

# Computing Laboratory

## Basics of Python - IV

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## 1 Importing Modules and Functions

## 2 Mathematical Functions

- Mathematical Functions for Real Numbers
- Mathematical Functions for Complex Numbers

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# Importing modules and functions

## Importing a module:

```
import <module_name>
```

OR

```
import <module_name> as <custom_name>
```

# Importing modules and functions

## Importing a module:

```
import <module_name>
```

OR

```
import <module_name> as <custom_name>
```

## Using a function:

```
import <module_name>  
<module_name>.<function_name>()
```

OR

```
import <module_name> as <custom_name>  
<custom_name>.<function_name>()
```

OR

```
from <module_name> import <function_name>  
<function_name>()
```

# Using built-in methods

There are some functions in Python that does not require to import a module because they work on specific data structures.

- Built-in methods for strings (e.g., `capitalize()`, `strip()`, `zfill()`, etc.)
- Built-in methods for lists/arrays (e.g., `sort()`, `reverse()`, `clear()`, etc.)
- Built-in methods for dictionaries (e.g., `keys()`, `values()`, `update()`, etc.)
- Built-in methods for tuples (e.g., `count()`)

**Note:** The `sort()` method in Python implements the hybrid algorithm *Timsort*, derived from merge sort and insertion sort.

# Mathematical functions for real numbers

## Using the math module:

```
import math  
math.<function_name>()
```

# Mathematical functions for real numbers

## Using the math module:

```
import math  
math.<function_name>()
```

<code>ceil(x)</code>	– Ceiling of $x$
<code>comb(n, r)</code>	– Number of ways to choose $r$ from $n$ ( ${}^n C_r$ )
<code>copysign(x, y)</code>	– Float with magnitude of $x$ but sign of $y$
<code>fabs(x)</code>	– Absolute value of $x$
<code>factorial(x)</code>	– Factorial of $x$
<code>floor(x)</code>	– Floor of $x$

**Table:** Functions in math module

# Mathematical functions for real numbers

## Library code of the factorial() function:

```
static PyObject * math_factorial(PyObject *module, PyObject *arg){
    ...
    /* use lookup table if x is small */
    if (x < (long)Py_ARRAY_LENGTH(SmallFactorials))
        return PyLong_FromUnsignedLong(SmallFactorials[x]);
    /* else express in the form odd_part * 2**two_valuation,
    and compute as odd_part << two_valuation. */
    odd_part = factorial_odd_part(x);
    if (odd_part == NULL)
        return NULL;
    two_valuation = x - count_set_bits(x);
    result = _PyLong_Lshift(odd_part, two_valuation);
    Py_DECREF(odd_part);
    return result;
}
```

Source: [github.com/python/cpython/blob/main/Modules/mathmodule.c](https://github.com/python/cpython/blob/main/Modules/mathmodule.c)

# Mathematical functions for real numbers

<code>fmod(x, y)</code>	– $x \% y$ (preferable for integers)
<code>frexp(x)</code>	– Mantissa and exponent of $x$ as a pair $(m, e)$
<code>fsum(iterable)</code>	– Accurate floating point sum of values in iterable
<code>gcd(x, y)</code>	– GCD of the integers $x$ and $y$
<code>isclose(x, y, *, rel_tol=1e-09, abs_tol=0.0)</code>	– Whether $x$ is close to $y$ w.r.t max/min allowed tolerance
<code>isfinite(x)</code>	– Whether $x$ is finite (or $\infty/\text{NaN}$ )
<code>isinf(x)</code>	– Whether $x$ is infinite
<code>isnan(x)</code>	– Whether $x$ is NaN (“not a number”)
<code>isqrt(x)</code>	– Integer square root of $x$
<code>ldexp(x, i)</code>	– $x * 2^i$

**Table:** Functions in `math` module

# Mathematical functions for real numbers

```
import math
print(sum([.1, .1, .1, .1, .1, .1, .1, .1]))
print(math.fsum([.1, .1, .1, .1, .1, .1, .1, .1]))
ls = [1/7, 1/7, 1/7, 1/7, 1/7, 1/7, 1/7]
print(sum(ls))
print(math.fsum(ls))
```

## Output:

```
0.7999999999999999
0.8
0.9999999999999998
1.0
```

# Mathematical functions for real numbers

<code>modf(x)</code>	– Fractional and integer parts of $x$
<code>perm(n, r=None)</code>	– Number of ways to choose $k$ from $n$ without repetition ( ${}^n P_r$ )
<code>prod(iterable, *, start=1)</code>	– Product of values in iterable
<code>remainder(x, y)</code>	– Remainder of $x$ when divided by $y$
<code>trunc(x)</code>	– Value of $x$ truncated to an integral
<code>exp(x)</code>	– $e^x$
<code>expm1(x)</code>	– $e^x - 1$ (provides a better precision)
<code>log(x[, base])</code>	– Natural logarithm of $x$ (base $e$ )
<code>log1p(x)</code>	– Natural logarithm of $1+x$ (base $e$ )
<code>log2(x)</code>	– Base-2 logarithm of $x$
<code>log10(x)</code>	– Base-10 logarithm of $x$
<code>pow(x, y)</code>	– $x^y$

**Table:** Functions in `math` module

# Mathematical functions for real numbers

```
import math
print(math.exp(0.8) - 1)
print(math.expm1(0.8))
print(math.log(math.exp(0.8)))
print(math.log(math.expm1(0.8)+1))
```

## Output:

```
1.2255409284924679
1.2255409284924677
0.8000000000000002
0.7999999999999999
```

# Mathematical functions for real numbers

- `sqrt(x)` – Square root of  $x$
- `acos(x)` – Arc cosine of  $x$  (result in radians)
- `asin(x)` – Arc sine of  $x$  (result in radians)
- `atan(x)` – Arc tangent of  $x$  (result in radians)
- `atan2(x, y)` – Arc tangent of  $x/y$  (in radians)
- `cos(x)` – Cosine of  $x$
- `dist(iterable1, iterable1)` – Euclidean distance between iterable 1 and iterable2
- `hypot(*coordinates)` – Euclidean norm  $\sqrt{x^2 + y^2}$  for the point  $(x, y)$ ,  $\sqrt{\text{sum}(x^2 \text{ for } x \text{ in coordinates})}$
- `sin(x)` – Sine of  $x$
- `tan(x)` – Tangent of  $x$

**Table:** Functions in `math` module

# Mathematical functions for real numbers

- `degrees(x)` – Convert angle  $x$  from radians to degrees
- `radians(x)` – Convert angle  $x$  from degrees to radians.
- `acosh(x)` – Inverse hyperbolic cosine of  $x$
- `asinh(x)` – Inverse hyperbolic sine of  $x$
- `atanh(x)` – Inverse hyperbolic tangent of  $x$
- `cosh(x)` – Hyperbolic cosine of  $x$
- `sinh(x)` – Hyperbolic sine of  $x$
- `tanh(x)` – Hyperbolic tangent of  $x$
- `erf(x)` – Error function at  $x$
- `erfc(x)` – Complementary error function at  $x$
- `gamma(x)` – Gamma function at  $x$
- `lgamma(x)` – Natural logarithm of absolute value of Gamma function at  $x$

**Table:** Functions in `math` module

# Mathematical functions for real numbers

- pi –  $\pi = 3.14\dots$ , to available precision
- e –  $e = 2.71\dots$ , to available precision
- tau –  $\tau = 6.28\dots$ , to available precision
- inf – Floating-point positive infinity
- nan – Floating-point NaN

**Table:** Functions in `math` module

# Mathematical functions for complex numbers

## Using the `cmath` module:

```
import cmath  
cmath.<function_name>()
```

OR

```
from cmath import <function_name>  
<function_name>()
```

# Mathematical functions for complex numbers

## Using the `cmath` module:

```
import cmath  
cmath.<function_name>()
```

OR

```
from cmath import <function_name>  
<function_name>()
```

<code>phase(x)</code>	– Phase of $x$ (in float)
<code>polar(x)</code>	– Representation of $x$ in polar coordinates as $(r, \phi)$
<code>rect(r, phi)</code>	– Complex number $x$ with polar coordinates $r$ and $\phi$
<code>exp(x)</code>	– $e^x$
<code>log(x[, base])</code>	– Natural logarithm of $x$ (base $e$ )
<code>log10(x)</code>	– Base-10 logarithm of $x$

Table: Functions in `cmath` module

# Mathematical functions for complex numbers

$\text{sqrt}(x)$	– Square root of $x$
$\text{acos}(x)$	– Arc cosine of $x$
$\text{asin}(x)$	– Arc sine of $x$
$\text{atan}(x)$	– Arc tangent of $x$
$\text{cos}(x)$	– Cosine of $x$
$\text{sin}(x)$	– Sine of $x$
$\text{tan}(x)$	– Tangent of $x$
$\text{acosh}(x)$	– Inverse hyperbolic cosine of $x$
$\text{asinh}(x)$	– Inverse hyperbolic sine of $x$
$\text{atanh}(x)$	– Inverse hyperbolic tangent of $x$
$\text{cosh}(x)$	– Hyperbolic cosine of $x$
$\text{sinh}(x)$	– Hyperbolic sine of $x$
$\text{tanh}(x)$	– Hyperbolic tangent of $x$

Table: Functions in `cmath` module

# Mathematical functions for complex numbers

<code>isclose(x, y, *, rel_tol=1e-09, abs_tol=0.0)</code>	– Whether $x$ is close to $y$ w.r.t max/min allowed tolerance
<code>isfinite(x)</code>	– Whether $x$ is finite (or $\infty$ /NaN)
<code>isinf(x)</code>	– Whether $x$ is infinite
<code>isnan(x)</code>	– Whether $x$ is NaN
<code>pi</code>	– $\pi = 3.14\dots$ , to available precision
<code>e</code>	– $e = 2.71\dots$ , to available precision
<code>tau</code>	– $\tau = 6.28\dots$ , to available precision
<code>inf</code>	– Floating-point positive infinity
<code>infj</code>	– Complex number with zero real and positive infinity imaginary parts
<code>nan</code>	– Floating-point NaN
<code>nanj</code>	– Complex number with zero real part and NaN imaginary part

Table: Functions in `cmath` module

# Statistical functions

## Using the statistics module:

```
import statistics
statistics.<function_name>()
```

# Statistical functions

## Using the statistics module:

```
import statistics  
statistics.<function_name>()
```

mean(X) – Arithmetic mean of the data in X

fmean(X) – Arithmetic mean of the data (converted to float) in X

geometric\_mean(X) – Geometric mean of the data (converted to float) in X

harmonic\_mean(X) – Harmonic mean of the data in X

**Table:** Functions in statistics module

# Statistical functions

<code>median(X)</code>	– Median of the data in X
<code>median_low(X)</code>	– Low median of the data in X
<code>median_high(X)</code>	– High median of the data in X
<code>median_grouped(X, interval)</code>	– Median of the grouped data in X
<code>mode(X)</code>	– Most frequent data item in X
<code>multimode(X)</code>	– Most frequent data items in the order they appear in X

**Table:** Functions in statistics module

# Statistical functions

- `pstdev(X, mu=None)` – Population standard deviation of the data in  $X$
- `pvariance(X, mu=None)` – Population variance of the data in  $X$
- `stdev(X, xbar=None)` – Sample standard deviation of the data in  $X$
- `variance(X, xbar=None)` – Sample variance of the data in  $X$
- `quantiles(X, *, n, method)` – Divide data in  $X$  into  $n$  continuous intervals with equal probability

**Table:** Functions in statistics module

# Shuffling of data

```
import random
ls = [10, 20, 30, 40, 50]
random.shuffle(ls)
print(ls)
```

**Output:** [40, 20, 10, 50, 30]

# Randomization of data

```
import random
random.randint(1, 100) # The interval is [1, 100]
```

**Output:** 15

```
import random
random.random() # The interval is [0.0, 1.0)
```

**Output:** 0.4289859005273219

# Random sampling with numpy

```
import numpy as np
np.random.randint(1, 100) # From uniform distribution
```

**Output:** 97

The interval is [1, 100)

## Random sampling with numpy

```
import numpy as np
np.random.randint(1, 100) # From uniform distribution
```

**Output:** 97 The interval is [1, 100)

```
import numpy as np
np.random.random(5) # From uniform distribution
```

**Output:** array([0.37076725, 0.10547203, 0.21298417, 0.96296838,  
0.15390583]) The interval is [0.0, 1.0)

## Random sampling with numpy

```
import numpy as np
np.random.randint(1, 100) # From uniform distribution
```

**Output:** 97 The interval is [1, 100)

```
import numpy as np
np.random.random(5) # From uniform distribution
```

**Output:** array([0.37076725, 0.10547203, 0.21298417, 0.96296838, 0.15390583]) The interval is [0.0, 1.0)

**Note:** Unlike the `random()` function in `random` module, the one in the `random` submodule of `numpy` module optionally takes size of the array.

## Random sampling with numpy

```
import numpy as np
np.random.randn(5) # From normal distribution
```

**Output:** array([ 1.62029576, 0.29112406, 1.21198839,  
0.26851418, -0.46712281])

```
import numpy as np
np.random.randn(3, 3) # From normal distribution
```

**Output:** array([[ 1.13217437, 0.56093212, -2.36792091],  
[ 1.42652606, 0.68953983, 0.521175 ],  
[ 1.36766496, 0.9488446 , -1.10042531]])

# Effect of the seed value on random sampling

## Code 1:

```
import numpy as np
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
```

# Effect of the seed value on random sampling

## Code 1:

```
import numpy as np
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
```

## Code 2:

```
import numpy as np
np.random.seed(20)
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
```

# Effect of the seed value on random sampling

## Code 1:

```
import numpy as np
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
```

## Code 2:

```
import numpy as np
np.random.seed(20)
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
print(np.random.randint(1, 100))
```

**Note:** Unlike Code 1, Code 2 will generate the same set of random values whenever executed.

# Drawing samples from different distributions using numpy

<code>beta(a, b[, size])</code>	– Samples from Beta distribution
<code>binomial(n, p[, size])</code>	– Samples from binomial distribution
<code>chisquare(df[, size])</code>	– Samples from chi-square distribution
<code>dirichlet(alpha[, size])</code>	– Samples from Dirichlet distribution
<code>exponential([scale, size])</code>	– Samples from exponential distribution
<code>f(dfnum, dfden[, size])</code>	– Samples from F distribution
<code>gamma(shape[, scale, size])</code>	– Samples from Gamma distribution
<code>geometric(p[, size])</code>	– Samples from geometric distribution
<code>gumbel([loc, scale, size])</code>	– Samples from Gumbel distribution

**Table:** Functions in `random` submodule of `numpy`

# Drawing samples from different distributions using numpy

<code>hypergeometric(ngood, nbad, nsample[, size])</code>	– Samples from hypergeometric distribution
<code>laplace([loc, scale, size])</code>	– Samples from Laplace distribution
<code>logistic([loc, scale, size])</code>	– Samples from logistic distribution
<code>lognormal([mean, sigma, size])</code>	– Samples from log-normal distribution
<code>logseries(p[, size])</code>	– Samples from logarithmic series distribution

**Table:** Functions in `random` submodule of `numpy`

# Drawing samples from different distributions using numpy

<code>multinomial(n, pvals[, size])</code>	– Samples from multinomial distribution
<code>multivariate_normal(mean, cov[, size, ...])</code>	– Samples from multivariate normal distribution
<code>negative_binomial(n, p[, size])</code>	– Samples from negative binomial distribution
<code>noncentral_chisquare(df, nonc[, size])</code>	– Samples from noncentral chi-square distribution
<code>noncentral_f(dfnum, dfden, nonc[, size])</code>	– Samples from noncentral F distribution
<code>normal([loc, scale, size])</code>	– Samples from normal distribution

**Table:** Functions in `random` submodule of `numpy`

# Drawing samples from different distributions using numpy

<code>pareto(a[, size])</code>	– Samples from pareto distribution
<code>poisson([lam, size])</code>	– Samples from Poisson distribution
<code>power(a[, size])</code>	– Samples from power distribution
<code>rayleigh([scale, size])</code>	– Samples from Rayleigh distribution
<code>standard_cauchy([size])</code>	– Samples from standard Cauchy distribution
<code>standard_exponential([size])</code>	– Samples from standard exponential distribution
<code>standard_gamma(shape[, size])</code>	– Samples from standard Gamma distribution
<code>standard_normal([size])</code>	– Samples from standard normal distribution

**Table:** Functions in `random` submodule of `numpy`

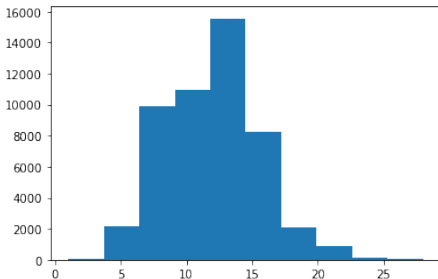
# Drawing samples from different distributions using numpy

<code>standard_t(df[, size])</code>	– Samples from standard Student's t distribution
<code>triangular(left, mode, right[, size])</code>	– Samples from triangular distribution
<code>uniform([low, high, size])</code>	– Samples from uniform distribution
<code>vonmises(mu, kappa[, size])</code>	– Samples from von Mises distribution
<code>wald(mean, scale[, size])</code>	– Samples from Wald distribution
<code>weibull(a[, size])</code>	– Samples from Weibull distribution
<code>zipf(a[, size])</code>	– Samples from Zipf distribution

**Table:** Functions in `random` submodule of `numpy`

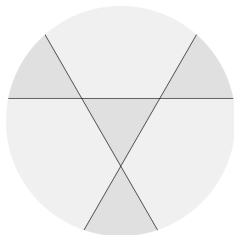
# Visualizing a distribution

```
import numpy as np
sample = np.random.poisson(10, 50000)
import matplotlib.pyplot as plt
plt.hist(sample, 10)
plt.show()
```



## Problems – Day 5

- 1 Suppose a caterer needs to cut a circular pancake into maximum number of pieces (not necessarily of the same shape or size) with a given number of straight cuts. For example, 3 straight cuts across a pancake will produce 6 pieces if the cuts are made intersecting at a common point, but 7 if they do not. Write a program to print the maximum number of pieces that can be obtained with a given number of straight cuts as input.



## Problems – Day 5

- 2 Write a program that will print '0' with a probability of 0.5 and '1' in rest of the cases. Now write a function RUN to count the number of runs in the binary stream generated with your program. The number of runs will simply explore and count dichotomous (reflecting contrast between two things) sequence of values.

**Note:** The following sequence of 0's and 1's has a run count of nine.

0 0 1 1 0 1 0 0 0 0 1 0 0 1 1 1 1 1 0 0

- 3 Write a program to find out the summation of the following series given  $n$  as the user input.

$$\frac{1}{1} + \frac{11}{12} + \frac{111}{123} + \dots + \frac{111 \dots n \text{ times}}{123 \dots n}$$