 2, 4, 6, 8])$

$np.shares_memory(a, b)$

\rightarrow True

{# are the two arrays

share memory}, i.e., they

have same values }

$b[0] = 10$

b

$\rightarrow \text{array}([10, 2, 4, 6, 8])$

a

$\rightarrow ([10, 1, 2, 3, 4, 5, 6, 7, 8, 9])$

it change the value in a
also it change in b.

100.) To overcome problem of pt (99) sleepy as created -

$$a = np \cdot \text{arrange}(5)$$

c = a[::2].copy()

\rightarrow array ([0, 2, 4])

np·shares_memory(a,c)

→ False

now it does not share memory as 'c' is a new copy of 'a'. there will be no self-blw a &c.

$$c [O] = 10$$

a

\rightarrow array ([0,1,2,3,4])

Fancy Indexing - It by default creates copy.

101.) using Boolean mask -

```
a = np.random.randint(0, 20, 15)
```

a

array ([18, 17, 1, 18, 5, 17, 0, 14, 12, 11, 4, 15, 16, 8, 7])

102.) mask = (a % 2 == 0)

mask

\rightarrow array ([True, False, T, F, T, -T, F, T, T, F, T, F, T, T, T, f])

103.) extract - from $a = a[mask]$ # we get value
 extract - from a .
 \rightarrow array ([18, 18, 0, 14, 12, 4, 16, 8]) where the
 "cond" is True.

104.) $a[mask] = -1$

a

\rightarrow array [-1, 17, 1, -1, 5, 17, -1, -1, -1, 11, -1, 15, -1, -1, 7])
 # every even element get the value -1.

105.) $s = np.array ([[1, 2, 3], [4, 5, 6]])$.

$s.flatten()$

\rightarrow array [1, 2, 3, 4, 5, 6])

convert a 2-D array into 1-D array.

106.) import numpy as np

$s = np.array ([[1, 2, 3], [4, 5, 6], [2, 4, 5], [1, 4, 2]])$

$s.shape$

$s.reshape (2, 2, 3)$ 3rd dim.

\rightarrow array ([[1, 2, 3],] 2nd dim.
 [4, 5, 6],] 3rd dim.)

* these two in 1st dimension

[[2, 4, 5],] 2nd dim.
 [1, 4, 2]]] 3rd dim.

107.) in above eg -

s. reshape (2, 2, 3, 1)

→ array ([[[1],
[2],
[3],

[4],
[5],
[6]]],

[[[2],
[4],
[5]]]]

[[[1],
[4],
[2]]]]]) .

108.) in pt. (106)

s. reshape (1, 2, 2, 3)

→ array ([[[1, 2, 3],
[4, 5, 6]]])

[[[2, 4, 5],

[1, 4, 2]]]])

109.) in pt (106)

s. reshape (2, 3, 2)

→ array ([[[1, 2],

[3, 4],
[5, 6]]],

[[[2, 4],
[5, 1],
[4, 2]]]]) .

109) - 110.) s. ravel ()

→ array ([1, 2, 3, 4, 5, 6]).

convert n-D array into 1-D array.

111.) → flatten always returns a copy.

Different → Ravel returns a view of the original array whenever possible. In this we don't know it return a "copy" or "view".

Element wise operation-

- if we add any no. then it will add its whole array
- if we multiply any no. it multiply with whole array
- square → all.
- arithmetic operation operate elementwise

112.) matrix multiplication -

`c = np.diag([1, 2, 3, 4])`

~~print(c * c)~~

`print(c * c)` → element wise multiplication.

`print(np.matmul(c, c))` → matrix multiply as

→ $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix}$ } dot multiplication we done in maths.

$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix}$ } maths matrix multiplication.

Comparisons - (element wise)

113.) `a = np.array([1, 2, 3])`

`b = np.array([5, 2, 2])`

→ `array([False, True, False], dtype=bool)`

114.) $a > b$

→ array ([False, False, True], dtype = bool)

115.) # element wise comparisons -

a = np.array ([1, 2, 3, 4])

b = np.array ([5, 2, 2, 4])

c = np.array ([1, 2, 3, 4])

np.array_equal(a, b)

→ False

np.array_equal(a, c)

→ True

Logical Operations -

116.) a = np.array ([1, 1, 0, 0], dtype = bool)

b = np.array ([1, 0, 1, 0], dtype = bool)

print (a)

np.logical_or (a, b)

0 → False

→ [True, True, False, False]

1 → True

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

117.) np.logical_and (a, b)

→ array ([True, False, False, False], dtype = bool).

Transcendental function:

118.) $a = \text{np.array}(4)$ # find value of sine.
 $\text{np.sin}(a)$
→ array ([0., 0.84147098, 0.90929743, 0.1411200])

119.) $\text{np.log}(a)$
→ array [-inf, 0., 0.69314718, 0.09861229]

120.) $\text{np.exp}(a)$
→ array [1., 2.71828183, 7.3890561, 20.08553692])

121.) Shape mismatch.

$a = \text{np.array}(4)$
 $a + \text{np.array}([1, 2])$
→ broadcasting error.
arrays with different shape cannot be add.

Basic Reduction -

122.) $x = \text{np.array}([1, 2, 3, 4])$
 $\text{np.sum}(x)$
→ 10

123.) # sum by rows & by columns.

$x = \text{np.array}([[1, 1], [2, 2]])$
 $\rightarrow \text{array}([[1, 1],$
 $[2, 2]])$.

$x.sum(axis=0)$ # columns set dim.

$\rightarrow \text{array}([3, 3])$

$x.sum(axis=1)$ # rows (2nd dim.)

$\rightarrow \text{array}([2, 4])$

Other reductions -

0 1 2 ← index
 ↑ ↓ ↓

124.) $x = \text{np.array}([1, 3, 2])$

$x.min()$
 $\rightarrow 1$

$x.argmax()$ # index of
 $\rightarrow 1$ min. element

$x.max()$

$\rightarrow 3$

$x.argmax()$ # index of
 $\rightarrow 0$ max. element

Logical operations :

125.) $\text{np.all}([True, True, False])$ { If any one is
 $\rightarrow \text{False}$ false it gives false }

126.) $\text{np.any}([True, False, False])$ { If any element is
 $\rightarrow \text{True}$ true it gives true }

127.) $\text{np.any}(a != 6)$
→ False.

can be used for array comparison.

128.) $\text{np.all}(a == a)$
→ True.

{ lets say -
 $a = [1, 2, 3, 4]$ }.

129.) $a = \text{np.array}([1, 2, 3, 2])$
 $b = \text{np.array}([2, 2, 3, 2])$
 $c = \text{np.array}([6, 4, 4, 5])$
 $((a \leq b) \& (b \leq c)).\text{all}()$
→ True.

statistics -

130.) $x = \text{np.array}([1, 2, 3, 1])$
 $y = \text{np.array}([[1, 2, 3], [5, 6, 1]])$
 $x.\text{mean}()$ | $\text{np.meanx}()$
→ 1.75

131.) $\text{np.medianx}()$
→ 1.75

132.) $\text{np.median}(y, \text{axis}=1)$ # last axis
→ $\text{array}([2., 5.])$ {
 $\text{axis} = 1$
 $y = \begin{bmatrix} 1 & 2 & 3 \\ 5 & 6 & 1 \end{bmatrix}$
 ↓ ↓ ↓ ↓
 axis 0 → mean -}

133.) $\text{np.median}(y, \text{axis}=0)$
→ $\text{array}([3., 4., 2.])$

134.) $x = \text{std}(l)$ $\rightarrow 0.8291561975884995$

full population std. dev.

135.) # load data into numpy array object
data = np.loadtxt('filename.txt')

Data:

 \rightarrow data of file.

	year	Rabbit	Cat	Carrot
1900		30000	4000	48300
1901		47200	6100	50000
1902		70200	5000	51000
1903		77400	6000	38200

array([[1900., 30000., 4000., 48300.],
 [1901., 47200., 6100., 50000.],
 [1902., 70200., 5000., 51000.],
 [1903., 77400., 6000., 38200.]]).

136.) rabbit, year, carrots, cat, carrots = data.T
 \rightarrow print(year) \rightarrow transpose
 \rightarrow [1900, 1901, 1902, 1903] # columns
 print(cat) \rightarrow data to variables.
 \rightarrow [4000, 6100, 5000, 51000]

print(carrots)

 \rightarrow [48300, 50000, 51000, 38200]

print(carrots)

 \rightarrow [48300, 50000, 51000, 38200]

137.) # The mean population over time

populations = data[:, 1:] \rightarrow take all rows
 populations \rightarrow , start from 2nd leave 0 under \rightarrow array([[30000., 4000., 48300.]])

[47200., 6100., 50000.]])

[70200., 5000., 51000.]])

[77400., 6000., 38200.]])

138.) #= sample std. dev.
populations. std (axis=0)
→ array([

139.) # which species has the highest population
each yr?
np.argmax(populations, axis=1) (horizontal)

→ array([2, 2, 0, 0])

140.) np.argmax(populations, axis=0)

→ array([3, 1, 2]) (vertically max. index?)

Broadcasting -

rule -

Two dimensions are compatible when-

- 1) They are equal, or (e.g. $4 \times 3 - 4 \times 3, 2 \times 4, 2 \times 4$).
- 2) one of them is ~~one~~ 1. (i.e., either row or column is 1 e.g. $3 \times 1, 1 \times 3, 4 \times 1, 1 \times 4$ etc.).

141.) a = np.tile(np.arange(0, 40, 10), (3, 1))
print(a)

$$a = a.T$$

print(a)

→ [[0 10 20 30] {3 rows}	[0 0 0]
[0 10 20 30] {4 rows}	[10 10 10]
[0 10 20 30]	[20 20 20]
1 column	[30 30 30]

142.) $b = \text{np.array}([10, 1, 2])$

b

$\rightarrow \text{array}([0, 1, 2])$

143.) $a+b$

$\rightarrow \text{array}([[0, 1, 2],$
 $[10, 11, 12],$
 $[20, 21, 22],$
 $[30, 31, 32]])$

144.) $a = \text{np.arange}(10, 40, 10)$

$a.$ shape

$\rightarrow \text{array}([0, 10, 20, 30])$

145.) $\#$ add a new axis \rightarrow 2D array.

$a = a[:, \text{np.newaxis}]$

in then

1D \rightarrow 2D

2D \rightarrow 3D

3D \rightarrow 4D
and so on.

$a.$ shape

$\rightarrow (4, 1)$

a

$\rightarrow ([0],$
 $[10],$
 $[20],$
 $[30]])$

146.) $a+b$

$\rightarrow \text{array}([[0, 1, 2],$
 $[10, 11, 12],$
 $[20, 21, 22],$
 $[30, 31, 32]])$

Array Shape Manipulation.

Flattening-

147.) $a = \text{np.array}([[1, 2, 3], [4, 5, 6]])$

$a.\text{ravel}()$ # Return a contiguous flattened array. A 1-D array, containing
 $\rightarrow \text{array}[1, 2, 3, 4, 5, 6]$ the elements of a .

148.) $a.T$ # Transpose

$\rightarrow \text{array}([[1, 4], [2, 5], [3, 6]])$

149.) $a.T.\text{ravel}()$

$\rightarrow \text{array}[1, 4, 2, 5, 3, 6]$

150.) Reshape (Done previously).

\rightarrow inverse ^{opn.} of flattening.

\rightarrow it may also return a copy.

Dimension shuffling -

$a = \text{np.arange}(4 * 3 * 2).\text{reshape}(4, 3, 2)$.

$a.\text{shape}$

$\rightarrow (4, 3, 2)$

a

$\rightarrow \text{array}([[[0, 1], [2, 3], [4, 5]],$

$[6, 7], [8, 9], [10, 11], [12, 13], [14, 15], [16, 17]]])$

$[[6, 7],$

$[8, 9],$

$[10, 11],$

$[12, 13],$

$[14, 15],$

$[18, 19],$

$[20, 21],$

$[22, 23]]])$

152.) $a[0, 2, 1]$
→ 5

153.) $a[[0, 2, 1]]$

→ array([0, 0, 0],
[20, 20, 20],

[10, 10, 10])

154.) Resizing -

$a = np.array(4)$

$a.resize(8, 0))$

a

→ array([0, 1, 2, 3, 0, 0, 0, 0])

155.) $b = a$

$a.resize((4,))$

→ Error (as b also refer to a so, when it a
resize b also change).

Sorting -

156.) # sorting along an axis:

$a = np.array([[5, 4, 6], [2, 3, 2]])$

$b = np.sort(a, axis=1)$

b

→ array([[4, 5, 6],
[2, 3, 2]])

157.) # in-place sort

a. $sort(axis=1)$

→ array([[4 5 6]
[2 3 2]])

158.) # sorting with fancy indexing -

a = np.array ([4, 3, 1, 2])

j = np.argsort(a)

j → array ([4, 5, 6], ([2, 3, 1, 0]). sorted.
[2, 2, 3])

index are ↗

159.) a[j]

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

~~PANDAS~~ (without pandas) ↴

160.) ~~parsed except~~

⇒ CSV → a module in python.

⇒ open → a "f" to open ^{type of} array file.

Eg. with open (-file-path, "r") as f:

f = open(-file-path...) → after this we have to close this file

⇒ after ":" a file block is started

⇒ reader.next() { change reader pointer to next row
move to next line.}

⇒ append → to add anything.

⇒ if __name__ == '__main__':

↳ to create main "f" in python.

PANDAS: Data frame: it is main obj. in pandas used to represent data with rows & columns.

Creating Dataframes -

160.) import pandas as pd

df = pd.read_csv("weather-data.csv") # read weather.csv data

df

	day	temp	windspeed	event
0	1/1/2017	32	6	Rain
1	1/2/2017	35	7	Sunny
2	1/3/2017	28	2	snow
3	1/4/2017	24	7	snow
4	1/5/2017	32	4	Rain
5	1/6/2017	31	2	Sunny

(file already created)

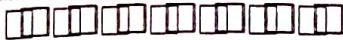
161.) # list of tuples [make list by ourself]

weather_data = [('1/1/2017', 32, 6, 'Rain'),
 ('1/2/2017', 35, 7, 'Sunny'),
 ('1/3/2017', 28, 2, 'snow')]

df = pd.DataFrame(weather_data, columns=['day',
 'temperature', 'windspeed', 'event'])

df

	day	temp.	windspeed	event
0	1/1/2017	32	6	Rain
1	1/2/2017	35	7	Sunny
2	1/3/2017	28	2	snow



16.2.) `# get dimensions of the table.`
`df.shape` # total no. of rows & columns.
 $\rightarrow (3, 4)$

16.3.) `# if you want to see initial some rows then use head command (default : 5 rows)`
`df.head()` / `df.head(1)`
 \rightarrow same table till 5 rows. \hookrightarrow only one row.

16.4.) `# if you want to see last few rows then use tail command (default : 5 rows).`
`df.tail()` / `df.tail(2)` \rightarrow last 2 rows.
 \rightarrow above table till 5 last 5 rows

16.5.) `# slicing`
`df[2:5]`
 \hookrightarrow start from 2nd & end to 5

16.6.) `df.columns` # print columns in a table
 \rightarrow `Index(['day', 'temp', 'windspeed', 'event'],`
`dtype='object')` {# print headings}

16.7.) `df.day` # print particular column data
 \rightarrow 0 1/1/2017
 1 1/2/2017
 2 1/3/2017

Name : day , dtype : object

168.) # another way of accessing column.
df['day'] # df.day (both are same)
→ same op as (167)

169.) # get 2 or more column.

df[['day', 'event']]

→

	day	event
0	1/1/2017	Rain
1	1/2/2017	sunny
2	1/3/2017	snow

max or min.

170.) # get all temp.

df['temperature'].max() / .min()

→ 35 / 28

171.) df['temperature'].describe()

→ count

mean

std

min

25%

50%

75%

max

Name: temp , dtype: float64

172.) # select rows which has max temp.

`df[df['temp'] == df['temperature'].max()]`

	day	temp.	ws	event
1	1/2/2017	35	7	Sunny

173.) # select rows which has max temp.

`df[df['temp'] < df['temp'].max()]`

	day	temp.	ws	event
0	1/1/2017	32	6	Rain
2	1/3/2017	28	2	snow

174.) Read & write XLS file.

Excel

Install : pip³ install xlrd

read excel file a library is required.

`df = pd.read_excel('weather_data.xlsx')`

`df`

→ excel file will open.

175.) # write DF to CSV → save to csv.

`df.to_csv('new.csv')`

→ w/ file with index (0, 1, 2...)

`df.to_csv('new_no_index.csv', index=False)`

→ file w/o index.

with index

	day	temp.	ws	event
0				
1				
2				

w/o index

	day	temp.	ws	event
0				
1				
2				

176.) # Install: pip3 install xlrd openpyxl
write DF to Excel { save dataframe
df.to_excel('new.xlsx',
sheet_name='weather_data').
to excel }

Group-by → a table divide in diff. parts
acc. to columns.

177.) import pandas as pd.
df = pd.read_csv('weather_data_cities.csv')
df # weather by cities.

	day	city	temp.	ws	event
→ 0	1/1/2017	newyork	32	6	Rain
1	1/2/2017	mumbai	87	15	Fog
2	1/3/2017	paris	50	20	Sunny

178.) ⚡ g = df.groupby('city')
→ {
g
<pandas.core.groupby.DataFrameGroupBy object at 0x13>

179.) for city, city-DF using:
print(city)
print(city-DF)
→ obj will try be data acc. to cities
mumbai data
N.Y. data
... data

180) # ou to get specific group:

g.get_group('mumbai')

→ only get data of mumbai.

181.) # find max. Temp. in each cities

print(g.max())

→ get max temp. of all cities.

182.) # print (g.mean())

→ get mean of temp & ws of all cities.

183.) print (g.describe())

→ get the mean, std., min., 25%, 50%, 75%, max.
(acc. to temp & ws).

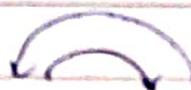
Concatenate Data frames-

184.) import pandas as pd

india_weather = pd.DataFrame({

after
space } "city": ["mumbai", "delhi", "pune"],
"temp": [32, 45, 30]
"humidity": [80, 60, 78]
})

india_weather



	city	temp	humidity
0	mumbai	32	80
1	delhi	45	60
2	pune	30	78