# SCRIPTING AND PROGRAMMING LABORATORY FOR DATA ANALYSIS

#### Lecture 11

Some more advanced topics & some gaming Sampling of distribution functions

## TRY, EXCEPT

```
try:
    f = open("myfile.txt", "r")
except:
    print("File could not be opened!")
```

```
some statements
try:
    statement1
except:
    statement2
```

This construct allows handle exceptions in a clever way.

The program executes the **try** block. If an exception (error) is encountered, the **except** block is executed.

```
try:
    f = open("myfile.txt", "r")
except FileNotFoundError:
    print("File could not be found!")
except:
    print("Another exception occurred")
```

```
some statements
try:
    statement1
except exception:
    statement2
```

```
some statements
try:
    statement1
except exception1:
    statement2
except exception2:
    statement3
except:
    statement4
```

## TRY, EXCEPT, ELSE, FINALLY

The try-except construct can be also used with an else block. The else block is executed if no exception is raised within the try block.

```
some statements
try:
    statement1

except:
    statement2
else:
    statement2
else:
    print("File could not be found!")
else:
    print("File has been successfully opened!")
```

The try-except construct can be terminated with a finally clausole, which is executed no matter what. In general, this is used to release external resources.

```
some statements
try:
    statement1
except:
    statement2
finally:
    statement_end
```

```
f = open("myfile.dat", "r")
try:
    l = f.readlines()
    w = [int(i) for i in l]
except ValueError:
    print("Lines do not contain integers!")
finally:
    f.close()
```

## **ODECORATORS**

Remember: everything in python is an object, including functions!

#### Example:

```
def first(msg):
    print(msg)

first("Hello")

Hello

second = first
    second("Hello")

Hello
```

This implies that functions can be passed as arguments to other functions (so-called higher-order functions) and can return other functions; in addition, functions can be defined within functions, and embedded functions are not destroyed right after the main function call (closure).

#### Examples:

```
[1]: def increment(x):
         return x + 1
     def decrement(x):
         return x - 1
     def operate(func, x):
         result = func(x)
         return result
    operate(increment,3)
[2]: 4
    operate(decrement,3)
[3]: 2
```

```
[1]: def is_called():
    def is_returned():
        print("Hello")
    return is_returned

[2]: new = is_called()
    new()
    Hello
```

## **ODECORATORS**

Functions and methods are called callable as they can be called. In fact, any object which implements the special \_\_call\_\_() method [see slides before] is said to be callable. So, in the most basic sense, a decorator is a callable that returns a callable.

A decorator takes in a function, adds some functionality and returns it.

#### Without decorators:

```
[1]: def make pretty(func):
         def inner():
             print("I got decorated")
             func()
         return inner
     def ordinary():
         print("I am ordinary")
[2]: ordinary()
     I am ordinary
[3]: pretty = make pretty(ordinary)
     pretty()
     I got decorated
     I am ordinary
[4]: #reassign
     ordinary = make pretty(ordinary)
     ordinary()
     I got decorated
     I am ordinary
```

#### Analogous with decorators:

```
[1]: def make_pretty(func):
         def inner():
             print("I got decorated")
             func()
         return inner
     #decorated function:
     @make pretty
     def ordinary():
         print("I am ordinary")
     #eauivalent to
     ## ordinary = make pretty(ordinary)
[2]: ordinary()
     I got decorated
     I am ordinary
```

## @DECORATION WITH PARAMETERS

#### Without decorators:

```
[1]: def divide(a, b):
         return a/b
    divide(5,3)
21: 1.6666666666666667
    divide(4,0)
[3]:
     ZeroDivisionError
     ~\AppData\Local\Tem
```

#### Decorated, fixed num. Of parameters:

```
[4]:
     def smart divide(func):
         def inner(a, b):
             print("I am going to divide", a, "and", b)
             if b == 0:
                 print("Whoops! cannot divide")
                 return
             return func(a, b)
         return inner
     @smart divide
     def divide(a, b):
         print(a/b)
[5]: divide(4,0)
     I am going to divide 4 and 0
     Whoops! cannot divide
```

Note: parameters of the nested inner() function inside the decorator is the same as the parameters of functions it decorates. This can be generalized!

#### Decorated, arbitrary params:

```
[1]: def smart_divide(func):
         def inner(*args, **kwargs):
             print("I am going to divide", args)
             if args[-1] == 0:
                 print("Whoops! cannot divide")
                 return
             return func(*args, **kwargs)
         return inner
     @smart divide
     def divide(a, b):
         print(a/b)
[2]: divide(4,0)
     I am going to divide (4, 0)
     Whoops! cannot divide
```

## CHAINING DECORATORS

Multiple decorators can be chained in Python.

A function can be decorated multiple times with different (or same) decorators. The decorators should be placed above the desired function. The last above the function call is the first to be applied.

```
[3]: def star(func):
       def inner(*args, **kwargs):
          print("*" * 30)
          func(*args, **kwargs)
          print("*" * 30)
       return inner
    def percent(func):
       def inner(*args, **kwargs):
          print("%" * 30)
          func(*args, **kwargs)
          print("%" * 30)
       return inner
[4]: @percent
    def printer(msg):
       print(msg)
    printer("Hello")
    Hello
    [6]: @percent
    @star
    def printer2(msg):
       print(msg)
[8]: printer2("Hello")
    **********
    Hello
```

#### **ITERABLES**

**Iterable:** An object is said to be iterable if we can get an iterator from it – that is, it can return its members one at a time.

Lists, tuples, strings, and dictionaries are all iterables

```
Objects that define either of the
__iter__ or __getitem__ methods are
iterables (also customized ones)
```

The **iter()** function (which in turn calls the \_\_iter\_\_() method) returns an **iterator** from them.

```
[2]: list_num = [4,5,7]
    for i in list_num:
        print(i)

4
    5
    7
```

```
list_num = [4,5,7]
print(dir(list_num))

['__add__', '__class__', '__class_getitem__', '__con
tains__', '__delattr__', '__delitem__', '__dir__',
'__doc__', '__eq__', '__format__', '__ge__', '__geta
ttribute__', '__getitem__', '__gt__', '__hash__', '__
_iadd__', '__imul__', '__init__', '__init_subclass__
_', '__iter__', '__le__', '__len__', '__lt__', '__mu
l__', '__ne__', '__new__', '__reduce__', '__reduce_e
x__', '__repr__', '__reversed__', '__rmul__', '__set
attr__', '__setitem__', '__sizeof__', '__str__', '_
subclasshook__', 'append', 'clear', 'copy', 'count',
'extend', 'index', 'insert', 'pop', 'remove', 'rever
se', 'sort']
```

#### **ITERATORS**

Iterator in Python is simply an object that can be iterated upon. An object which will return data, one element at a time.

Technically speaking, a Python iterator object must implement two special methods, \_\_iter\_\_() and \_\_next\_\_(), collectively called the iterator protocol.

Note: object.\_\_iter\_\_() and iter(object) do nearly the same thing! Same for object.\_\_next\_\_() and next(object)

```
list_num = [4,5,7]
print(iter(list_num))

list_iterator object at 0x000001B31C0163A0>
```

```
[1]: list num = [4,5,7]
     num it=list num. iter ()
     val1=num it. next ()
     print(val1)
     val2=num it. next ()
                                   list num = [4,5,7]
     print(val2)
                                   num it=iter(list num)
                                   val1=next(num it)
                                   print(val1)
[3]: val3=num it. next ()
     print(val3)
                                   val2=next(num it)
                                   print(val2)
                             [3]: val3=next(num it)
                                   print(val3)
```

#### ITERATORS AND LOOPS

next() manually iterates
through all the items;
when it reaches the end,
it raises the
StopIteration Exception.

for loops are more elegant ways of iterating. A for loop is in fact an infinite loop that stops when the **StopIteration** Exception is raised!

```
[1]: list num = [4,5,7]
     num it=iter(list num)
     val1=next(num it)
     print(val1)
[2]: val2=next(num it)
     print(val2)
[3]: val3=next(num it)
     print(val3)
[4]: val4=next(num it)
     print(val4)
     StopIteration
     Traceback (most recent call last)
     Input In [4], in <cell line: 1>()
     ----> 1 val4=next(num it)
           2 print(val4)
     StopIteration:
```

```
list num = [4,5,7]
for i in list num:
     print(i)
             'Implemented as:
list num = [4,5,7]
my iterator = iter(list num)
while True:
 try:
   i = next(my iterator)
   #do stuff, for example:
   print(i)
 except StopIteration:
   break
```

## BUILD YOUR CUSTOM ITERATOR

When you define your own class, you can make it iterable by defining the \_\_iter\_\_() and \_\_next\_\_() functions.

\_\_iter\_\_() can contain
initializations and should
return the iterator itself

\_\_next\_\_() returns the next
item. When there is no next
element, it should raise a
StopIteration Exception

```
[1]: class PowTwo:
         """Class to implement an iterator
         of powers of two"""
         def init (self, max=0):
             self.max = max
         def iter (self):
             self.n = 0
             return self
         def next (self):
            if self.n <= self.max:</pre>
                 result = 2 ** self.n
                self.n += 1
                return result
             else:
                 raise StopIteration
[2]: numbers = PowTwo(3)
     it = iter(numbers)
                               [4]: for i in PowTwo(5):
     print(next(it))
                                             print(i)
[3]: print(next(it))
     print(next(it))
     print(next(it))
     print(next(it))
                                       16
                                       32
     StopIteration
     <ipython-input-3-3c913438779d> in <mod</pre>
           2 print(next(it))
```

#### INFINITE ITERATORS

An iterator can be defined so that it never ends.

But: not recommended!!

```
[1]: class InfIter:
         """Infinite iterator
         to return odd numbers"""
         def iter (self):
             self.num = 1
             return self
         def next (self):
             num = self.num
             self.num += 2
             return num
[2]: oddnum = InfIter()
     it = iter(oddnum)
     print(next(it))
[3]: print(next(it))
     print(next(it))
     print(next(it))
```

#### GENERATORS

Generators are simple ways of generating some types of iterators without having to rely on a class.

In fact, a generator is a normal functions that returns an iterator, over which we can iterate. The function scope should contain a **yield** statement.

yield pauses the function, saving all its current states for the next successive calls

#### This is long and counterintuitive:

```
class Even:
    """Returns even numbers"""
    def init (self, nmax):
        self.n=2
        self.max=nmax
    def iter (self):
        return self
    def next (self):
        result = self.n
        self.n += 2
        if self.n<=self.max:</pre>
            return result
        else:
            raise StopIteration
numbers = Even(12)
print(next(numbers))
print(next(numbers))
print(next(numbers))
```

A generator is a much better solution (here still not optimal, see next slide):

```
[1]: def my gen():
        n = 2
         print('This is printed first')
        vield n
         n += 2
         print('This is printed second')
        vield n
         n += 2
         print('This is printed at last')
        vield n
[2]: a = my gen()
     print(next(a))
     This is printed first
[3]: print(next(a))
    This is printed second
[4]: print(next(a))
    This is printed at last
```

#### GENERATORS

Difference between a generator and a normal function:

- A generator function contains yield statements.
- When called, it returns an object
   (iterator) but does not start execution
   immediately.
- Methods like \_\_iter\_\_() and \_\_next\_\_()
   are implemented automatically → we can
   iterate through the items using next().
- Once the function yields, it is is paused and the control is transferred to the caller.
- Local variables and their states are remembered between successive calls.
- When the function terminates,
   StopIteration is raised automatically.

```
[1]: def EvenGen(max=0):
          n = 0
          while n < max:
              yield 2*n
              n += 1
[2]: a=EvenGen(2)
     print(next(a))
[3]: print(next(a))
[4]: print(next(a))
     StopIteration
     <ipython-input-4-94b12d</pre>
     ----> 1 print(next(a))
     StopIteration:
[5]: b=EvenGen(5)
     for val in b:
                          <u></u>sa<mark>m</mark>e
          print(val)
```

Note: we have to create a new object to restart the generator

```
for val in EvenGen(5):
    print(val)

0
2
4
6
```

#### MORE ON GENERATORS

A more useful example:

```
[1]: def rev str(my str):
         length = len(my str)
         for i in range(length - 1, -1, -1):
             vield my str[i]
     # For Loop to reverse the string
     for char in rev str("hello"):
         print(char)
```

Normally, generator functions are implemented with a loop having a suitable terminating condition.

#### Python generator expression:

```
[1]: mv list = [1, 3, 6, 10]
                                                         Generator expressions
    # sauare each term
                                                         work very similarly to
    # using list comprehension []
                                                         list comprehension, but
    list = [x^{**2} \text{ for } x \text{ in } my \text{ list}]
                                                         they are surrounded by
    # using a generator ()
                                                         parentheses
    generator = (x^{**}2 \text{ for } x \text{ in my list})
[2]: print(list)
    [1, 9, 36, 100]
                                                         List comprehension
3]: print(generator)
                                                         immediately produces all
                                                         elements, generator
    <generator object <genexpr> at 0x000001F77CFEC120>
                                                         expressions are "lazy":
                                                         they produce an element at
[4]: a = (x**2 \text{ for } x \text{ in my list})
    print(next(a))
                                                         a time
    print(next(a))
    print(next(a))
                                                         → Memory efficient!
    print(next(a))
                                                         Note: generators can be
     100
                                                         used as arguments for
                                                         functions!
[5]: for val in (x**2 for x in my list):
        print(val)
                                                    sum(x**2 for x in my list)
     36
                                                    146
     100
```

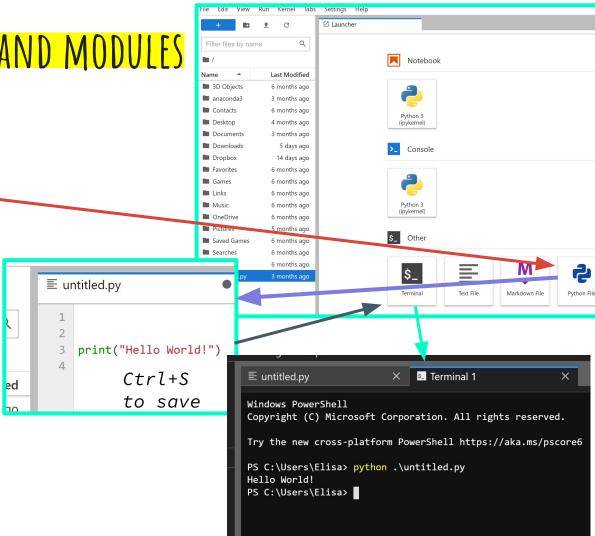
## WRITE YOUR OWN SCRIPT AND MODULES

Any file with extension
•py can be a module (and
your own script!)

- 1. From the launcher:
  create a file .py in
  a directory, put
  there the code to be
  executed. Save it!
- 2. Again from the launcher: open a Terminal in the same directory (or navigate there) and execute the program with:

python name.py

Then press Enter

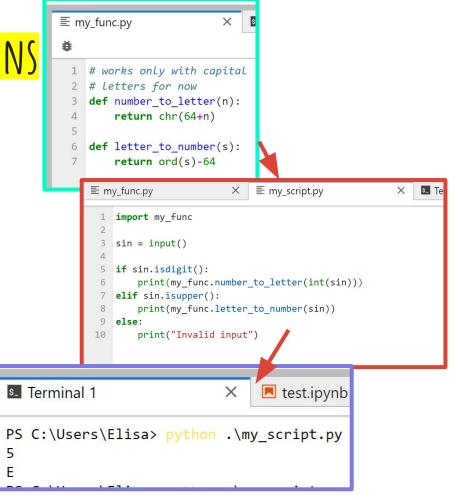


### CREATE A MODULE WITH FUNCTIONS

Knowing this, you can create your own module to be imported if needed. A module can contain the definitions of classes, functions and so forth.

The name of the module should then be imported in the notebook or .py file in which we want to use it.

Note: the **input()** command takes as input the string from the prompt!



#### MONTE CARLO METHODS: GENERATE YOUR DISTRIBUTION FUNCTIONS

Broadly speaking, Monte Carlo methods are methods that rely on the generation of random numbers. They are non-deterministic, but they can give approximate results in a number of situations. Monte Carlo methods are frequently adopted for the solution of problems where the complexity of the model makes a direct solution method infeasible.

**Problem:** Let's assume that we want to generate a set of data between a certain minimum and maximum value, that follow a given distribution function f(x):

How can we do it?

#### **IMPORTANT NOTIONS**

#### Distribution function:

f(x) dx represents the probability  $\mathscr{P}$  for x to fall in the range  $\{x, x + dx\}$ . This means that the probability for x to assume a value in the interval  $\{a, b\}$  is given by

$$\mathscr{P}{a,b} = \int_a^b f(x) \, \mathrm{d}x.$$

Must be normalized to 1

#### Cumulative distribution function:

*cumulative distribution function* (CDF) F(X), which represents the probability for x to have a value minor or equal to X:

$$F(X) = \int_{-\infty}^{X} f(x) dx = \mathcal{P}\{x < X\}.$$

It is an increasing function between 0 and 1

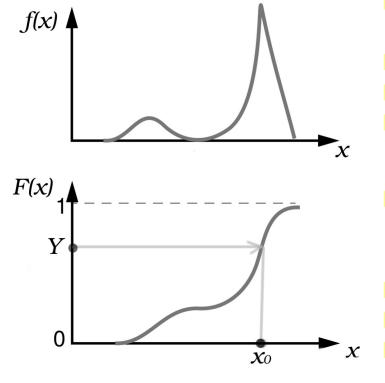
#### <u>It follows that:</u>

$$\frac{\mathrm{d}F(x)}{\mathrm{d}x} = f(x);$$

$$\mathscr{P}\{a,b\} = \int_{a}^{b} f(x) \, \mathrm{d}x = F(b) - F(a).$$

<u>Let's also recall the fundamental</u>
law of probability:

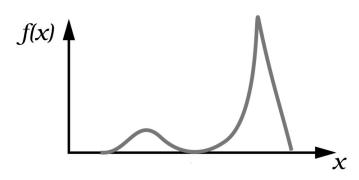
#### MONTE CARLO TRANSFORMATION METHOD

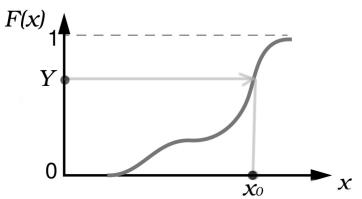


Basic idea: if F(x) = Y is a value assumed by the cumulative distribution function generated from f(x), it must be uniformly distributed between 0 and 1.

This notion can be used to generate values distributed as f(x). NOTE: this method relies on the assumption that we are able to compute the inverse of the cumulative F(x)! [not necessarily analytically]

### MONTE CARLO TRANSFORMATION METHOD





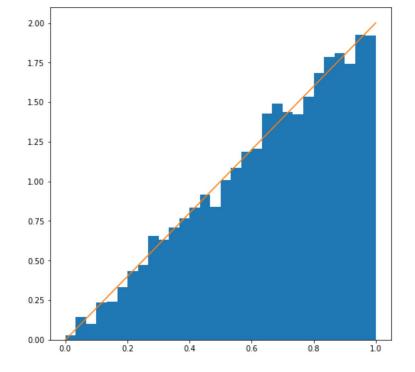
- Take a probability distribution function f(x) of the quantity x you want to sample
- Compute its cumulative distribution function (analytically if possible, otherwise numerically, but be careful!)
- 3. Extract a random number for Y = F(x) of the cumulative distribution function between 0 and 1
- 4. Invert the function F(x) to get X=F<sup>-1</sup>(Y); X is distributed according to f(x)

REPEAT 3 & 4 as many times as you need to get your complete sample

## MONTE CARLO TRANSFORMATION METHOD

An example from jupyter notebook

```
# actual execution of the monte carlo sampling
# f(x) = 2*x
# F(x) = x^2
# --> x = sqrt(Y)
Y = np.random.uniform(0,1,10000)
X = np.sqrt(Y)
fig, ax=plt.subplots(figsize=(8,8))
plt.label("hist(x)")
plt.hist(X,bins=30,range=(0,1),density=True)
plt.plot(x,2*x)
plt.show()
```



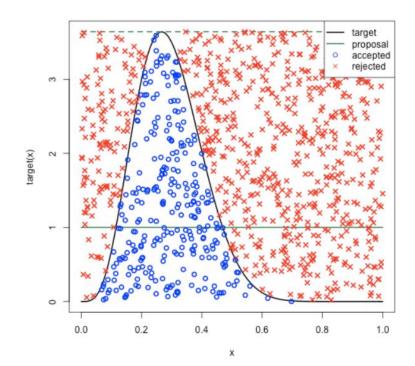
## MONTE CARLO REJECTION METHOD

The inverse transform sampling is not useful when  $F^{-1}$  does not have an explicit form. In this situations a different techniques, always using the inverse sampling, can be used. Suppose we have an additional pdf function g(x) with a cdf that has an explicit form and such that  $f(x) \leq Mg(x)$  over the whole domain on which f(x) is defined, with M a positive constant. In this case to get x samples we use the inverse sampling on the cdf of g(x). Then in order to decide if we can accept the x value we proceed as follows:

- we sample a random number u from a uniform distribution with limits 0 to Mg(x);
- if  $u \le f(x)$  we accept x, otherwise we reject it.

Note that in principle for using this method the pdf does not have to be normalized at 1, since any shift by a constant do not change the relative difference between f(x) and g(x).

The simplest g(x) that we can choose is a constant function i.e. g(x) = C, with  $C \ge \max(f(x))$ :

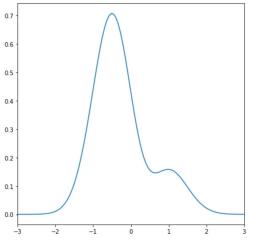


## MONTE CARLO REJECTION METHOD

```
import numpy as np
import matplotlib.pyplot as plt

def f(x):
    return np.sin(x-0.5)**2 * np.exp(-x**2) + .2*np.exp(-x**2)

fig,ax = plt.subplots(figsize=(7,7))
x=np.linspace(-3,3,100)
plt.plot(x, f(x))
plt.xlim(-3,3)
plt.show()
```

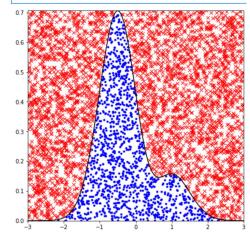


```
N=5000

XRAND = np.random.uniform(-3,3,N)

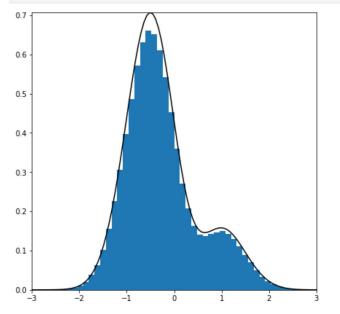
YRAND = np.random.uniform(0,np.amax(f(x)), N)|
filt = YRAND < f(XRAND)

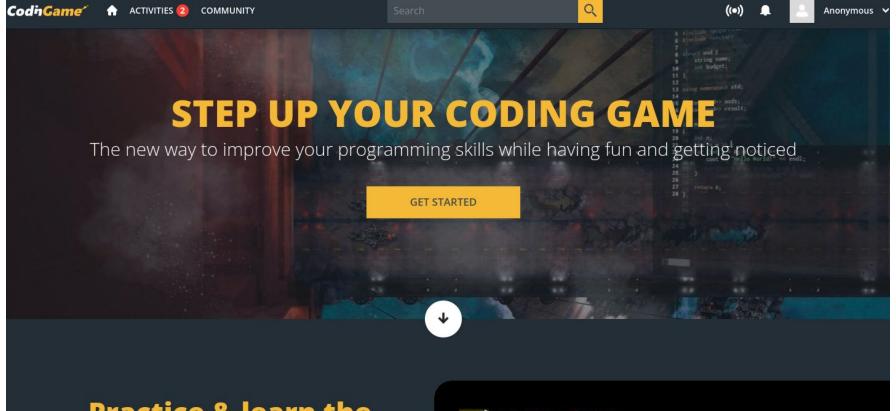
fig,ax = plt.subplots(figsize=(7,7))
plt.plot(XRAND[filt], "RAND[filt], "b.")
plt.plot(XRAND[-filt], YRAND[-filt], "rx")
plt.plot(x, f(x), "k-")
plt.xlim(-3,3)
plt.ylim(0, np.amax(f(x)))
plt.show()
```



## xrand is now distributed as f(x)

```
plt.hist(XRAND[filt], bins=50,range=(-3,3), density=True)
plt.plot(x, f(x), "k-")
plt.xlim(-3,3)
plt.ylim(0, np.amax(f(x)))
plt.show()
```





Practice & learn the fun way



### CODINGAME.COM

Registration needed First: solve a simple puzzle

Then: decide on what you

want to do!

