

MSC IN FINANCE

PRE-TERM COURSE WEEK2:

Object-Oriented Programming, Numerical Computing with NumPy

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Agenda:

1. Object-Oriented Programming (OOP)

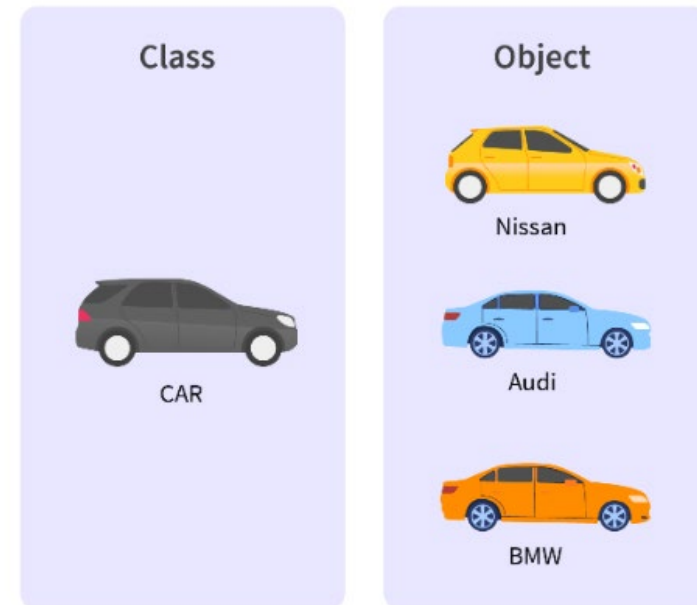
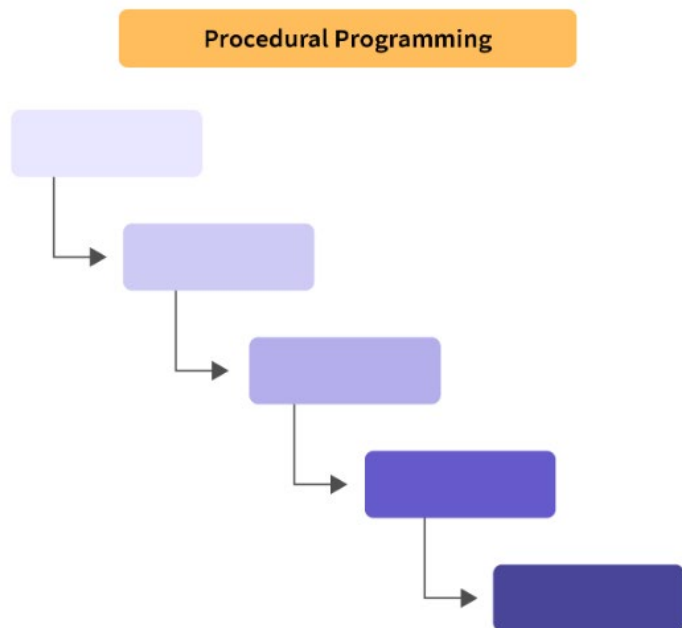
- Class and Object
- Class Attributes and Methods

2. NumPy

- ndarray
- Numerical Computing with NumPy

Object-Oriented Programming (OOP)

- Two basic programming paradigms:
 - Procedural
 - Organizing programs around functions or blocks of statements which manipulate data.
 - Object-Oriented
 - Structuring a program by bundling related properties and behaviors into individual objects.

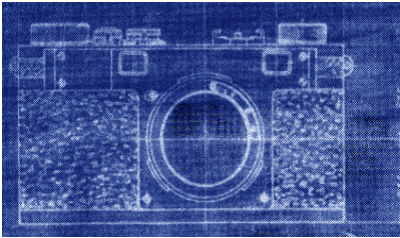


Object-Oriented Programming(Cont.)

- **Classes** and **objects** are the two main aspects of object-oriented programming.
- A **class** creates a new *type*.
- An **object** is the *instance* of the class.
- Everything in Python is really an object.
- E.g.: An analogy is that we can have variables of type *int* which translates to saying that variables that store integers are variables which are instances (objects) of the *int* class.

Class and Object

- Defining a class
 - A template that describes that class: how many fields, what type of information will be stored by each field, what default information will be stored in a field.
- Creating objects
 - Instances of that class (during instantiation) which can take on different forms.



Class: Define A Composite Type

- Classes can be used to define a generic template for a new non-homogeneous composite type.
- The class definition specifies the type of information (called “**attributes**”) that each instance (example) tracks.



Name:
Phone:
Email:
Purchases:



Name:
Phone:
Email:
Purchases:



Name:
Phone:
Email:
Purchases:

Define a Class

- **Format:**

```
class <Name of the class>:  
    name of first field = <default value>  
    name of second field = <default value>
```

Note the convention: The first letter is capitalized.

- **Example:**

```
class Client:  
    name = "default"  
    phone = "(123)456-7890"  
    email = "foo@bar.com"  
    purchases = 0
```

Describes what information that would be tracked by a “Client” but doesn’t actually create a client variable

Contrast this with a list definition of a client

```
client = ["xxxxxxxxxxxxxxxxxx",  
          "0000000000",  
          "xxxxxxxxxx",  
          0]
```

Create an Instance (Object) of a Class

- Creating an actual instance (instance = object) is referred to as

- **Format:**

```
<reference name> = <name of class>()      # Create a new object  
<reference name>.<field name>              # Accessing value  
<reference name>.<field name> = <value>    # Changing value
```

- **Example:**

```
aClient = Client()  
aClient.name = "Lily"  
aClient.email = "python@cuhk.edu.hk"
```


Example: The Client Class

```
class Client:
    name = "default"
    phone = "(123)456-7890"
    email = "foo@bar.com"
    purchases = 0
```

```
firstClient = Client()
firstClient.name = "Lily"
firstClient.email = "python@cuhk.edu.hk"
print(firstClient.name)
print(firstClient.phone)
print(firstClient.email)
print(firstClient.purchases)
```



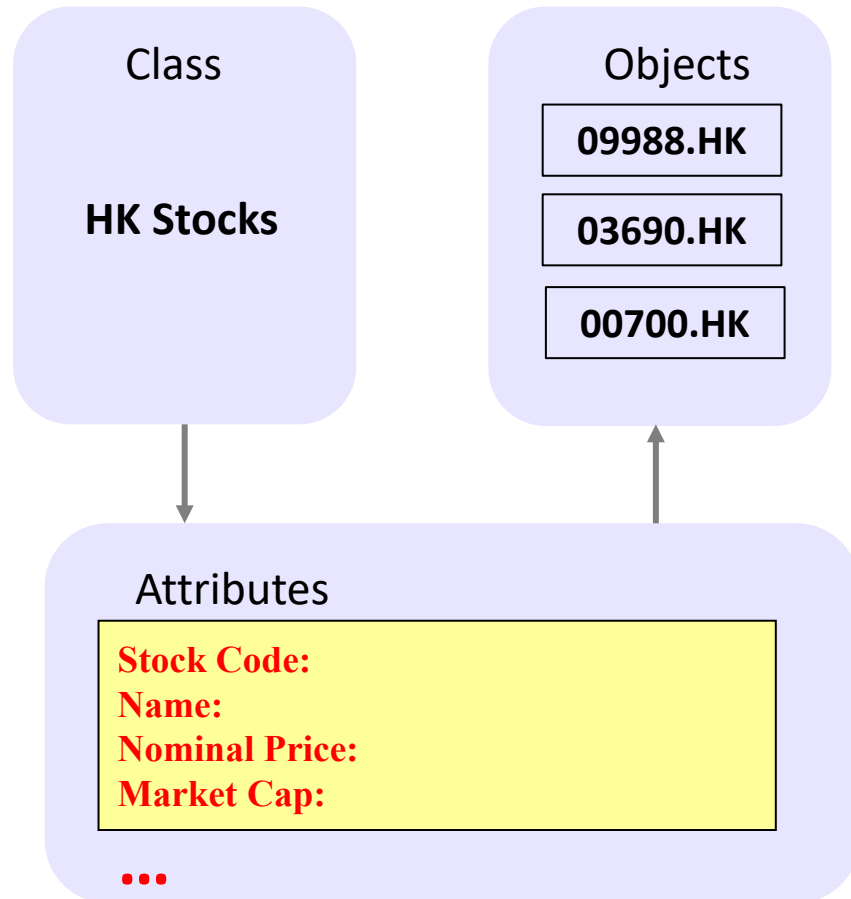
```
name = "default"
phone = "(123)456-7890"
email = "foo@bar.com"
purchases = 0
```

```
name = "Lily"
email = "python@cuhk.edu.hk"
```

```
Lily
(123)456-7890
python@cuhk.edu.hk
0
```

Example: HK Stocks

```
class HKStocks:
    def __init__(self, stock_code = "default", name = "NA", price = 0, market_cap = 0 ):
        self.profile = {'stock_code': stock_code, 'name': name, 'price': price, 'market_cap': market_cap}
```



```
# object 1 of Class-HKStocks
s1 = HKStocks('09988.HK', 'BABA-SW', 89.55, 1897.12)
print(s1.profile)
print(s1.profile['stock_code'])
print(s1.profile['name'])
print(s1.profile['price'])
print(s1.profile['market_cap'])
```

```
{'stock_code': '09988.HK', 'name': 'BABA-SW', 'price': 89.55, 'market_cap': 1897.12}
09988.HK
BABA-SW
89.55
1897.12
```

```
# object 3 of Class-HKStocks
s3 = HKStocks('00700.HK', 'TENCENT', 325.00, 3111.63)
print(s3.profile)
print(s3.profile['stock_code'])
print(s3.profile['name'])
print(s3.profile['price'])
print(s3.profile['market_cap'])
```

```
{'stock_code': '00700.HK', 'name': 'TENCENT', 'price': 325.0, 'market_cap': 3111.63}
00700.HK
TENCENT
325.0
3111.63
```

The Benefits of Defining a Class

- It allows new types of variables to be declared. The new type can model information about most any arbitrary entity:
 - Car, Movie, A bacteria or virus in a medical simulation, An 'object' (e.g., sword, ray gun, food, treasure) in a video game.
 - A member of a website (e.g., a social network user could have attributes to specify the person's: images, videos, links, comments and other posts associated with the 'profile' object).
 - Firms (e.g. firms with different characteristics and also different investment choices)
 - Snapshots of order books (e.g. at each timestamp, the order book for each stock records information of price and volume and they will be changed by the incoming orders)
- Unlike creating a composite type by using a list, a predetermined number of fields can be specified, and those fields can be named.

class Client:

name = "default"

phone = "(123)456-7890"

email = "foo@bar.com"

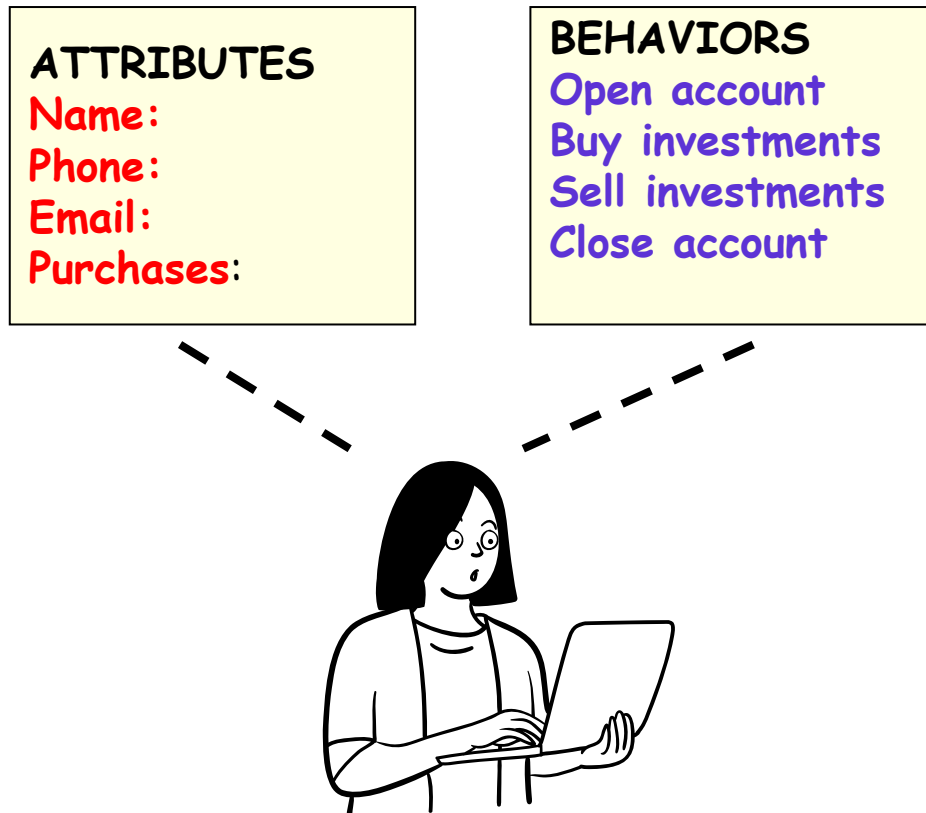
purchases = 0

firstClient = Client ()

print(firstClient.middleName) **# Error: no such field defined**

Class Methods (“Behaviors”)

Classes Have **Attributes**, But Also **Behaviors**



Functions: not tied to a composite type or object.

The call is ‘stand alone’, just name of function

E.g.,

print(), **input()**

Methods: must be called through an instance of a composite.

E.g.,

filename = "foo.txt"

name, suffix = filename.**split('.')**

Functions that are associated with classes are referred to as **methods**.

Define Class Methods

- **Format:**

```
class <classname>:  
    def <method name> (self <other parameters>):  
        <method body>
```

- **Example:**

```
class Person:  
    name = "I have no name :("  
    def sayName (self):  
        print ("My name is...", self.name)
```

Unlike functions, every method of a class must have the 'self' parameter (more on this later)

When the attributes are accessed inside the methods of a class they **MUST** be preceded by the prefix "self."

Example: Defining Class Methods

```
class Person:  
    name = "I have no name :("  
    def sayName(self):  
        print("My name is...", self.name)
```

```
aPerson = Person()  
aPerson.sayName()
```

```
My name is... I have no name :(
```

```
aPerson.name = "HappyLily :D"  
aPerson.sayName()
```

```
My name is... HappyLily :D
```

Recap: Accessing Attributes & Methods

- **Inside the class definition** (inside the body of the class methods)

- Prefix the attribute or method using the '**self**' reference

```
class Person:
```

```
    name = " I have no name :( "
```

```
    def sayName(self):
```

```
        print("My name is...", self.name)
```

- **Outside the class definition**

- Prefix the attribute or method using the **name of the reference** used when creating the object.

```
lisa = Person()
```

```
lisa.name = "Lisa, Nice to meet you."
```

```
lisa.sayName()
```

Constructor: A Special Method

- Classes have a special method ‘`__init__()`’ that can be used to initialize the starting values of a class to some specific values.
- This method is automatically called whenever an object is created.

- **Format:**

```
class <Class name>:  
    def __init__(self, <other parameters>):  
        <body of the method>
```

- **Example:**

```
class Person:  
    name = ""  
    def __init__(self):  
        self.name = "No name"
```

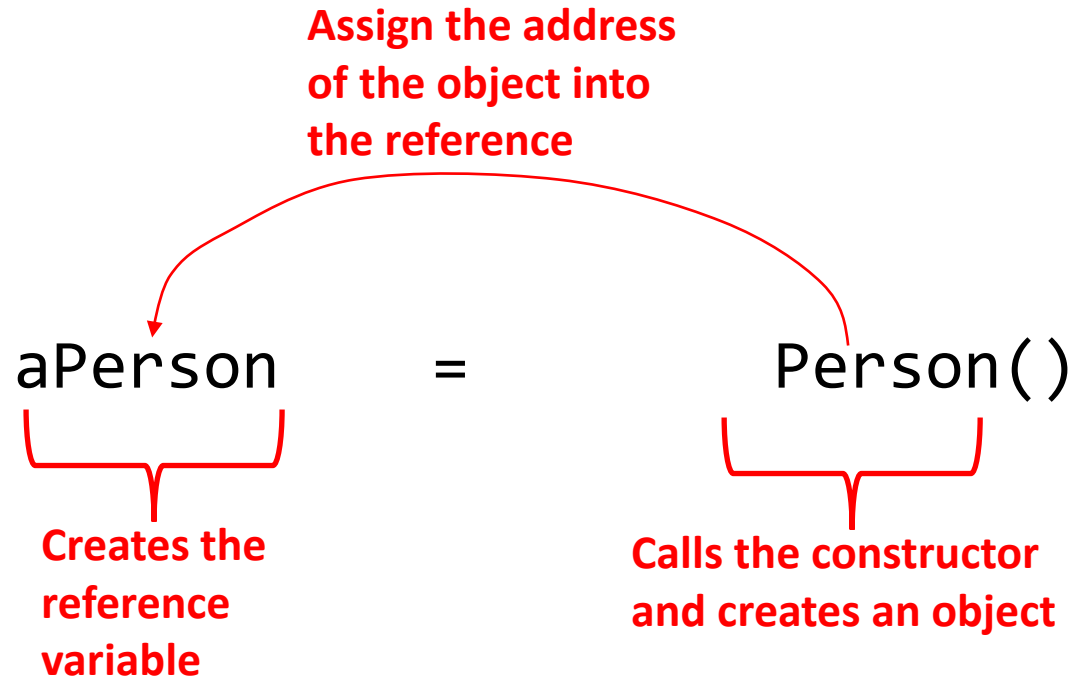
This design approach is consistent with many languages



```
bPerson = Person()  
bPerson.name
```

'No name'

Example: Using The “__Init__()” Method



```
class Person:  
    name = ""  
  
    def __init__(self, aName):  
        self.name = aName
```

```
cPerson = Person("Lynn")  
cPerson.name
```

'Lynn'

Example: Using The “__Init__()” Method (Contd.)

- Similar to other methods, ‘init’ can be defined so that if parameters aren’t passed into them then default values can be assigned.

- **Example:**

```
def __init__ (self, aName="No name"):  
    self.name = aName
```

This method can be called either when a personalized name is given or if the name is left out.

- Method calls (to ‘init’), both will work

```
dPerson = Person()  
ePerson = Person("Rose")
```

```
dPerson = Person()  
print(dPerson.name)  
  
ePerson = Person("Rose")  
print(ePerson.name)
```

No name
Rose

Complete Example: Class Person with Birthday

```
class Person:
    name = "No Name"
    age = 0

    def __init__(self, newName, newAge):
        self.name = newName
        self.age = newAge

    def haveBirthday(self):
        print("Happy Birthday!")
        self.mature()

    def mature(self):
        self.age = self.age + 1
```

```
def __init__(self, newName, newAge):
    self.name = newName
    self.age = newAge

aPerson = Person("Cartman", 8)
print("%s is %d." % (aPerson.name, aPerson.age))
aPerson.haveBirthday()
print("%s is %d." % (aPerson.name, aPerson.age))

def haveBirthday(self):
    print("Happy Birthday!")
    self.mature()

def mature(self):
    self.age = self.age + 1
```

```
aPerson = Person("Cartman", 8)
print("%s is %d." % (aPerson.name, aPerson.age))
```

Cartman is 8.

```
aPerson.haveBirthday()
print("%s is %d." % (aPerson.name, aPerson.age))
```

Happy Birthday!
Cartman is 9.

Complete Example: Class Client with Purchases

```
class Client:
    name = "No Name"
    purchase = 0

    def __init__(self, newName, newPurchase):
        self.name = newName
        self.purchase = newPurchase

    def makePurchase(self, addPurchase):
        print("Making new purchase "+ str(addPurchase))
        self.purchase = self.purchase + addPurchase
```

```
aClient = Client("Lily",100)
print("%s's purchase position: %d." %(aClient.name,aClient.purchase))
```

Lily's purchase position: 100.

```
aClient.makePurchase(150)
print("%s's purchase position: %d." %(aClient.name,aClient.purchase))
```

Making new purchase150
Lily's purchase position: 250.

Numpy

1. Why NumPy?
2. ndarray
3. Copies vs Views
4. Broadcasting
5. Iterations

Why NumPy?

- ndarray stands for N-dimensional array.
- A NumPy array is a grid of values, all of the same data type.
- This simplifies the data storage process and makes numerical operations faster.
- NumPy arrays form the core of nearly the entire ecosystem of data science tools in Python, such as
 - Pandas,
 - Scikit Learn
 - and more ...



Why NumPy: Compared with Excel in Matrix Operations

Matrices multiplication in Excel

1	4
2	3
1	4

1	2
1	2

=MMULT(K21:L23,O21:P22)

- Three steps:
- Select the input and output areas
- array formula =MMULT(XX:XX,XX:XX)
- ctrl+shift+enter

Matrices multiplication in NumPy

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


array([[ 5, 10],
       [ 5, 10],
       [ 5, 10]])
```

- Only need one line:
- a.dot(b)

ndarray: import

- We begin using modules by importing them

import directive
is used for
importing
modules and its
methods to the
working
environment

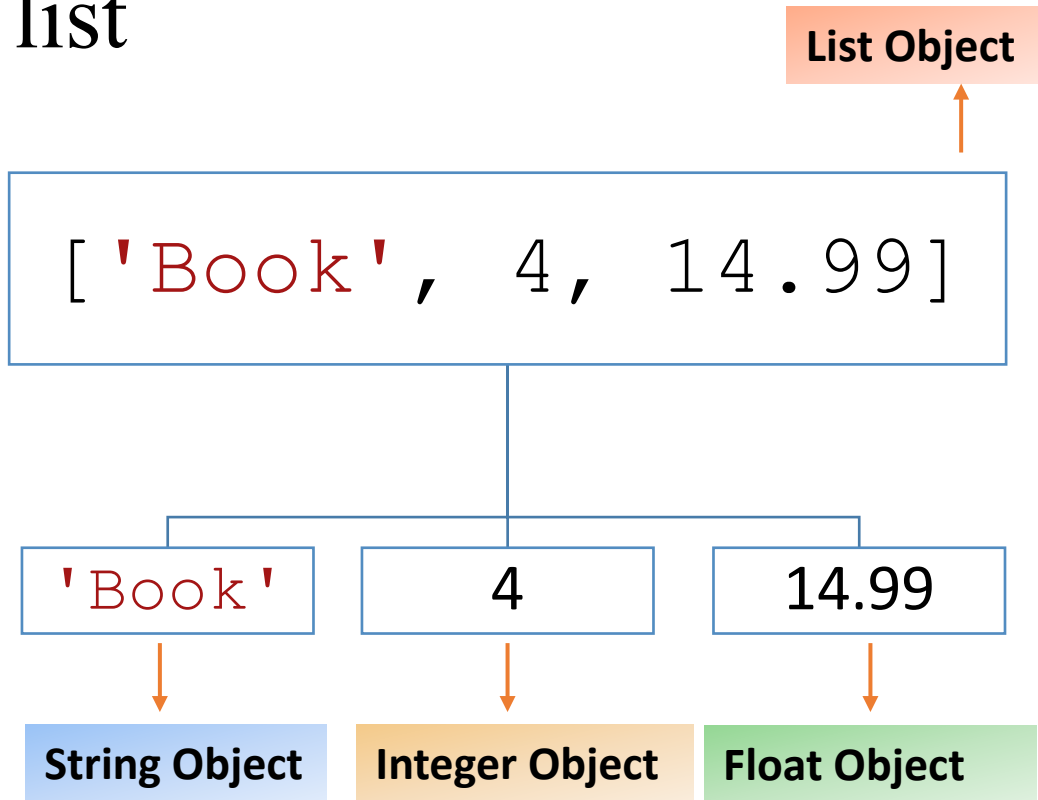


```
import numpy as np
```

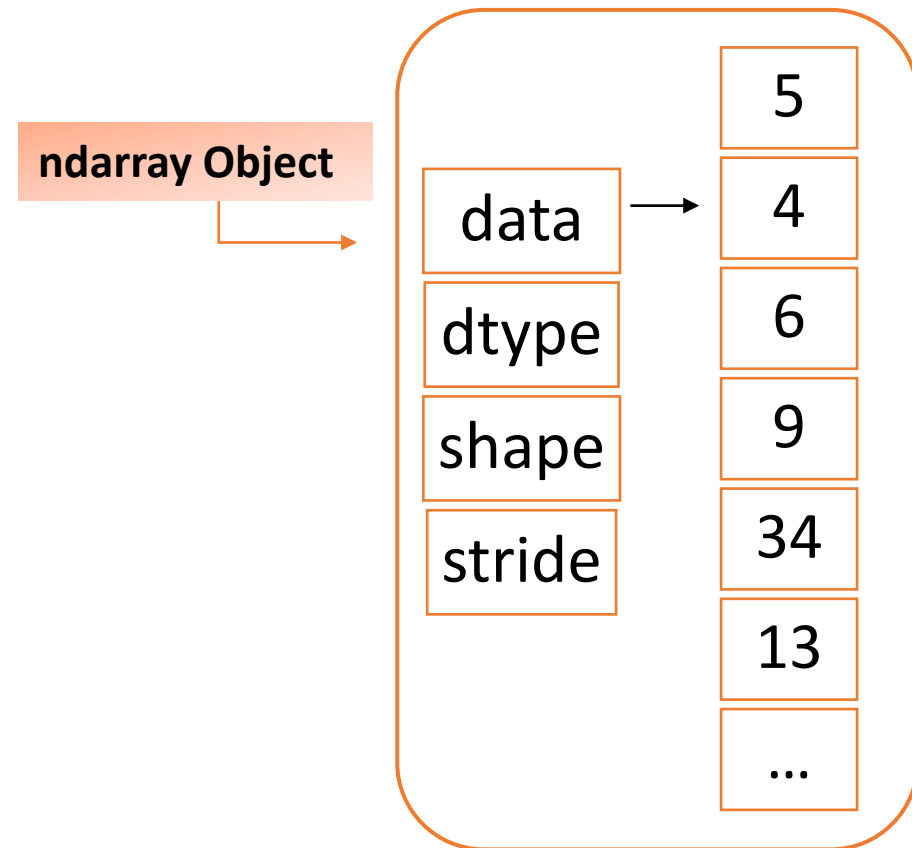
an alias such as `np`
can be given to avoid having
to type the name of the
module every time we want to
call a method available in it

List vs ndarray data structure

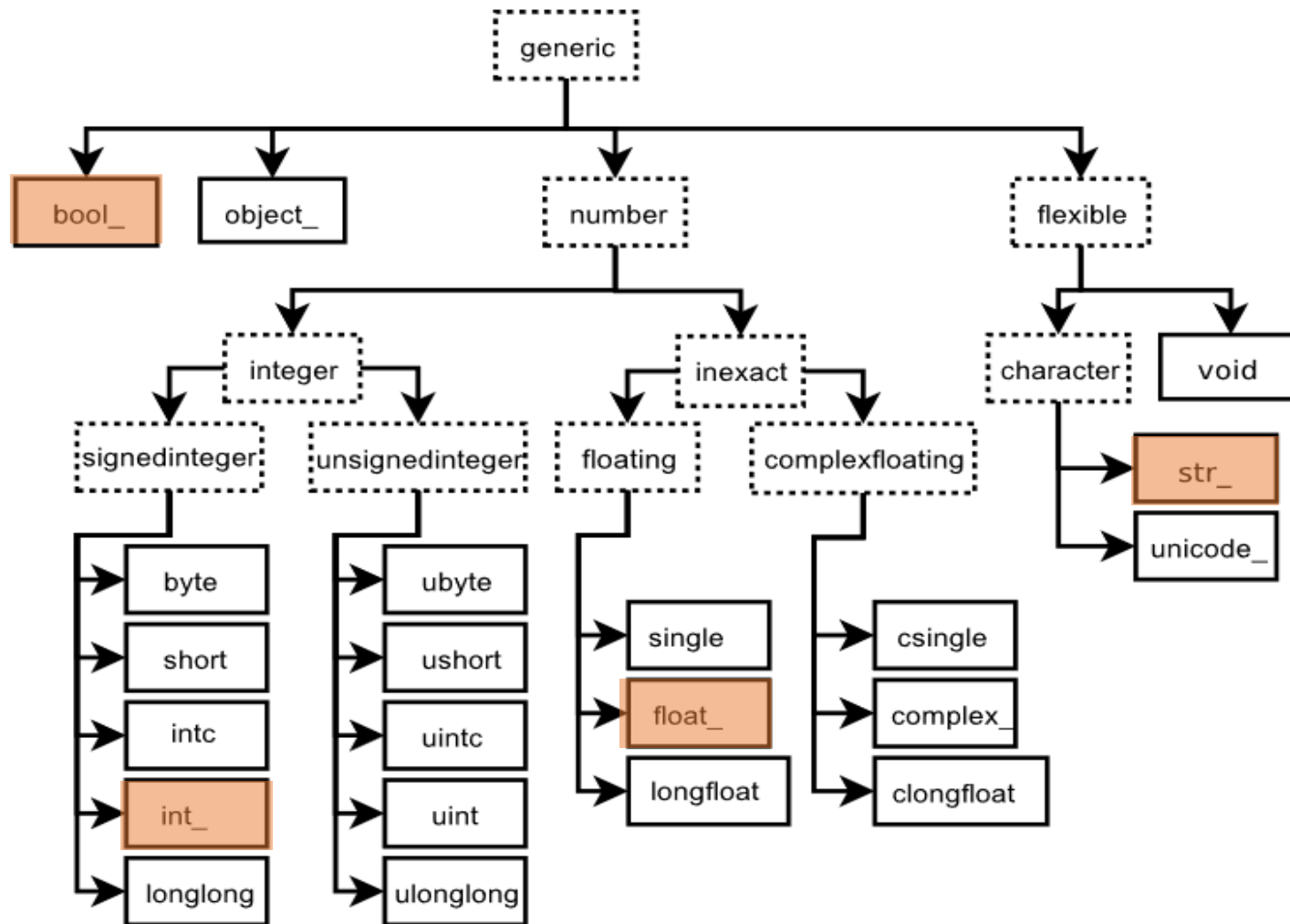
list



ndarray



ndarray: dtype



and there are more data types supported by NumPy

such as datetime objects

ndarray vs list operation

add 2 to **each item** of -

➤ a list

```
numlist = [1, 2, 3]
```

```
doubled_list = []  
for i in numlist:  
    doubled_list.append(i+2)
```

➤ an array

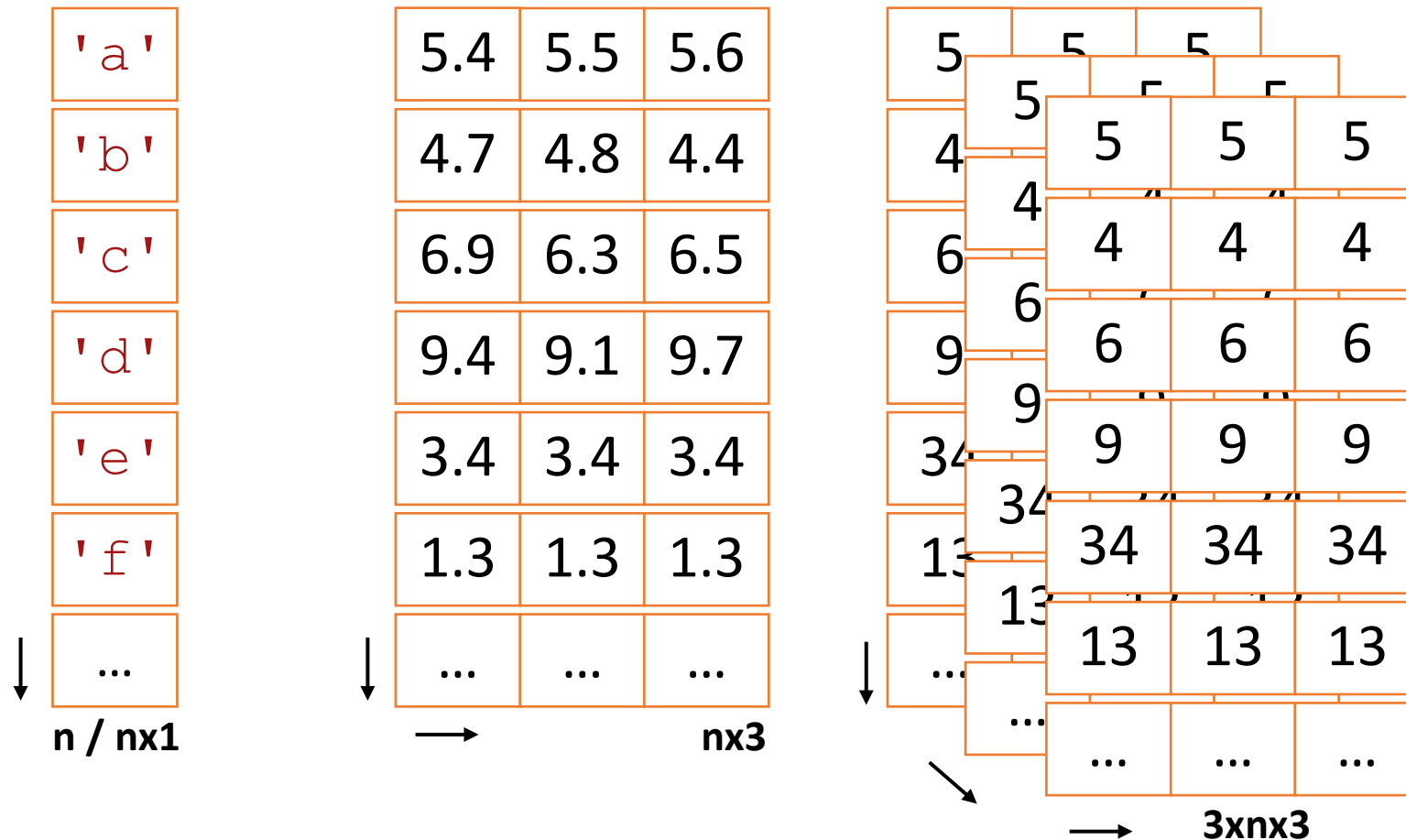
```
A = np.array([1, 2, 3])
```

```
doubleA = np.add(A, 2)
```

numerical operations are made easier
with NumPy

click [here](#) to see a list of them

ndarray: shape



can be even more dimensions!

ndarray: numerical operations

x	y
0	5
1	4
2	3
3	2
4	1
5	0

Arithmetic

$x / 3$	$x + y$
0	5
0.33	5
0.67	5
1.0	5
1.33	5
1.67	5

Conditional

$x \geq 3$	$x < y$
False	True
False	True
False	True
True	False
True	False
True	False

ndarray: numerical operations (Cont.)

What if you want to perform arithmetic operation on arrays with different shapes?


x

0
1
2
3
4
5

z

0	0
0	0
0	0

$x + z$

not possible! 

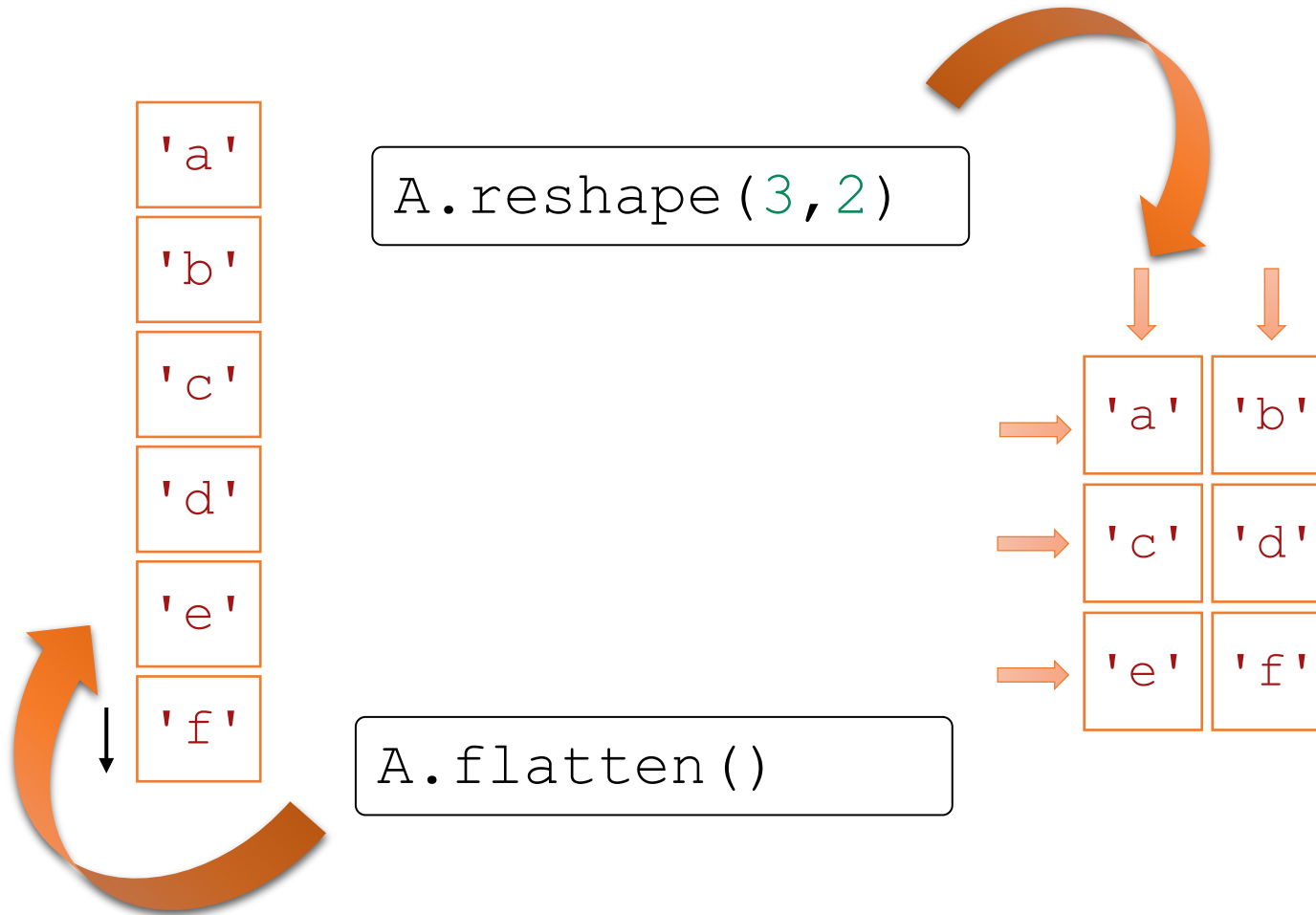
**operations are
done element
wise**

**so the dimensions
must match**

we must reshape the array to
perform this operation

ndarray: reshape

```
A = np.array(['a', 'b', 'c', 'd', 'e', 'f'])
```



other reshape methods are

- `np.concatenate()`
- `np.hstack()`
- `np.vstack()`
- `np.transpose()`

ndarray: initialization

```
x = np.arange(6)
```

0
1
2
3
4
5

↓

```
x = np.zeros(3)
```

0
0
0

```
x = np.zeros((3, 2))
```

0	0
0	0
0	0

and there are more ways
such as

- np.full
- np.random.rand
- np.ones
- np.eye ...

ndarray: methods

```
B = np.array([ (2, 3, 4),  
              (12, 14, 15),  
              (9, 10, 6) ])
```

```
print(B.ndim)
```

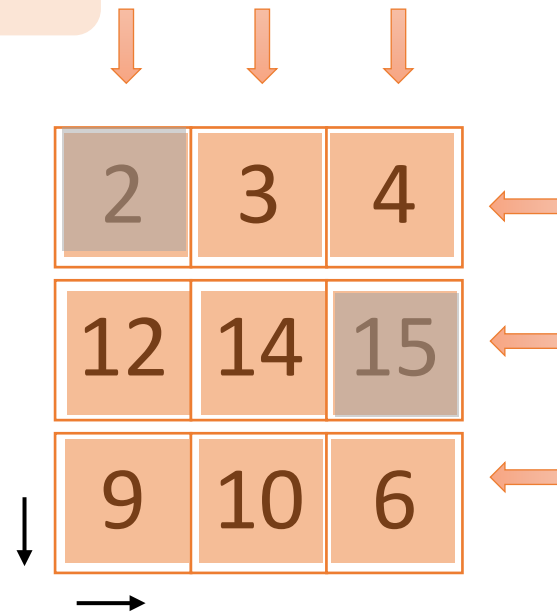
→ 2

```
print(B.shape)
```

→ (3, 3)

```
print(B.size)
```

→ 9



```
print(B.dtype)
```

→ int64 or int32

```
print(B.max())
```

→ 15

```
print(B.min())
```

→ 2

ndarray: methods (Cont.)

```
B = np.array([ (2, 3, 4),  
               (12, 14, 15),  
               (9, 10, 6) ])
```

```
np.sum(B, axis=1)
```

2	3	4	9
12	14	15	41
9	10	6	25
23	27	25	

```
np.sum(B, axis=0)
```

ndarray: indexing, slicing, stepping

```
A = np.array(['a', 'b', 'c', 'd', 'e', 'f'])
```

Indexing

```
A[0]
```

'a'

Slicing

```
A[1:4]
```

```
array(['b', 'c', 'd'])
```

Stepping

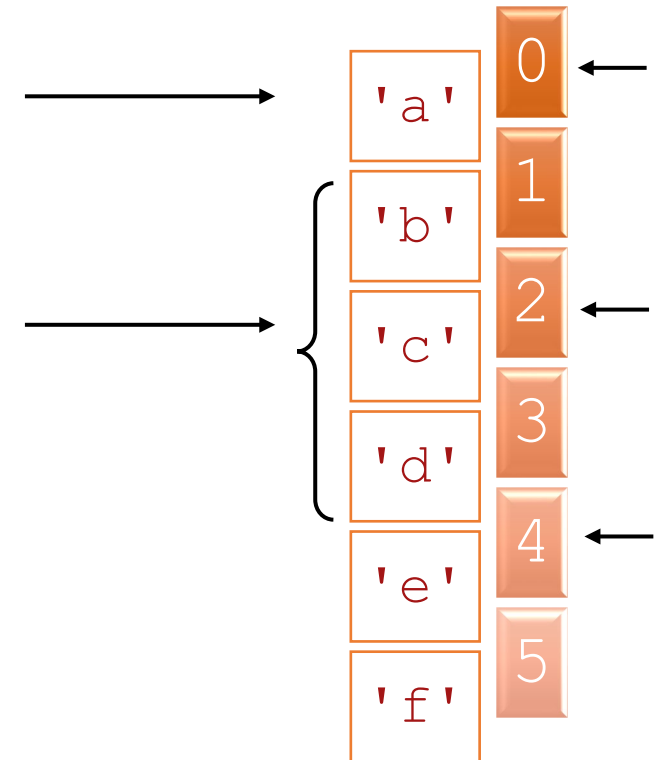
```
A[::2]
```

```
array(['a', 'c', 'e'])
```

Negative Indexing

```
A[::-1]
```

```
nd.array(['f', 'e', 'd', 'c', 'b', 'a'])
```



(6,1)

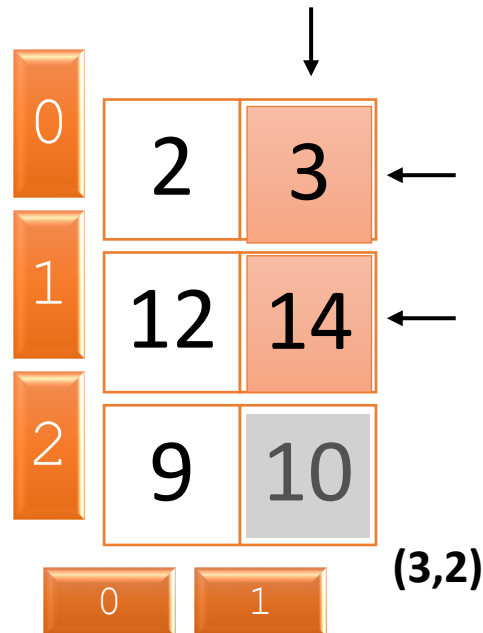
ndarray: indexing, slicing, stepping (Cont.)

```
B = np.array([ (2, 3),  
              (12, 14),  
              (9, 10) ])
```

```
print(B[2, 1])
```

```
print(B[:, 2, 1])
```

- comma separates the each dimension
- colon (:) selects everything in that dimension



```
print(B[::-1, ::-1])
```



ndarray: indexing, slicing, stepping (Cont.)

```
C = np.array([
    [1, 2, 3, 4],
    [4, 5, 6, 7],
    [8, 9, 10, 11]],

    [
    [12, 13, 14, 15],
    [16, 17, 18, 19],
    [20, 21, 22, 23]]
    )
```

```
print(C[0, 2, -1])
```

```
print(C[1, 1, 2])
```

```
print(C[:, 1:2, 3:])
```

1	2	3	4				
4	5	6	7	12	13	14	15
8	9	10	11	16	17	18	19
				20	21	22	23

(2,3,4)

ndarray: copies vs views

- Slicing creates a view on the original array, not a new array.

```
B = np.array([ (2, 3),  
               (12, 14),  
               (9, 10) ])
```

```
C = B[:2, 1]
```

2	3
12	14
9	10

view of array B

3
14

ndarray: copies vs views (Cont.)

- Changing a view changes the original array.

```
B = np.array([ (2, 3),  
              (12, 14),  
              (9, 10) ])
```

```
c = B[:2, 1]
```

2	3
12	14
9	10

this also changes
the original array B

```
c[0] = 40
```

2	40
12	14
9	10

ndarray: copies vs views (Cont.)

- Use `.copy()` method to not change the original array.

```
B = np.array([(2, 3),  
             (12, 14),  
             (9, 10)])
```

2	3
12	14
9	10

```
c = B[:, 1].copy()
```

3
14

```
c[0] = 40
```



B remains unchanged
because c is a copy of B,
not a view of B

Broadcasting

- It is possible to do operations on arrays of different dimensions if NumPy can transform these arrays so that they all have same size.
- This conversion is called **broadcasting**.

Broadcasting (Cont.)

- RULE 1: If the two arrays differ in their number of dimensions, the shape of the **one with fewer dimensions is *padded*** with ones on its leading side.

				Let's say you get a delivery – no more erasers, 1 book for each basket and 2 pens for each basket							
<u>Existing Stock</u>								<u>Updated Stock</u>			

Broadcasting (Cont.)

- RULE 2: If the shape of the two arrays does not match in any dimension, the array with **shape equal to 1 in that dimension is stretched** to match the other shape.

Initial stock for
erasers, books
and pens each

Basket 1	0	0	0
Basket 2	10	10	10
Basket 3	20	20	20
Basket 4	30	30	30

(4,1)

+

Let's say you get a delivery –
no more erasers,
1 book for each basket and
2 pens for each basket

0	1	2
0	1	2
0	1	2
0	1	2

(3,)

=

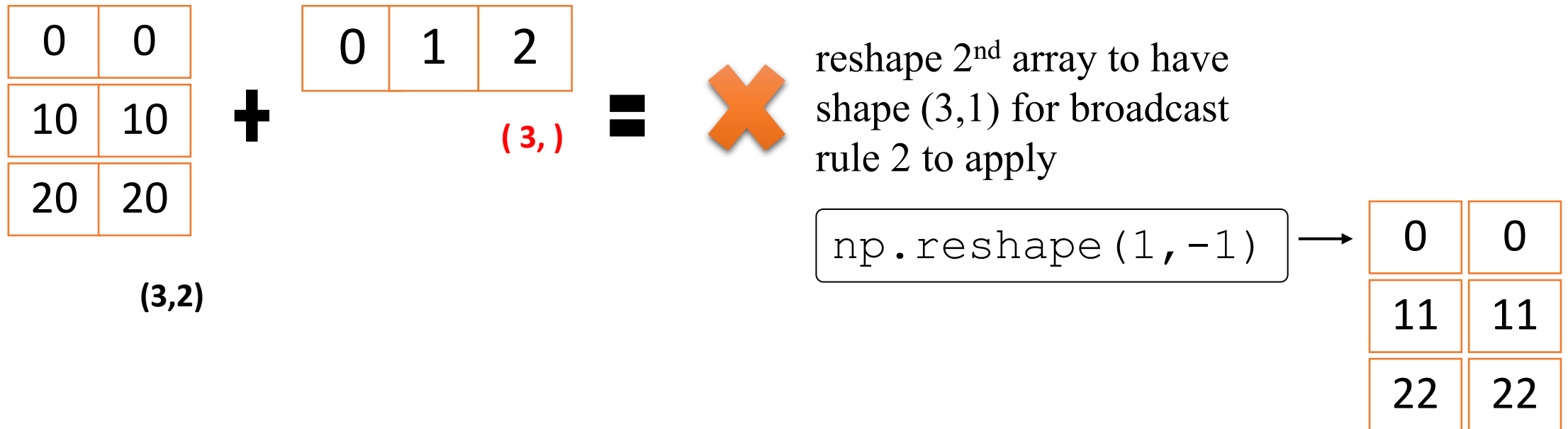
Updated Stock

Eraser	Books	Pens
0	1	2
10	11	12
20	21	22
30	31	32

(4,3)

Broadcasting (Cont.)

- RULE 3: If in any dimension the sizes disagree and neither is equal to 1, an error is raised.



Broadcasting (Cont.)

- In fact, many examples in previous sections are all using broadcasting technique.
- Being able to write short codes is not the only reason for broadcasting in NumPy.
- Broadcasting often makes array operation faster.

Iteration

daily price
of a stock

5.49		
5.65	$(5.49 + 5.65) / 2$	-
5.87	$(5.65 + 5.87) / 2$	-
6.32	$(5.87 + 6.32) / 2$	5.57
6.59		5.76
6.85		6.10
7.56		6.45
8.56		6.72
9.25		7.20
9.99		8.06

N-
days

Simple model that predict today's price as
the average of the price of the last **2** days

-
-
5.57
5.76
6.10
6.45
6.72
7.20
8.06
8.90

N-2
days

How do we do this in NumPy?

```
stockpred[0] = stockprice[0:2].mean()
```

```
stockpred[1] = stockprice[1:3].mean()
```

```
stockpred[2] = stockprice[2:4].mean()
```

```
stockpred[3] = stockprice[3:5].mean()
```

```
stockpred[i] = stockprice[i:i+2].mean()
```

Iteration(Cont.)

Simple model that predicts today's price as the average of the price of the last 2 days

Initialize the
array with
desired length

```
stockpred = np.zeros(N-2)
```

Run a
loop over
the
indices of
array to
fill

```
for i in range(N-2):
```

```
    stockpred[i] = stockprice[i:i+2].mean()
```

The exact
position to fill

Slice of the
original array
to be used in
operation

Value to fill
in that
position

the
operation to
perform

This Week

1. Object-Oriented Programming (OOP)

- Class and Object
- Class Attributes and Methods

2. NumPy

- ndarray
- Numerical Computing with NumPy

Additional sources for beginner Python programmers:

- NumPy official webpage <https://numpy.org/>
- NumPy official documentation <https://docs.scipy.org/doc/numpy/reference/>
- SciPy Lectures <https://scipy-lectures.org/>
- Python for data Analysis <http://shop.oreilly.com/product/0636920023784.do>
- Guide to NumPy <https://web.mit.edu/dvp/Public/numpybook.pdf>
- Python Data Science Handbook <https://jakevdp.github.io/PythonDataScienceHandbook/>

Next Weeks

1. **Pandas**
2. **Matplotlib**
3. **Linear Algebra in Python**
4. **...**