

# Case study definition

Let's define a function with a complicated expression:

$$f: (x, y) \rightarrow \cos \left[ (xy + (x - 4)^2 + \arctan((y + 3)^2))x \right]$$

With  $x > 0$  and  $y > 0$ .

`sympy` will be used to compute partial derivatives of this function.

## Key idea of `sympy`

`sympy` can handle mathematical objects in a formal way, i.e. **without using numerical discretization but relying on the properties of the defined objects.**

With `sympy`, objects are defined with a mathematical meaning that goes beyond the software meaning of traditional Python code. Then some operations are performed on these objects:

- differentiation
- integration
- limits
- etc...

Code

## Mathematical objects declaration

Let's define two mathematical variables `x` and `y`. `sympy` is told these variables must take positive real values, using `positive=True`.

```
In [1]: import sympy as sym

x = sym.Symbol('x', positive=True)
y = sym.Symbol('y', positive=True)
```

Then the function is defined:

```
In [2]: f = sym.Lambda((x, y),
                        sym.cos((x*y + (x-4)**2 + sym.atan((y+3)**2))*x))
f
```

```
Out[2]: 
$$\left( (x, y) \mapsto \cos\left(x\left(xy + (x - 4)^2 + \operatorname{atan}\left((y + 3)^2\right)\right)\right)\right)$$

```

**Be careful!** Functions `cos`, `atan` and `log` are the one of `sympy` !

The function can be evaluated at a specific point:

```
In [3]: function_call = f(2, 4)
```

Yet, `evalf` must be used to get a numerical approximation. `evalf` returns a `sympy`-related type, it can be converted using `float`:

```
In [4]: print(type(function_call.evalf()))  
print(float(function_call.evalf()))
```

```
<class 'sympy.core.numbers.Float'>  
-0.3868788252007259
```

## Operations on defined objects

Let's compute the partial derivative of `f` with respect to `x`, using `diff`:

```
In [5]: f(x, y).diff(x)
```

```
Out[5]: -\left(xy + x(2x + y - 8) + (x - 4)^2 + \operatorname{atan}\left((y + 3)^2\right)\right)\sin\left(x\left(xy + (x - 4)^2 + \operatorname{atan}\left((y + 3)^2\right)\right)\right)
```

Or compute the second derivative with respect to  $x$  and then the first derivative with respect to  $y$ :

```
In [6]: der = f(x, y).diff(x, 2, y)
der
```

```
Out[6]:
```

$$-2x \left( x + \frac{2(y+3)}{(y+3)^4 + 1} \right) (3x + y - 8) \cos \left( x \left( xy + (x-4)^2 + \operatorname{atan}((y+3)^2) \right) \right) + x \left( x + \frac{2(y+3)}{(y+3)^4 + 1} \right) \left( xy + x(2x + y - 8) + ( \right.$$

---



In this expression,  $x$  and  $y$  are still unknown. Let's replace  $x$  using `subs` :

```
In [7]: der = der.subs({x: 5})  
der
```

```
Out[7]:
```

$$-10(y+7)\left(\frac{2(y+3)}{(y+3)^4+1}+5\right)\cos\left(25y+5\operatorname{atan}\left((y+3)^2\right)+5\right)-4\left(\frac{y+3}{(y+3)^4+1}+5\right)\left(10y+\operatorname{atan}\left((y+3)^2\right)+11\right)\cos(2$$

---

## numpy conversion

The evaluation of `der` is not fast. Thus it can be converted to a `numpy` function using `lambdify`. Then evaluation on arrays is possible:

```
In [8]: import numpy as np
        numpy_func = sym.lambdify(y, der)
        Y = np.linspace(0, 10, 5)
        numpy_func(Y)
```

```
Out[8]: array([ -1618.36732557,  -8543.34409677, -44619.86174987,
                -132966.53918376, -279318.28704865])
```

## When to use `sympy`

`sympy` can provide exact mathematical solutions **for simple problems only**. Thus, it can be used to check the results of some very specific calculation steps performed in a more numerical way.

