

# math — Mathematical Functions

**Purpose:** Provides functions for specialized mathematical operations.

The `math` module implements many of the IEEE functions that would normally be found in the native platform C libraries for complex mathematical operations using floating point values, including logarithms and trigonometric operations.

## Special Constants

Many `math` operations depend on special constants. `math` includes values for  $\pi$  (pi),  $e$ , `nan` (not a number), and infinity.

```
# math_constants.py

import math

print('  pi: {:.30f}'.format(math.pi))
print('  e: {:.30f}'.format(math.e))
print('nan: {:.30f}'.format(math.nan))
print('inf: {:.30f}'.format(math.inf))
```

Both  $\pi$  and  $e$  are limited in precision only by the platform's floating point C library.

```
$ python3 math_constants.py

pi: 3.141592653589793115997963468544
e: 2.718281828459045090795598298428
nan: nan
inf: inf
```

## Testing for Exceptional Values

Floating point calculations can result in two types of exceptional values. The first of these, `inf` (infinity), appears when the double used to hold a floating point value overflows from a value with a large absolute value.

```
# math_isinf.py

import math

print('{:^3} {:6} {:6} {:6}'.format(
    'e', 'x', 'x**2', 'isinf'))
print('{:^3} {:^-6} {:^-6} {:^-6}'.format(
    '', '', '', ''))

for e in range(0, 201, 20):
    x = 10.0 ** e
    y = x * x
    print('{:3d} {:<6g} {:<6g} {!s:6}'.format(
        e, x, y, math.isinf(y),
    ))
```

When the exponent in this example grows large enough, the square of  $x$  no longer fits inside a double, and the value is recorded as infinite.

```
$ python3 math_isinf.py

e  x      x**2  isinf
---
0  1      1      False
20 1e+20  1e+40  False
40 1e+40  1e+80  False
60 1e+60  1e+120 False
```

```

80 1e+80 1e+160 False
100 1e+100 1e+200 False
120 1e+120 1e+240 False
140 1e+140 1e+280 False
160 1e+160 inf True
180 1e+180 inf True
200 1e+200 inf True

```

Not all floating point overflows result in `inf` values, however. Calculating an exponent with floating point values, in particular, raises `OverflowError` instead of preserving the `inf` result.

```

# math_overflow.py

x = 10.0 ** 200

print('x      =', x)
print('x*x    =', x * x)
print('x**2   =', end=' ')
try:
    print(x ** 2)
except OverflowError as err:
    print(err)

```

This discrepancy is caused by an implementation difference in the library used by C Python.

```

$ python3 math_overflow.py

x      = 1e+200
x*x    = inf
x**2   = (34, 'Result too large')

```

Division operations using infinite values are undefined. The result of dividing a number by infinity is `nan` (not a number).

```

# math_isnan.py

import math

x = (10.0 ** 200) * (10.0 ** 200)
y = x / x

print('x =', x)
print('isnan(x) =', math.isnan(x))
print('y = x / x =', x / x)
print('y == nan =', y == float('nan'))
print('isnan(y) =', math.isnan(y))

```

`nan` does not compare as equal to any value, even itself, so to check for `nan` use `isnan()`.

```

$ python3 math_isnan.py

x = inf
isnan(x) = False
y = x / x = nan
y == nan = False
isnan(y) = True

```

Use `isfinite()` to check for regular numbers or either of the special values `inf` or `nan`.

```

# math_isfinite.py

import math

for f in [0.0, 1.0, math.pi, math.e, math.inf, math.nan]:
    print('{:5.2f} {}'.format(f, math.isfinite(f)))

```

`isfinite()` returns `false` for either of the exceptional cases, and `true` otherwise.

```

$ python3 math_isfinite.py

```

```
0.00 True
1.00 True
3.14 True
2.72 True
inf False
nan False
```

## Comparing

Comparisons for floating point values can be error prone, with each step of the computation potentially introducing errors due to the numerical representation. The `isclose()` function uses a stable algorithm to minimize these errors and provide a way for relative as well as absolute comparisons. The formula used is equivalent to

```
abs(a-b) <= max(rel_tol * max(abs(a), abs(b)), abs_tol)
```

By default, `isclose()` uses relative comparison with the tolerance set to  $1e-09$ , meaning that the difference between the values must be less than or equal to  $1e-09$  times the larger absolute value between `a` and `b`. Passing a keyword argument `rel_tol` to `isclose()` changes the tolerance. In this example, the values must be within 10% of each other.

```
# math_isclose.py

import math

INPUTS = [
    (1000, 900, 0.1),
    (100, 90, 0.1),
    (10, 9, 0.1),
    (1, 0.9, 0.1),
    (0.1, 0.09, 0.1),
]

print('{:^8} {:^8} {:^8} {:^8} {:^8} {:^8}'.format(
    'a', 'b', 'rel_tol', 'abs(a-b)', 'tolerance', 'close'
))
print('{:~8} {:~8} {:~8} {:~8} {:~8} {:~8}'.format(
    '_', '_', '_', '_', '_', '_'
))

fmt = '{:8.2f} {:8.2f} {:8.2f} {:8.2f} {:8.2f} {!s:>8}'

for a, b, rel_tol in INPUTS:
    close = math.isclose(a, b, rel_tol=rel_tol)
    tolerance = rel_tol * max(abs(a), abs(b))
    abs_diff = abs(a - b)
    print(fmt.format(a, b, rel_tol, abs_diff, tolerance, close))
```

The comparison between 0.1 and 0.09 fails because of the error representing 0.1.

```
$ python3 math_isclose.py
```

a	b	rel_tol	abs(a-b)	tolerance	close
1000.00	900.00	0.10	100.00	100.00	True
100.00	90.00	0.10	10.00	10.00	True
10.00	9.00	0.10	1.00	1.00	True
1.00	0.90	0.10	0.10	0.10	True
0.10	0.09	0.10	0.01	0.01	False

To use a fixed or “absolute” tolerance, pass `abs_tol` instead of `rel_tol`.

```
# math_isclose_abs_tol.py

import math

INPUTS = [
    (1.0, 1.0 + 1e-07, 1e-08),
    (1.0, 1.0 + 1e-08, 1e-08),
```

```

(1.0, 1.0 + 1e-09, 1e-08),
]

print('{:^8} {:^11} {:^8} {:^10} {:^8}'.format(
    'a', 'b', 'abs_tol', 'abs(a-b)', 'close')
)
print('{:-^8} {:-^11} {:-^8} {:-^10} {:-^8}'.format(
    '_', '_', '_', '_', '_'
))

for a, b, abs_tol in INPUTS:
    close = math.isclose(a, b, abs_tol=abs_tol)
    abs_diff = abs(a - b)
    print('{:8.2f} {:11} {:8} {:0.9f} {!s:>8}'.format(
        a, b, abs_tol, abs_diff, close))

```

For an absolute tolerance, the difference between the input values must be less than the tolerance given.

```
$ python3 math_isclose_abs_tol.py
```

a	b	abs_tol	abs(a-b)	close
1.00	1.0000001	1e-08	0.000000100	False
1.00	1.00000001	1e-08	0.000000010	True
1.00	1.000000001	1e-08	0.000000001	True

nan and inf are special cases.

```

# math_isclose_inf.py

import math

print('nan, nan:', math.isclose(math.nan, math.nan))
print('nan, 1.0:', math.isclose(math.nan, 1.0))
print('inf, inf:', math.isclose(math.inf, math.inf))
print('inf, 1.0:', math.isclose(math.inf, 1.0))

```

nan is never close to another value, including itself. inf is only close to itself.

```
$ python3 math_isclose_inf.py
```

```

nan, nan: False
nan, 1.0: False
inf, inf: True
inf, 1.0: False

```

## Converting Floating Point Values to Integers

The math module includes three functions for converting floating point values to whole numbers. Each takes a different approach, and will be useful in different circumstances.

The simplest is `trunc()`, which truncates the digits following the decimal, leaving only the significant digits making up the whole number portion of the value. `floor()` converts its input to the largest preceding integer, and `ceil()` (ceiling) produces the largest integer following sequentially after the input value.

```

# math_integers.py

import math

HEADINGS = ('i', 'int', 'trunk', 'floor', 'ceil')
print('{:^5} {:^5} {:^5} {:^5} {:^5}'.format(*HEADINGS))
print('{:-^5} {:-^5} {:-^5} {:-^5} {:-^5}'.format(
    '_', '_', '_', '_', '_'
))

fmt = '{:5.1f} {:5.1f} {:5.1f} {:5.1f} {:5.1f}'

TEST_VALUES = [

```

```

-1.5,
-0.8,
-0.5,
-0.2,
0,
0.2,
0.5,
0.8,
1,
]

for i in TEST_VALUES:
    print(fmt.format(
        i,
        int(i),
        math.trunc(i),
        math.floor(i),
        math.ceil(i),
    ))

```

`trunc()` is equivalent to converting to `int` directly.

```
$ python3 math_integers.py
```

i	int	trunk	floor	ceil
-1.5	-1.0	-1.0	-2.0	-1.0
-0.8	0.0	0.0	-1.0	0.0
-0.5	0.0	0.0	-1.0	0.0
-0.2	0.0	0.0	-1.0	0.0
0.0	0.0	0.0	0.0	0.0
0.2	0.0	0.0	0.0	1.0
0.5	0.0	0.0	0.0	1.0
0.8	0.0	0.0	0.0	1.0
1.0	1.0	1.0	1.0	1.0

## Alternate Representations of Floating Point Values

`modf()` takes a single floating point number and returns a tuple containing the fractional and whole number parts of the input value.

```

# math_modf.py

import math

for i in range(6):
    print('{} / 2 = {}'.format(i, math.modf(i / 2.0)))

```

Both numbers in the return value are floats.

```
$ python3 math_modf.py
```

```

0/2 = (0.0, 0.0)
1/2 = (0.5, 0.0)
2/2 = (0.0, 1.0)
3/2 = (0.5, 1.0)
4/2 = (0.0, 2.0)
5/2 = (0.5, 2.0)

```

`frexp()` returns the mantissa and exponent of a floating point number, and can be used to create a more portable representation of the value.

```

# math_frexp.py

import math

print('{:^7} {:^7} {:^7}'.format('x', 'm', 'e'))
print('{:~7} {:~7} {:~7}'.format(' ', ' ', ' '))

```

```
for x in [0.1, 0.5, 4.0]:
    m, e = math.frexp(x)
    print('{:7.2f} {:7.2f} {:7d}'.format(x, m, e))
```

`frexp()` uses the formula  $x = m * 2^{**}e$ , and returns the values `m` and `e`.

```
$ python3 math_frexp.py
```

x	m	e
0.10	0.80	-3
0.50	0.50	0
4.00	0.50	3

`ldexp()` is the inverse of `frexp()`.

```
# math_ldexp.py

import math

print('{:^7} {:^7} {:^7}'.format('m', 'e', 'x'))
print('{:~7} {:~7} {:~7}'.format('', '', ''))

INPUTS = [
    (0.8, -3),
    (0.5, 0),
    (0.5, 3),
]

for m, e in INPUTS:
    x = math.ldexp(m, e)
    print('{:7.2f} {:7d} {:7.2f}'.format(m, e, x))
```

Using the same formula as `frexp()`, `ldexp()` takes the mantissa and exponent values as arguments and returns a floating point number.

```
$ python3 math_ldexp.py
```

m	e	x
0.80	-3	0.10
0.50	0	0.50
0.50	3	4.00

## Positive and Negative Signs

The absolute value of a number is its value without a sign. Use `fabs()` to calculate the absolute value of a floating point number.

```
# math_fabs.py

import math

print(math.fabs(-1.1))
print(math.fabs(-0.0))
print(math.fabs(0.0))
print(math.fabs(1.1))
```

In practical terms, the absolute value of a float is represented as a positive value.

```
$ python3 math_fabs.py
```

```
1.1
0.0
0.0
1.1
```

To determine the sign of a value, either to give a set of values the same sign or to compare two values, use `copysign()` to set the sign of a known good value.

```
# math_copysign.py

import math

HEADINGS = ('f', 's', '< 0', '> 0', '= 0')
print('{:^5} {:^5} {:^5} {:^5} {:^5}'.format(*HEADINGS))
print('{:~^5} {:~^5} {:~^5} {:~^5} {:~^5}'.format(
    ' ', ' ', ' ', ' ', ' ',
))

VALUES = [
    -1.0,
    0.0,
    1.0,
    float('-inf'),
    float('inf'),
    float('-nan'),
    float('nan'),
]

for f in VALUES:
    s = int(math.copysign(1, f))
    print('{:5.1f} {:5d} {!s:5} {!s:5} {!s:5}'.format(
        f, s, f < 0, f > 0, f == 0,
    ))
```

An extra function like `copysign()` is needed because comparing `nan` and `-nan` directly with other values does not work.

```
$ python3 math_copysign.py

  f      s    < 0    > 0    = 0
-----
-1.0    -1  True   False  False
 0.0     1  False  False   True
 1.0     1  False   True  False
-inf    -1  True   False  False
 inf     1  False   True  False
 nan    -1  False  False  False
 nan     1  False  False  False
```

## Commonly Used Calculations

Representing precise values in binary floating point memory is challenging. Some values cannot be represented exactly, and the more often a value is manipulated through repeated calculations, the more likely a representation error will be introduced. `math` includes a function for computing the sum of a series of floating point numbers using an efficient algorithm that minimizes such errors.

```
# math_fsum.py

import math

values = [0.1] * 10

print('Input values:', values)

print('sum()          : {:.20f}'.format(sum(values)))

s = 0.0
for i in values:
    s += i
print('for-loop       : {:.20f}'.format(s))

print('math.fsum()    : {:.20f}'.format(math.fsum(values)))
```

Given a sequence of ten values, each equal to `0.1`, the expected value for the sum of the sequence is `1.0`. Since `0.1` cannot be represented exactly as a floating point value, however, errors are introduced into the sum unless it is calculated with

represented exactly as a floating point value, however, errors are introduced into the sum unless it is calculated with fsum().

```
$ python3 math_fsum.py
```

```
Input values: [0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1]
sum()         : 0.99999999999999988898
for-loop      : 0.99999999999999988898
math.fsum()   : 1.00000000000000000000
```

factorial() is commonly used to calculate the number of permutations and combinations of a series of objects. The factorial of a positive integer  $n$ , expressed  $n!$ , is defined recursively as  $(n-1)! * n$  and stops with  $0! == 1$ .

```
# math_factorial.py
```

```
import math
```

```
for i in [0, 1.0, 2.0, 3.0, 4.0, 5.0, 6.1]:
    try:
        print('{:2.0f} {:6.0f}'.format(i, math.factorial(i)))
    except ValueError as err:
        print('Error computing factorial({}): {}'.format(i, err))
```

factorial() only works with whole numbers, but does accept float arguments as long as they can be converted to an integer without losing value.

```
$ python3 math_factorial.py
```

```
0      1
1      1
2      2
3      6
4     24
5    120
```

```
Error computing factorial(6.1): factorial() only accepts integral
values
```

gamma() is like factorial(), except it works with real numbers and the value is shifted down by one (gamma is equal to  $(n - 1)!$ ).

```
# math_gamma.py
```

```
import math
```

```
for i in [0, 1.1, 2.2, 3.3, 4.4, 5.5, 6.6]:
    try:
        print('{:2.1f} {:6.2f}'.format(i, math.gamma(i)))
    except ValueError as err:
        print('Error computing gamma({}): {}'.format(i, err))
```

Since zero causes the start value to be negative, it is not allowed.

```
$ python3 math_gamma.py
```

```
Error computing gamma(0): math domain error
1.1    0.95
2.2    1.10
3.3    2.68
4.4   10.14
5.5   52.34
6.6  344.70
```

lgamma() returns the natural logarithm of the absolute value of gamma for the input value.

```
# math_lgamma.py
```

```
import math
```

```
for i in [0, 1.1, 2.2, 3.3, 4.4, 5.5, 6.6]:
```



```

for i in [0, 1.1, 2.2, 3.3, 4.4, 5.5, 6.6]:
    try:
        print('{:2.1f} {:.20f} {:.20f}'.format(
            i,
            math.lgamma(i),
            math.log(math.gamma(i)),
        ))
    except ValueError as err:
        print('Error computing lgamma({}): {}'.format(i, err))

```

Using `lgamma()` retains more precision than calculating the logarithm separately using the results of `gamma()`.

```

$ python3 math_lgamma.py

Error computing lgamma(0): math domain error
1.1 -0.04987244125984036103 -0.04987244125983997245
2.2 0.09694746679063825923 0.09694746679063866168
3.3 0.98709857789473387513 0.98709857789473409717
4.4 2.31610349142485727469 2.31610349142485727469
5.5 3.95781396761871651080 3.95781396761871606671
6.6 5.84268005527463252236 5.84268005527463252236

```

The modulo operator (%) computes the remainder of a division expression (i.e.,  $5 \% 2 = 1$ ). The operator built into the language works well with integers but, as with so many other floating point operations, intermediate calculations cause representational issues that result in a loss of data. `fmod()` provides a more accurate implementation for floating point values.

```

# math_fmod.py

import math

print('{:^4} {:^4} {:^5} {:^5}'.format(
    'x', 'y', '%', 'fmod'))
print('{:~4} {:~4} {:~5} {:~5}'.format(
    '-', '-', '-', '-'))

INPUTS = [
    (5, 2),
    (5, -2),
    (-5, 2),
]

for x, y in INPUTS:
    print('{:4.1f} {:4.1f} {:5.2f} {:5.2f}'.format(
        x,
        y,
        x % y,
        math.fmod(x, y),
    ))

```

A potentially more frequent source of confusion is the fact that the algorithm used by `fmod()` for computing modulo is also different from that used by `%`, so the sign of the result is different.

```

$ python3 math_fmod.py

x      y      %      fmod
-----
5.0    2.0    1.00    1.00
5.0   -2.0   -1.00    1.00
-5.0    2.0    1.00   -1.00

```

Use `gcd()` to find the largest integer that can divide evenly into two integers, the greatest common divisor.

```

# math_gcd.py

import math

print(math.gcd(10, 8))
print(math.gcd(10, 0))
print(math.gcd(50, 225))
print(math.gcd(11, 0))

```

```
print(math.gcd(11, 9))
print(math.gcd(0, 0))
```

If both values are 0, the result is 0.

```
$ python3 math_gcd.py

2
10
25
1
0
```

## Exponents and Logarithms

Exponential growth curves appear in economics, physics, and other sciences. Python has a built-in exponentiation operator (`**`), but `pow()` can be useful when a callable function is needed as an argument to another function.

```
# math_pow.py

import math

INPUTS = [
    # Typical uses
    (2, 3),
    (2.1, 3.2),

    # Always 1
    (1.0, 5),
    (2.0, 0),

    # Not-a-number
    (2, float('nan')),

    # Roots
    (9.0, 0.5),
    (27.0, 1.0 / 3),
]

for x, y in INPUTS:
    print('{:5.1f} ** {:5.3f} = {:6.3f}'.format(
        x, y, math.pow(x, y)))
```

Raising 1 to any power always returns 1.0, as does raising any value to a power of 0.0. Most operations on the not-a-number value nan return nan. If the exponent is less than 1, `pow()` computes a root.

```
$ python3 math_pow.py

2.0 ** 3.000 = 8.000
2.1 ** 3.200 = 10.742
1.0 ** 5.000 = 1.000
2.0 ** 0.000 = 1.000
2.0 ** nan = nan
9.0 ** 0.500 = 3.000
27.0 ** 0.333 = 3.000
```

Since square roots (exponent of 1/2) are used so frequently, there is a separate function for computing them.

```
# math_sqrt.py

import math

print(math.sqrt(9.0))
print(math.sqrt(3))
try:
    print(math.sqrt(-1))
except ValueError as err:
    print('Cannot compute sqrt(-1):', err)
```

Computing the square roots of negative numbers requires *complex numbers*, which are not handled by `math`. Any attempt to calculate a square root of a negative value results in a `ValueError`.

```
$ python3 math_sqrt.py
3.0
1.7320508075688772
Cannot compute sqrt(-1): math domain error
```

The logarithm function finds  $y$  where  $x = b ** y$ . By default, `log()` computes the natural logarithm (the base is  $e$ ). If a second argument is provided, that value is used as the base.

```
# math_log.py

import math

print(math.log(8))
print(math.log(8, 2))
print(math.log(0.5, 2))
```

Logarithms where  $x$  is less than one yield negative results.

```
$ python3 math_log.py
2.0794415416798357
3.0
-1.0
```

There are three variations of `log()`. Given floating point representation and rounding errors, the computed value produced by `log(x, b)` has limited accuracy, especially for some bases. `log10()` computes `log(x, 10)`, using a more accurate algorithm than `log()`.

```
# math_log10.py

import math

print('{:2} {:^12} {:^10} {:^20} {:8}'.format(
    'i', 'x', 'accurate', 'inaccurate', 'mismatch',
))
print('{:~2} {:~12} {:~10} {:~20} {:~8}'.format(
    '', '', '', '', '',
))

for i in range(0, 10):
    x = math.pow(10, i)
    accurate = math.log10(x)
    inaccurate = math.log(x, 10)
    match = '' if int(inaccurate) == i else '*'
    print('{:2d} {:12.1f} {:10.8f} {:20.18f} {:^5}'.format(
        i, x, accurate, inaccurate, match,
    ))
```

The lines in the output with trailing `*` highlight the inaccurate values.

```
$ python3 math_log10.py
```

i	x	accurate	inaccurate	mismatch
0	1.0	0.00000000	0.000000000000000000	
1	10.0	1.00000000	1.000000000000000000	
2	100.0	2.00000000	2.000000000000000000	
3	1000.0	3.00000000	2.99999999999999956	*
4	10000.0	4.00000000	4.000000000000000000	
5	100000.0	5.00000000	5.000000000000000000	
6	1000000.0	6.00000000	5.999999999999999112	*
7	10000000.0	7.00000000	7.000000000000000000	
8	100000000.0	8.00000000	8.000000000000000000	
9	1000000000.0	9.00000000	8.999999999999998224	*

```
# math_log2.py

import math

print('{: >2} {: ^5} {: ^5}'.format(
    'i', 'x', 'log2',
))

print('{: -^2} {: -^5} {: -^5}'.format(
    'i', 'x', 'log2',
))

for i in range(0, 10):
    x = math.pow(2, i)
    result = math.log2(x)
    print('{: 2d} {: 5.1f} {: 5.1f}'.format(
        i, x, result,
    ))
```

Depending on the underlying platform, using the built-in and special-purpose function can offer better performance and accuracy by using special-purpose algorithms for base 2 that are not found in the more general purpose function.

```
$ python3 math_log2.py
```

i	x	log2
0	1.0	0.0
1	2.0	1.0
2	4.0	2.0
3	8.0	3.0
4	16.0	4.0
5	32.0	5.0
6	64.0	6.0
7	128.0	7.0
8	256.0	8.0
9	512.0	9.0

`log1p()` calculates the Newton-Mercator series (the natural logarithm of  $1+x$ ).

[illegible]

`log1p()` is more accurate for values of `x` very close to zero because it uses an algorithm that compensates for round-off errors from the initial addition.

```
$ python3 math_log1p.py
```

```
x      : 1e-25
1 + x  : 1.0
log(1+x): 0.0
log1p(x): 1e-25
```

`exp()` computes the exponential function ( $e^x$ ).

```
# math exp.py
```

```
import math
```

$$x = 2$$

with other special-case functions, it uses an algorithm that produces more accurate results than the general-purpose `math.pow(math.e, x)`.

`expm1()` is the inverse of `log1p()`, and calculates  $e^x - 1$ .

Small values of  $x$  lose precision when the subtraction is performed separately, like with `log1p()`.

## Angles

The circumference is calculated as  $2\pi r$ , so there is a relationship between radians and  $\pi$ , a value that shows up frequently in trigonometric calculations. That relationship leads to radians being used in trigonometry and calculus, because they result in more compact formulas.

```
# math_radians.py

import math

print('{:^7} {:^7} {:^7}'.format(
    'Degrees', 'Radians', 'Expected'))
print('{:~7} {:~7} {:~7}'.format(
    '', '', ''))

INPUTS = [
    (0, 0),
    (30, math.pi / 6),
    (45, math.pi / 4),
    (60, math.pi / 3),
    (90, math.pi / 2),
    (180, math.pi),
    (270, 3 / 2.0 * math.pi),
    (360, 2 * math.pi),
]

for deg, expected in INPUTS:
    print('{:7d} {:7.2f} {:7.2f}'.format(
        deg, math.radians(deg), expected))
```

```

    print('{:^8} {:^8} {:^8}'.format(
        deg,
        math.radians(deg),
        expected,
    ))

```

The formula for the conversion is  $\text{rad} = \text{deg} * \pi / 180$ .

```
$ python3 math_radians.py
```

Degrees	Radians	Expected
0	0.00	0.00
30	0.52	0.52
45	0.79	0.79
60	1.05	1.05
90	1.57	1.57
180	3.14	3.14
270	4.71	4.71
360	6.28	6.28

To convert from radians to degrees, use `degrees()`.

```
# math_degrees.py
```

```
import math
```

```

INPUTS = [
    (0, 0),
    (math.pi / 6, 30),
    (math.pi / 4, 45),
    (math.pi / 3, 60),
    (math.pi / 2, 90),
    (math.pi, 180),
    (3 * math.pi / 2, 270),
    (2 * math.pi, 360),
]

print('{:^8} {:^8} {:^8}'.format(
    'Radians', 'Degrees', 'Expected'))
print('{:-^8} {:-^8} {:-^8}'.format('', '', ''))
for rad, expected in INPUTS:
    print('{:8.2f} {:8.2f} {:8.2f}'.format(
        rad,
        math.degrees(rad),
        expected,
    ))

```

The formula is  $\text{deg} = \text{rad} * 180 / \pi$ .

```
$ python3 math_degrees.py
```

Radians	Degrees	Expected
0.00	0.00	0.00
0.52	30.00	30.00
0.79	45.00	45.00
1.05	60.00	60.00
1.57	90.00	90.00
3.14	180.00	180.00
4.71	270.00	270.00
6.28	360.00	360.00

## Trigonometry

Trigonometric functions relate angles in a triangle to the lengths of its sides. They show up in formulas with periodic properties such as harmonics, circular motion, or when dealing with angles. All of the trigonometric functions in the standard library take angles expressed as radians.

Given an angle in a right triangle, the *sine* is the ratio of the length of the side opposite the angle to the hypotenuse ( $\sin A =$

Given an angle in a right triangle, the *sine* is the ratio of the length of the side opposite the angle to the hypotenuse ( $\sin A = \text{opposite/hypotenuse}$ ). The *cosine* is the ratio of the length of the adjacent side to the hypotenuse ( $\cos A = \text{adjacent/hypotenuse}$ ). And the *tangent* is the ratio of the opposite side to the adjacent side ( $\tan A = \text{opposite/adjacent}$ ).

```
# math_trig.py

import math

print('{:^7} {:^7} {:^7} {:^7} {:^7}'.format(
    'Degrees', 'Radians', 'Sine', 'Cosine', 'Tangent'))
print('{:~^7} {:~^7} {:~^7} {:~^7} {:~^7}'.format(
    '-', '-', '-', '-', '-'))

fmt = '{:7.2f} {:7.2f} {:7.2f} {:7.2f} {:7.2f}'

for deg in range(0, 361, 30):
    rad = math.radians(deg)
    if deg in (90, 270):
        t = float('inf')
    else:
        t = math.tan(rad)
    print(fmt.format(deg, rad, math.sin(rad), math.cos(rad), t))
```

The tangent can also be defined as the ratio of the sine of the angle to its cosine, and since the cosine is 0 for  $\pi/2$  and  $3\pi/2$  radians, the tangent is infinite.

```
$ python3 math_trig.py
```

Degrees	Radians	Sine	Cosine	Tangent
0.00	0.00	0.00	1.00	0.00
30.00	0.52	0.50	0.87	0.58
60.00	1.05	0.87	0.50	1.73
90.00	1.57	1.00	0.00	inf
120.00	2.09	0.87	-0.50	-1.73
150.00	2.62	0.50	-0.87	-0.58
180.00	3.14	0.00	-1.00	-0.00
210.00	3.67	-0.50	-0.87	0.58
240.00	4.19	-0.87	-0.50	1.73
270.00	4.71	-1.00	-0.00	inf
300.00	5.24	-0.87	0.50	-1.73
330.00	5.76	-0.50	0.87	-0.58
360.00	6.28	-0.00	1.00	-0.00

Given a point  $(x, y)$ , the length of the hypotenuse for the triangle between the points  $[(0, 0), (x, 0), (x, y)]$  is  $(x^2 + y^2)^{1/2}$ , and can be computed with `hypot()`.

```
# math_hypot.py

import math

print('{:^7} {:^7} {:^10}'.format('X', 'Y', 'Hypotenuse'))
print('{:~^7} {:~^7} {:~^10}'.format('-', '-', '-'))

POINTS = [
    # simple points
    (1, 1),
    (-1, -1),
    (math.sqrt(2), math.sqrt(2)),
    (3, 4), # 3-4-5 triangle
    # on the circle
    (math.sqrt(2) / 2, math.sqrt(2) / 2), # pi/4 rads
    (0.5, math.sqrt(3) / 2), # pi/3 rads
]

for x, y in POINTS:
    h = math.hypot(x, y)
    print('{:7.2f} {:7.2f} {:7.2f}'.format(x, y, h))
```

Points on the circle always have hypotenuse equal to 1.

```
$ python3 math_hypot.py
```

X	Y	Hypotenuse
1.00	1.00	1.41
-1.00	-1.00	1.41
1.41	1.41	2.00
3.00	4.00	5.00
0.71	0.71	1.00
0.50	0.87	1.00

The same function can be used to find the distance between two points.

```
# math_distance_2_points.py
```

```
import math
```

```
print('{:^8} {:^8} {:^8} {:^8} {:^8}'.format(
    'X1', 'Y1', 'X2', 'Y2', 'Distance',
))
print('{:~8} {:~8} {:~8} {:~8} {:~8}'.format(
    ' ', ' ', ' ', ' ', ' ',
))
```

```
POINTS = [
    ((5, 5), (6, 6)),
    ((-6, -6), (-5, -5)),
    ((0, 0), (3, 4)), # 3-4-5 triangle
    ((-1, -1), (2, 3)), # 3-4-5 triangle
]
```

```
for (x1, y1), (x2, y2) in POINTS:
    x = x1 - x2
    y = y1 - y2
    h = math.hypot(x, y)
    print('{:8.2f} {:8.2f} {:8.2f} {:8.2f} {:8.2f}'.format(
        x1, y1, x2, y2, h,
    ))
```

Use the difference in the x and y values to move one endpoint to the origin, and then pass the results to `hypot()`.

```
$ python3 math_distance_2_points.py
```

X1	Y1	X2	Y2	Distance
5.00	5.00	6.00	6.00	1.41
-6.00	-6.00	-5.00	-5.00	1.41
0.00	0.00	3.00	4.00	5.00
-1.00	-1.00	2.00	3.00	5.00

`math` also defines inverse trigonometric functions.

```
# math_inverse_trig.py
```

```
import math
```

```
for r in [0, 0.5, 1]:
    print('arcsine({:.1f}) = {:.5.2f}'.format(r, math.asin(r)))
    print('arccosine({:.1f}) = {:.5.2f}'.format(r, math.acos(r)))
    print('arctangent({:.1f}) = {:.5.2f}'.format(r, math.atan(r)))
    print()
```

1.57 is roughly equal to  $\pi/2$ , or 90 degrees, the angle at which the sine is 1 and the cosine is 0.

```
$ python3 math_inverse_trig.py
```

```
arcsine(0.0) = 0.00
arccosine(0.0) = 1.57
```



```

arcsine(0.0) = 0.00
arctangent(0.0) = 0.00

arcsine(0.5) = 0.52
arccosine(0.5) = 1.05
arctangent(0.5) = 0.46

arcsine(1.0) = 1.57
arccosine(1.0) = 0.00
arctangent(1.0) = 0.79

```

## Hyperbolic Functions

Hyperbolic functions appear in linear differential equations and are used when working with electromagnetic fields, fluid dynamics, special relativity, and other advanced physics and mathematics.

```

# math_hyperbolic.py

import math

print('{:^6} {:^6} {:^6} {:^6}'.format(
    'X', 'sinh', 'cosh', 'tanh',
))
print('{:-^6} {:-^6} {:-^6} {:-^6}'.format(' ', ' ', ' ', ' '))

fmt = '{:6.4f} {:6.4f} {:6.4f} {:6.4f}'

for i in range(0, 11, 2):
    x = i / 10.0
    print(fmt.format(
        x,
        math.sinh(x),
        math.cosh(x),
        math.tanh(x),
    ))

```

Whereas the cosine and sine functions describe a circle, the hyperbolic cosine and hyperbolic sine form half of a hyperbola.

```

$ python3 math_hyperbolic.py

  X      sinh      cosh      tanh
-----
0.0000  0.0000  1.0000  0.0000
0.2000  0.2013  1.0201  0.1974
0.4000  0.4108  1.0811  0.3799
0.6000  0.6367  1.1855  0.5370
0.8000  0.8881  1.3374  0.6640
1.0000  1.1752  1.5431  0.7616

```

Inverse hyperbolic functions `acosh()`, `asinh()`, and `atanh()` are also available.

## Special Functions

The Gauss Error function is used in statistics.

```

# math_erf.py

import math

print('{:^5} {:7}'.format('x', 'erf(x)'))
print('{:-^5} {:-^7}'.format(' ', ' '))

for x in [-3, -2, -1, -0.5, -0.25, 0, 0.25, 0.5, 1, 2, 3]:
    print('{:5.2f} {:7.4f}'.format(x, math.erf(x)))

```

For the error function,  $\text{erf}(-x) == -\text{erf}(x)$ .

```

$ python3 math_erf.py

```

x	erf(x)
-3.00	-1.0000
-2.00	-0.9953
-1.00	-0.8427
-0.50	-0.5205
-0.25	-0.2763
0.00	0.0000
0.25	0.2763
0.50	0.5205
1.00	0.8427
2.00	0.9953
3.00	1.0000

The complimentary error function is  $1 - \text{erf}(x)$ .

```
# math_erfc.py

import math

print('{:^5} {:^7}'.format('x', 'erfc(x)'))
print('{:~^5} {:~^7}'.format('', ''))

for x in [-3, -2, -1, -0.5, -0.25, 0, 0.25, 0.5, 1, 2, 3]:
    print('{:5.2f} {:7.4f}'.format(x, math.erfc(x)))
```

The implementation of `erfc()` avoids precision errors for small values of `x` when subtracting from 1.

```
$ python3 math_erfc.py
```

x	erfc(x)
-3.00	2.0000
-2.00	1.9953
-1.00	1.8427
-0.50	1.5205
-0.25	1.2763
0.00	1.0000
0.25	0.7237
0.50	0.4795
1.00	0.1573
2.00	0.0047
3.00	0.0000

## See also

- [Standard library documentation for math](#)
- [IEEE floating point arithmetic in Python](#) – Blog post by John Cook about how special values arise and are dealt with when doing math in Python.
- [SciPy](#) – Open source libraries for scientific and mathematical calculations in Python.
- [PEP 485](#) – “A function for testing approximate equality”

Quick Links

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