# Lecture 7 Object Oriented Programming

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Computer Science and Mathematics Preparatory Course

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# Overview

#### 1. Object Oriented Programming

- What is OOP?
- ► Example Project
- ➤ Inheritance
- ➤ Modules in Python

## 2. Tasks

#### 3. Outlook: Matrices and Scientific Programming

- Matrices Quick Summary
- ➤ The Numpy Module
- Matrix Calculation with Numpy

# **Programming Paradigms**

#### **Procedural Programming**

- A problem is solved by manipulating data structures through procedures
- The key is to write the right logic
- Efficiency is a main focus of procedural programming

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#### **Procedural Programming**

- A problem is solved by manipulating data structures through procedures
- ► The key is to write the right logic
- Efficiency is a main focus of procedural programming

#### **Object oriented Programming**

- A problem is solved by modeling it's processes
- The key is to figure out the relevant entities and their relations
- Programming Logic is tightly coupled to entities

# **Classes vs. Objects**

# Class

# Person

first name last name age email

## **Classes vs. Objects**

Class

# **Objects (Instances)**



# **Classes Bind Variables Together**

Instead of writing something like this

```
#Alice's attributes
alice_name = "Alice"
alice_last_name = "Anderson"
alice_age = 28
```

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```

Objects encapsulate multiple variables in one place

```
#A Person-object variable
alice = Person("Alice","Anderson",28)
```

# **Classes are Advanced Data Types**

Object variables can be treated like simple types

```
#Two Person-object variables
alice = Person("Alice","Anderson",28)
rob = Person("Rob","Robertson",17)
#Objects can be stored in lists
myPersonList = [] #I want to manage persons
myPersonList.append(rob)
#Objects can be arguments of self-defined functions
calculate_year_of_birth(alice)
```

## **Class Definition**

A class needs to be defined

This is enough to create a class-object

robby = Person("Rob", "Robertson", 17)

# **Accessing Class Attributes**

Class attributes can be accessed via the '.' operator

robby = Person("Rob", "Robertson", 17)

```
f_name = robby.first_name #"Rob"
l_name = robby.last_name #"Robertson"
age = robby.age #17
```

## **Accessing Class Attributes**

Class attributes can be accessed via the '.' operator

robby = Person("Rob", "Robertson", 17)

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f_name = robby.first_name #"Rob"
l_name = robby.last_name #"Robertson"
age = robby.age #17
```

They can also be assigned after initialization

robby.age = 18 #As he gets older robby.l\_name = "Peterson" #If he marries

## **Objects and Functions**

▶ We can use objects as function arguments

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We can use objects as function arguments

```
Usage:
```

```
robby = Person("Rob","Robertson",17)
print_info(robby)
#This prints: "Rob Robertson is 17 years old"
alice = Person("Alice","Anderson",28)
print_info(alice)
#This prints: "Alice Anderson is 28 years old"
```

#### **Function Encapsulation**

Functions can even be defined inside classes

```
class Person: #This defines the class Name
   #The __init__ function
   def __init__(self, first_name,last_name,age):
       #The passed values are stored in the class
       self.first_name = first_name
       self.last_name = last_name
       self.age = age
   #Our print_info function
   def print_info(self): #Note how the argument changed
       print(self.first_name +" " +self.last_name +" is
           \hookrightarrow " +str(self.age) +" years old.")
```

### **Function Encapsulation**

A function can be called directly from the object

```
robby = Person("Rob","Robertson",17)
robby.print_info()
#This prints: "Rob Robertson is 17 years old"
alice = Person("Alice","Anderson",28)
```

```
alice.print_info()
#This prints: "Alice Anderson is 28 years old"
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## **Function Encapsulation**

A function can be called directly from the object

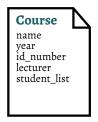
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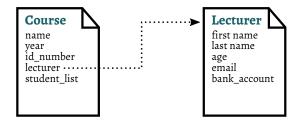
This way a potential programmer/user does not need to know the internal structure of the particular class, e.g. *list.append()*.

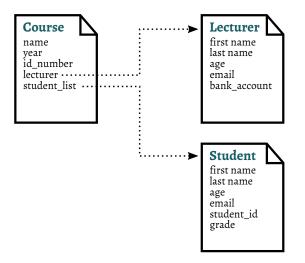
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#### How would an OOP model look like?







#### The course class

```
class Course: #This defines the class Name
  #The __init__ function
  def __init__(self, name,year,id_number,lecturer):
    #The passed values are stored in the class
    self.name = name
    self.year = year
    self.id_number = id_number
    self.lecturer = lecturer
    self.student_list = [] #empty upon creation
```

#### The lecturer class

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c\_bank\_account = cscience\_course.lecturer.bank\_account

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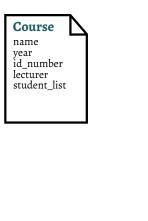
c\_bank\_account = cscience\_course.lecturer.bank\_account

This works independent of course and lecturer

#### **The Student Class**

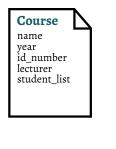
This class looks similar to the lecturer

```
class Student: #This defines the class Name
   #The init function
   def __init__(self, first_name,last_name,age,email,
       \hookrightarrow student id):
       #The passed values are stored in the class
       self.first_name = first_name
       self.last name = last name
       self.age = age
       self.email = email
       self.student_id = student_id
       self.grade = -1
```



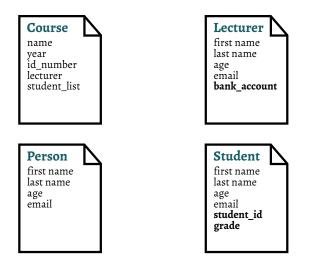


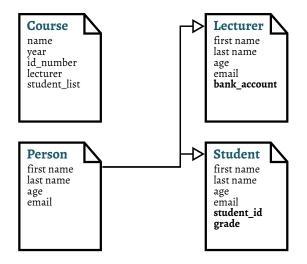


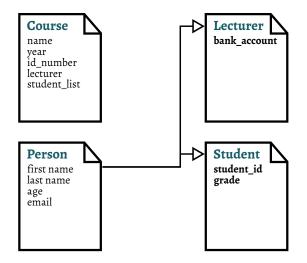




Student first name last name age email student\_id grade







#### **The Person Class**

We will use the Class Person as Super-Class

```
class Person: #This defines the class Name
  #The __init__ function
  def __init__(self, first_name,last_name,age,email):
    #The passed values are stored in the class
    self.first_name = first_name
    self.last_name = last_name
    self.age = age
    self.email = email
```

## Inheritance

Lecturer and Student will inherit from Person

```
class Lecturer(Person): #Brackets declare inheritance
   #The __init__ function is overrriden
   def __init__(self,f_name,l_name,age,email,b_acc):
       #The super() calls the parent function
       super().__init__(f_name,l_name,age,email)
       self.bank account = b acc
class Student(Person): #Brackets declare inheritance
   #The __init__ function is overrriden
   def __init__(self,f_name,l_name,age,email,stud_id):
       super().__init__(f_name,l_name,age,email)
       self.student_id = stud_id
       self.grade = -1
```

# **Modifiying the Parent Class**

Functions of the parent class are available to child classes

```
class Person: #This defines the class Name
```

def \_\_init\_\_(self, first\_name,last\_name,age,email):
 #The passed values are stored in the class
 self.first\_name = first\_name
 self.last\_name = last\_name
 self.age = age
 self.email = email

#### **Using Parent Functions**

Functions of the parent class are available to child classes

```
student_rob.print_info()
lecturer_jan.print_info()
#Prints:
#Rob Robertson is 25 years old.
#Jan Tekuelve is 30 years old.
```

# **Completing the Example**

The course needs to be able to add students

```
#Inside the Course class
def enroll(self,student):
    self.student_list.append(student)
    #Enroll adds them to the course internal list
```

Minimal example:

# Creating your own Python Modules

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► This allows for flexible re-usability of code

# Advantages/Disadvantages of OOP

#### Advantages:

- Design Benefit: Real/World processes are easily transferable in code
- Modularity: Extending and reusing software is easy
- Software Maintenance: Modular code is easier to debug

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#### **Disadvantages:**

- Desing Overhead: Modeling requires longer initial development time
- Originally OOP required more "coding"

## Tasks

- 1. Download todays class definitions *unimanager.py* and create a separate script that uses this module to create a course, a lecturer and three sample students.
  - Enroll all students to the course.
  - After enrolling iterate through the student list to print the info of all enrolled students. You can access the student\_list via the course object.
  - ▶ In the loop use the *print\_info()* function.
- 2. Add a *print\_info()* function to the class definition of Course in *unimanager.py*. This function should print the course name, its lecturer and each student of the course with his/her student ID.
  - The function should be defined in the Course class and its only argument should be self
  - The course name, the lecturer and its student\_list can be accessed via the self keyword.

A Matrix  $\mathbf{A}_{m,n}$  is a rectangular array arranged in *m* rows and *n* columns.

► Example:

$$oldsymbol{A}_{3,4}=egin{pmatrix} 1&2&3&4\5&6&7&8\9&10&11&12 \end{pmatrix}$$

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► Example:

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A single element in a matrix is usually denoted by  $a_{i,j}$ , where *i* is the row and *j* the column index. For example  $a_{2,3} = 7$ .

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- A matrix  $A_m$ , n, where m = n is called a square matrix
- A matrix that has only entries on the diagonal is called a **diagonal matrix**

$$\mathbf{D}_{3,3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 4 \end{pmatrix}$$
 Special case **identity matrix**  $\mathbf{I}_{3,3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ 

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Subtraction works analogously:

$$\boldsymbol{A}_{3,2} - \boldsymbol{B}_{3,2} = \begin{pmatrix} 1 & 2 \\ 5 & 6 \\ 9 & 10 \end{pmatrix} - \begin{pmatrix} 4 & 2 \\ 3 & 1 \\ 8 & 2 \end{pmatrix} = \begin{pmatrix} 1-4 & 2-2 \\ 5-3 & 6-1 \\ 9-8 & 10-2 \end{pmatrix} = \begin{pmatrix} -3 & 0 \\ 2 & 5 \\ 1 & 8 \end{pmatrix}$$

#### Scalar Multiplication and Transposition

Multiplication with scalar values is also applied element-wise:

$$\mathbf{A}_{3,2} \cdot 3 = \begin{pmatrix} 1 & 2 \\ 5 & 6 \\ 9 & 10 \end{pmatrix} \cdot 3 = \begin{pmatrix} 1 \cdot 3 & 2 \cdot 3 \\ 5 \cdot 3 & 6 \cdot 3 \\ 9 \cdot 3 & 10 \cdot 3 \end{pmatrix} = \begin{pmatrix} 3 & 6 \\ 15 & 18 \\ 27 & 30 \end{pmatrix}$$

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The transposition A<sup>T</sup> of a matrix switches the roles of row and columns Example:

$$\boldsymbol{A}_{3,2}^{T} = \begin{pmatrix} 1 & 2 \\ 5 & 6 \\ 9 & 10 \end{pmatrix}^{T} = \begin{pmatrix} 1 & 5 & 9 \\ 2 & 6 & 10 \end{pmatrix}$$

The transposition turns a  $m \times n$  matrix into a  $n \times m$  matrix.

- Matrices A and B can be multiplied with each other, if the number of columns of A<sub>m,n</sub> matches the number of rows in B<sub>n,o</sub>.
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   Multiply Row by Column

$$\boldsymbol{A}_{2,3} \cdot \boldsymbol{B}_{3,3} = \begin{pmatrix} 3 & 6 & 5 \\ 4 & 2 & 1 \end{pmatrix} \cdot \begin{pmatrix} 4 & 3 & 8 \\ 1 & 2 & 10 \\ 7 & 3 & 2 \end{pmatrix} = \begin{pmatrix} - & - & - \\ - & - & - \end{pmatrix}$$

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#### The Numpy Module



- Numpy is part of SciPy the module for scientific programming
- ▶ It should have been installed with matplotlib
- It is usually imported like this:

import numpy as np

#### The Numpy Array

Numpy brings its own data structure the numpy array

```
import numpy as np
#Arrays can be created from lists
array_example = np.array([1,6,7,9])
#Arrays can be created with arange
#An array with numbers from 4 to 5 and step size 0.2
array2 = np.arange(4,5,0.2) #5 is not in the array
print(array2) # [4.0 4.2 4.4 4.6 4.8]
```

Elements of an array can be manipulated simultaneously

array3 = array2\*array2 #For example with multiplication
print(array3)# [16.0 16.64 19.36 21.16 23.04]

# Matplotlib and Numpy

```
Plotting sin(x) from 0 to \pi with lists
```

Plotting sin(x) from 0 to  $\pi$  with numpy

```
xValues = np.arange(0,math.pi,0.5)
yValues = np.sin(xValues)
plt.plot(xValues,yValues)
```

#### Numpy Arrays as Matrices

• Creating the following matrix: 
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \end{pmatrix}$$

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# This creates a 3x4 Matrix

A = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])

#### Numpy Arrays as Matrices

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# This creates a 3x4 Matrix

A = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])

Numpy treats such an array as a matrix

```
arr_dim = A.shape #Gives you the shape of your matrix
print(arr_dim) #Prints (3,4)
# Access elements with indexing
single_number = A[1,3] #8, 2nd list,4th element
num2 = A[0,1] #2, 1st list, 2nd element
```

## **Matrix Operations in Numpy**

Matrix Addition: 
$$\begin{pmatrix} 1 & 2 & 3 \\ 5 & 6 & 7 \end{pmatrix} + \begin{pmatrix} 3 & 5 & 1 \\ 5 & -3 & 1 \end{pmatrix} = \begin{pmatrix} 4 & 7 & 4 \\ 10 & 3 & 8 \end{pmatrix}$$

In numpy code:

```
A = np.array([[1,2,3], [5,6,7]])
B = np.array([[3,5,1], [5,-3,1]])
C = A + B
D = A - B #Subtraction works analogously
print(D) #[[-2 -3 2],[0 9 6]]
```

# Matrix Operations in Numpy

• Matrix Multiplication: 
$$\begin{pmatrix} 1 & 2 & 3 \\ 5 & 6 & 7 \end{pmatrix} * \begin{pmatrix} 3 & 5 \\ 5 & -3 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 16 & 2 \\ 52 & 14 \end{pmatrix}$$

- In numpy code:
  - A = np.array([[1,2,3], [5,6,7]]) E = np.array([[3,5], [5,-3],[1,1]])
  - F = np.matmul(A,E)

print(F) # [[16,2],[52,14]]

# **Matrix Operations in Numpy**

• Matrix Multiplication: 
$$\begin{pmatrix} 1 & 2 & 3 \\ 5 & 6 & 7 \end{pmatrix} * \begin{pmatrix} 3 & 5 \\ 5 & -3 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 16 & 2 \\ 52 & 14 \end{pmatrix}$$

In numpy code:

Do not confuse with element-wise multiplication

```
A = np.array([[1,2,3], [5,6,7]])
```

- B = np.array([[3,5,1], [5,-3,1]])
- G = A\*B # [[3,10,3], [25,-18,7]]

/ \

### Matrix Operations in Numpy

It also works for vectors:

$$\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = \mathbf{v}_1^T \mathbf{v}_2 = \begin{pmatrix} \mathbf{I} & \mathbf{2} & \mathbf{3} \end{pmatrix} * \begin{pmatrix} \mathbf{3} \\ \mathbf{5} \\ \mathbf{1} \end{pmatrix} = \mathbf{16}$$

► In numpy code:

V1 = np.array([1,2,3])
V2 = np.array([3,5,1])
R = np.matmul(V1,V2)
print(R) # 16

/ \

## Matrix Operations in Numpy

It also works for vectors:

$$\langle v_1, v_2 \rangle = v_1^T v_2 = \begin{pmatrix} 1 & 2 & 3 \end{pmatrix} * \begin{pmatrix} 3 \\ 5 \\ 1 \end{pmatrix} = 16$$

► In numpy code:

V1 = np.array([1,2,3]) V2 = np.array([3,5,1]) R = np.matmul(V1,V2) print(R) # 16

#### Or vectors and matrices if you want to

/

`

#### **Other helpful Operations**

► Transpose Matrices: 
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 5 & 6 & 7 \end{pmatrix}$$
  $\mathbf{A}^{\mathrm{T}} = \begin{pmatrix} 1 & 5 \\ 2 & 6 \\ 3 & 7 \end{pmatrix}$ 

In numpy:

- A = np.array([[1,2,3], [5,6,7]])
- H = A.T # [[1,5], [2,6], [3,7]]

#### Element-wise summing across arrays:

```
sum = np.sum(H) #24,
V1 = np.array([1,2,3]) #works also for 1D-arrays
sum_v = np.sum(V1) # 6
```

#### This concludes the Preparatory Course.

# Any Questions or Feedback?