



Assignment 1 due tonight!

Assignment 2 out this afternoon



In Python, we can write code to work with data represented as:

*Integers*

42, -3, 10000

*Floating-point numbers*

0.0, -3.6, 4.2

*Booleans*

True, False

*Text strings*

"Alan Turing", "50%", "\$3.50"

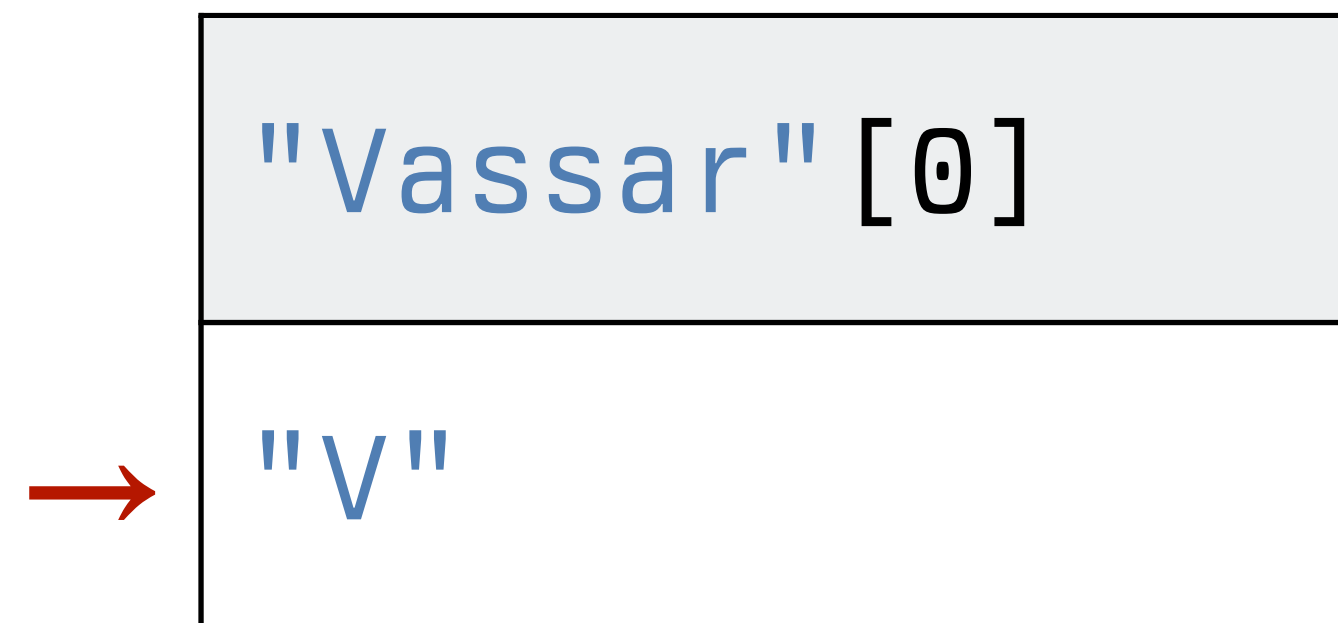


A text string is a *sequence* of *characters* – letters, numbers, punctuation.

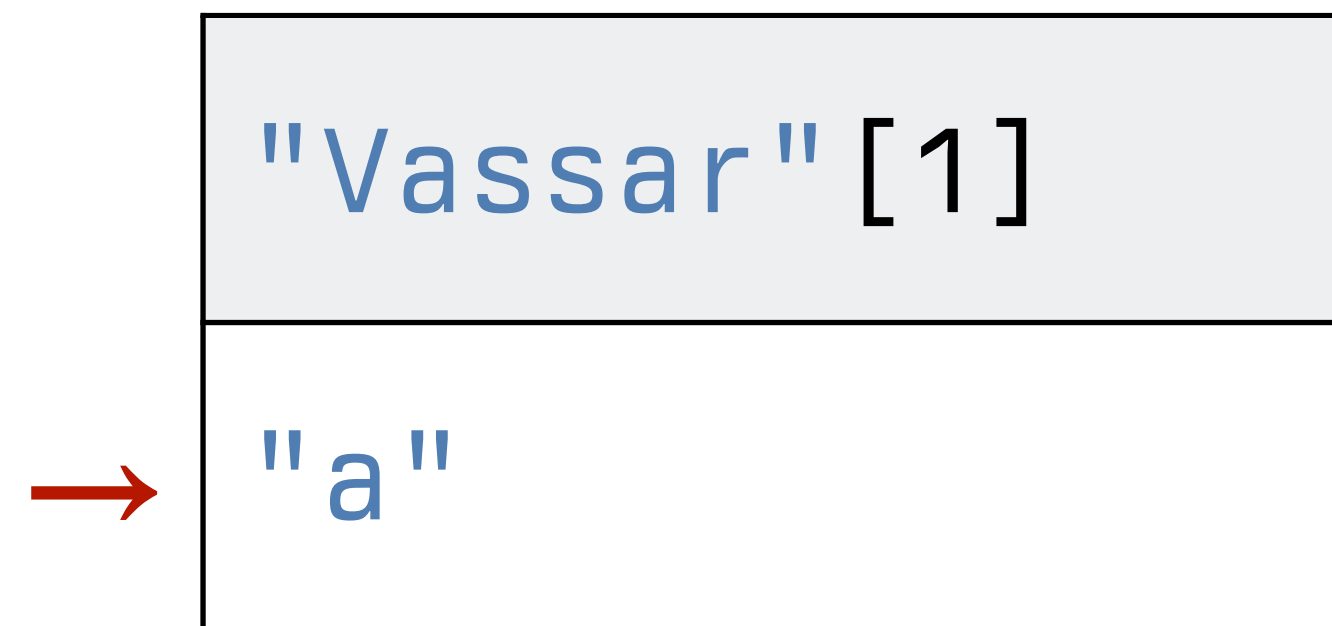
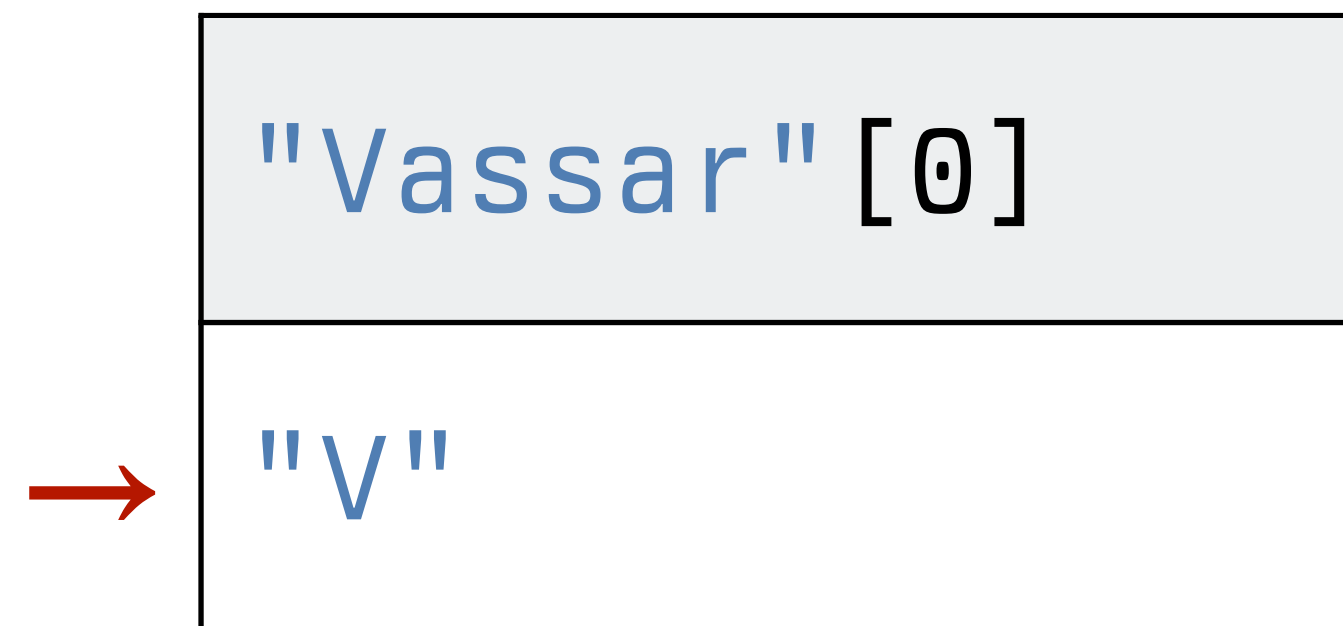
We can access these characters by their position in the sequence.

	0	1	2	3	4	5	
"	v	a	s	s	a	r	"

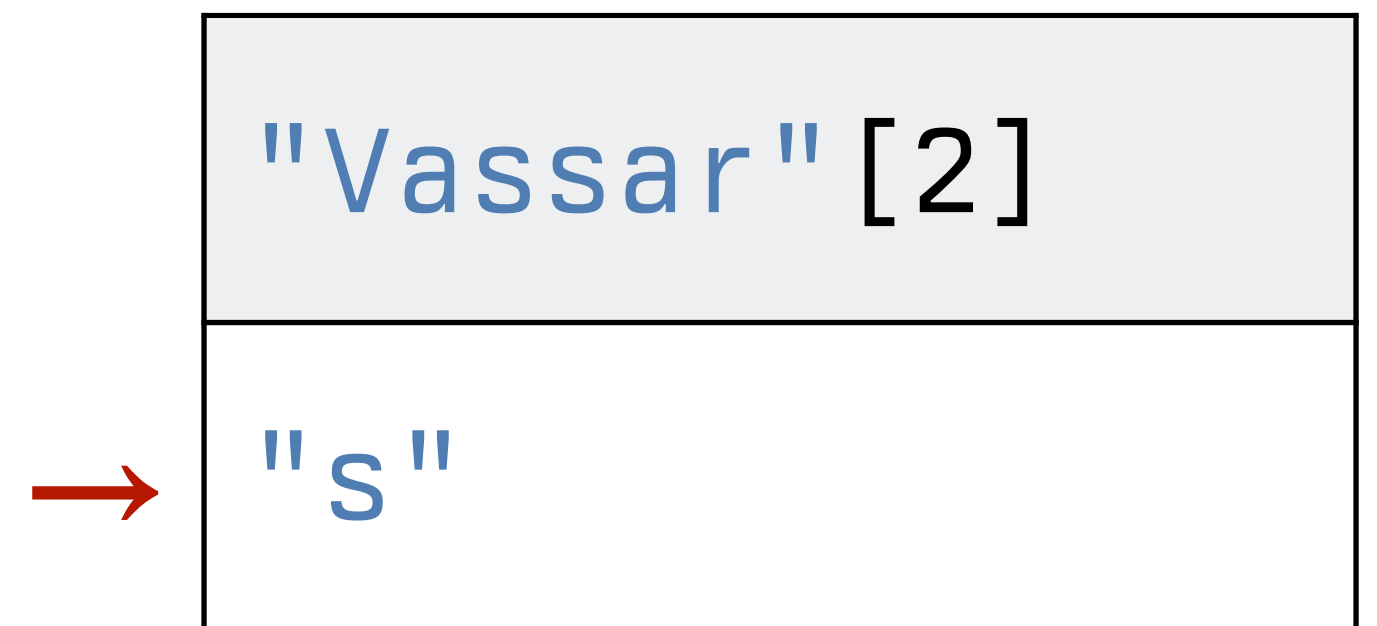
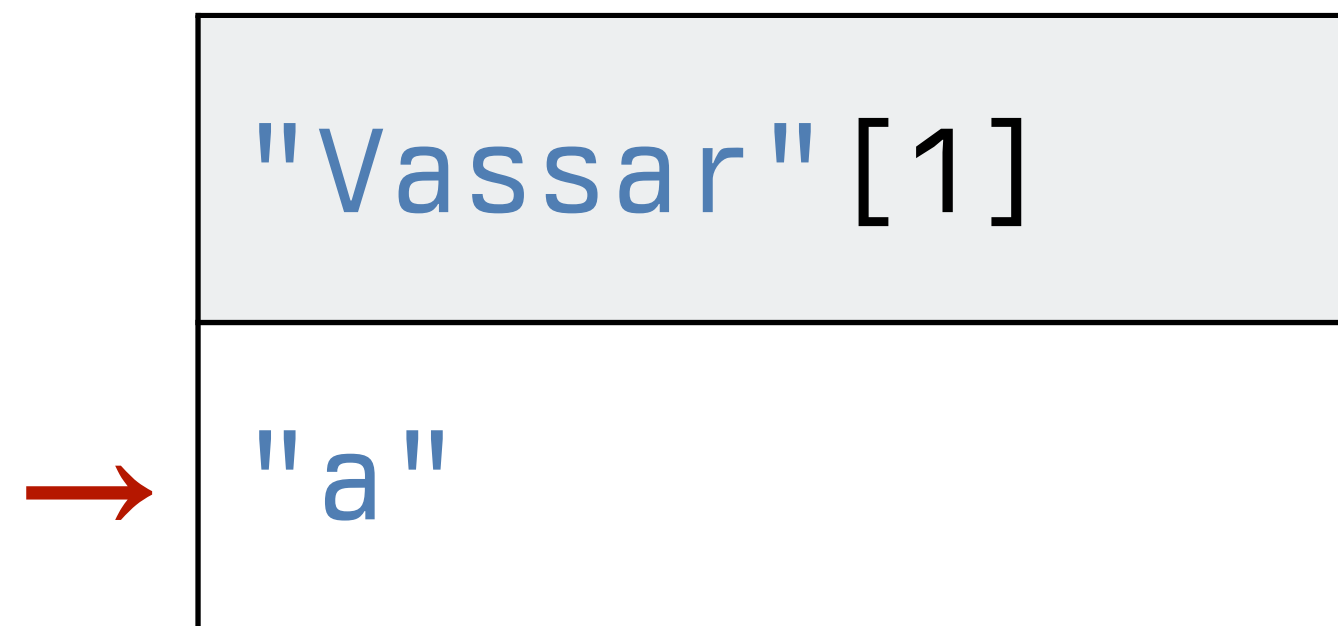
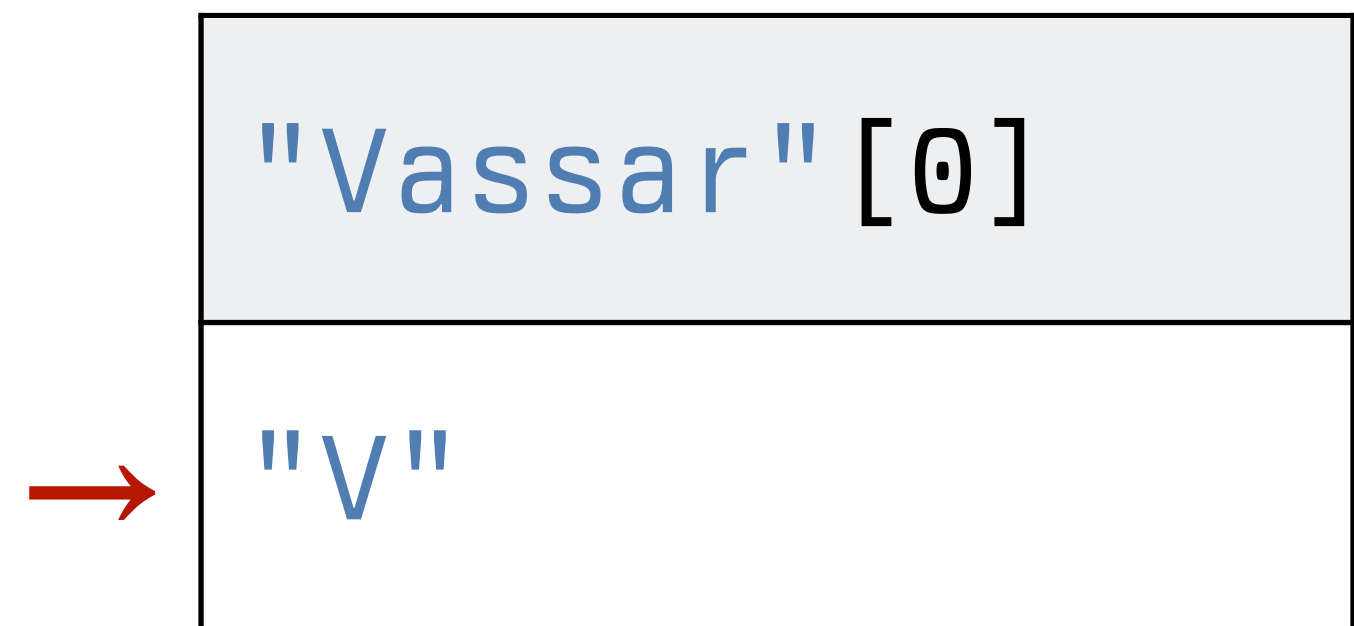
0	1	2	3	4	5		
"	V	a	s	s	a	r	"



0 | 1 | 2 | 3 | 4 | 5  
" v | a | s | s | a | r "



0 | 1 | 2 | 3 | 4 | 5  
" v | a | s | s | a | r "



	0	1	2	3	4	5	
"	v	a	s	s	a	r	"
	-6	-5	-4	-3	-2	-1	

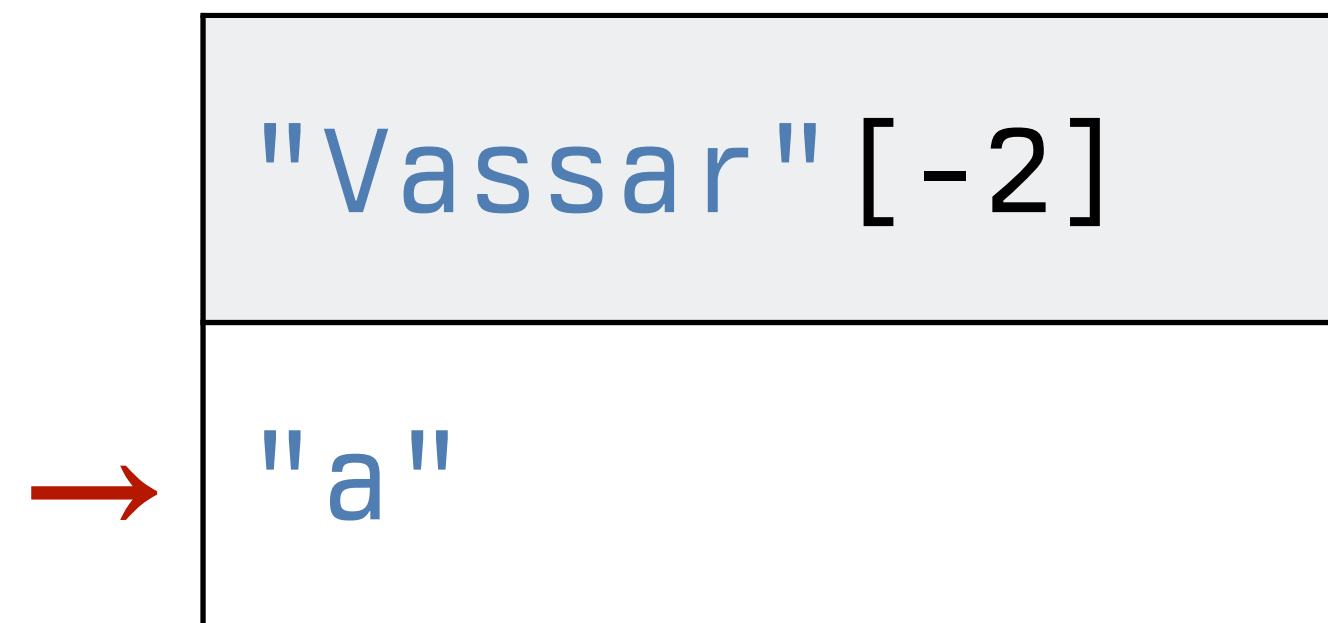
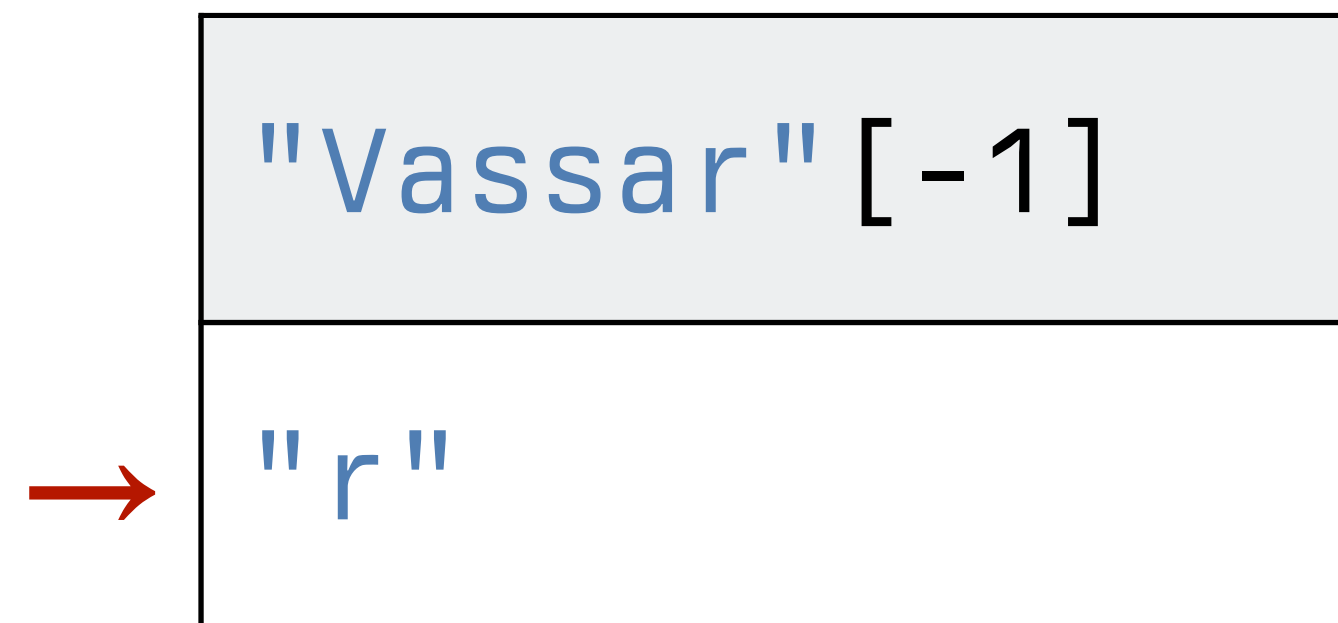
*We can use negative numbers to index from the end of the string.*

	0	1	2	3	4	5	
"	V	a	s	s	a	r	"
	-6	-5	-4	-3	-2	-1	

	"Vassar" [-1]
→	"r"

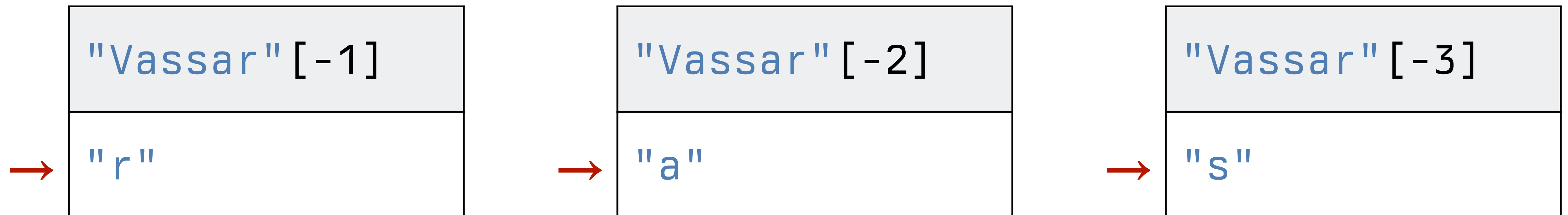
*We can use negative numbers to index from the end of the string.*

	0	1	2	3	4	5	
"	V	a	s	s	a	r	"
	-6	-5	-4	-3	-2	-1	



*We can use negative numbers to index from the end of the string.*

	0	1	2	3	4	5	
"	V	a	s	s	a	r	"
	-6	-5	-4	-3	-2	-1	



*We can use negative numbers to index from the end of the string.*

We can also *slice* a sequence to access a subsequence.

Instead of a single index, we specify

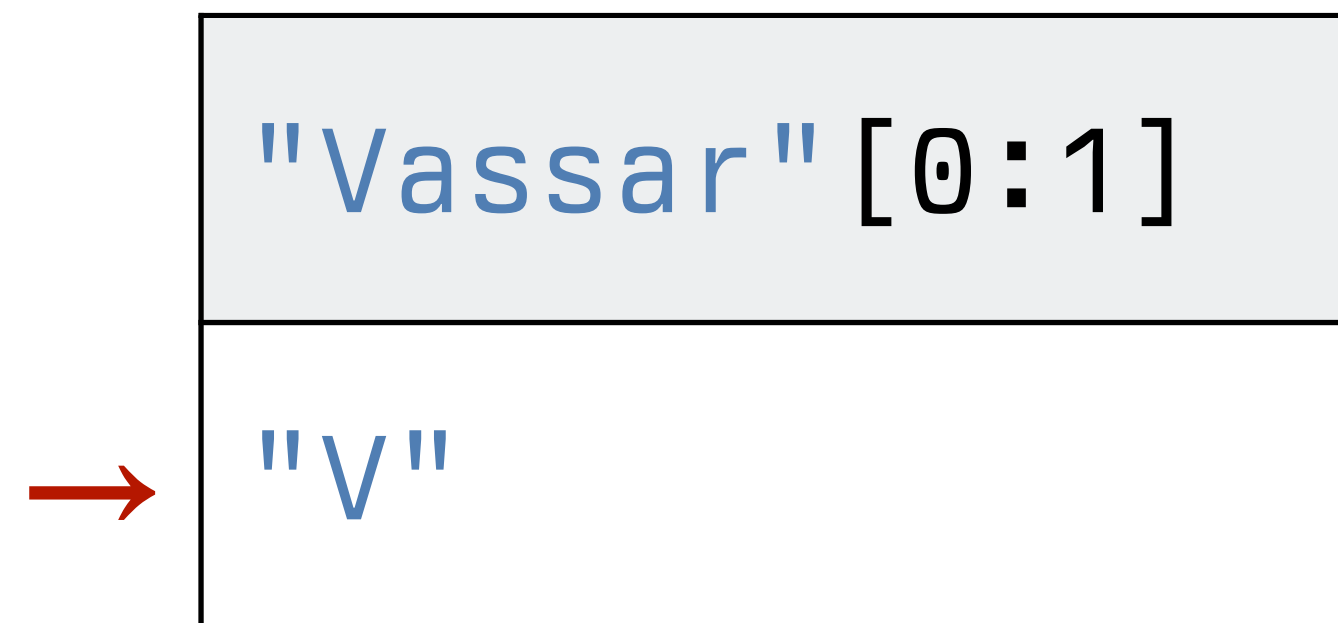
**[*start*:*end*]**

*the position to start  
at, including that  
character*

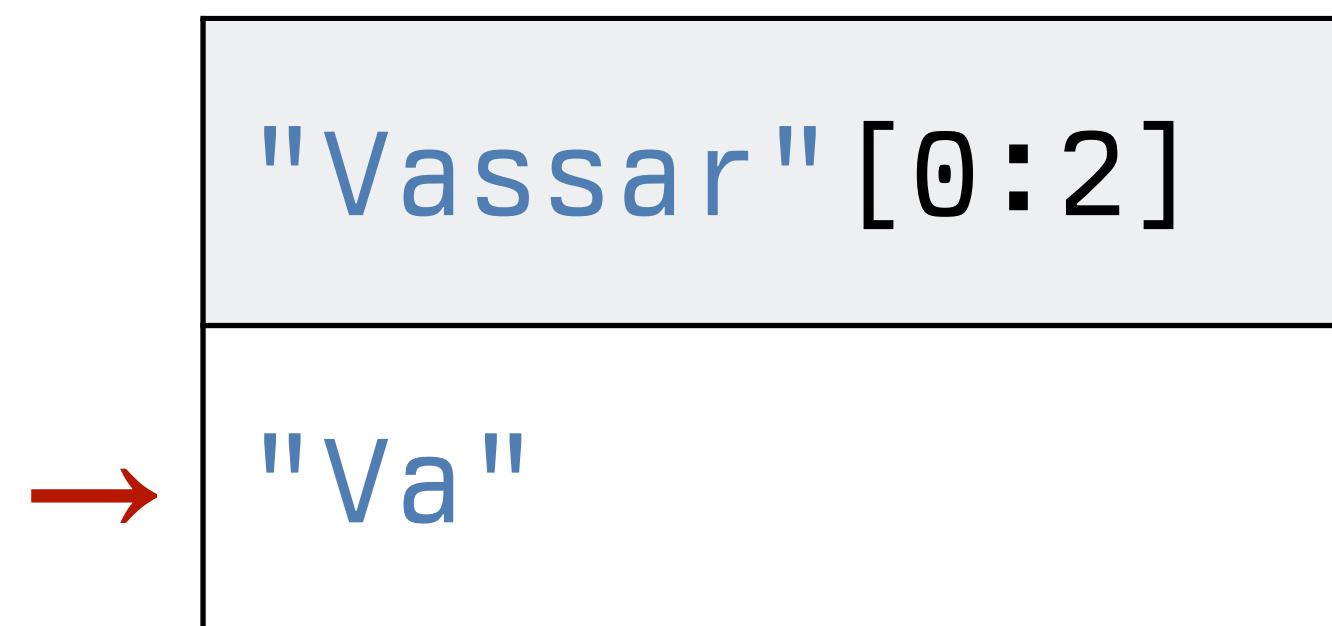
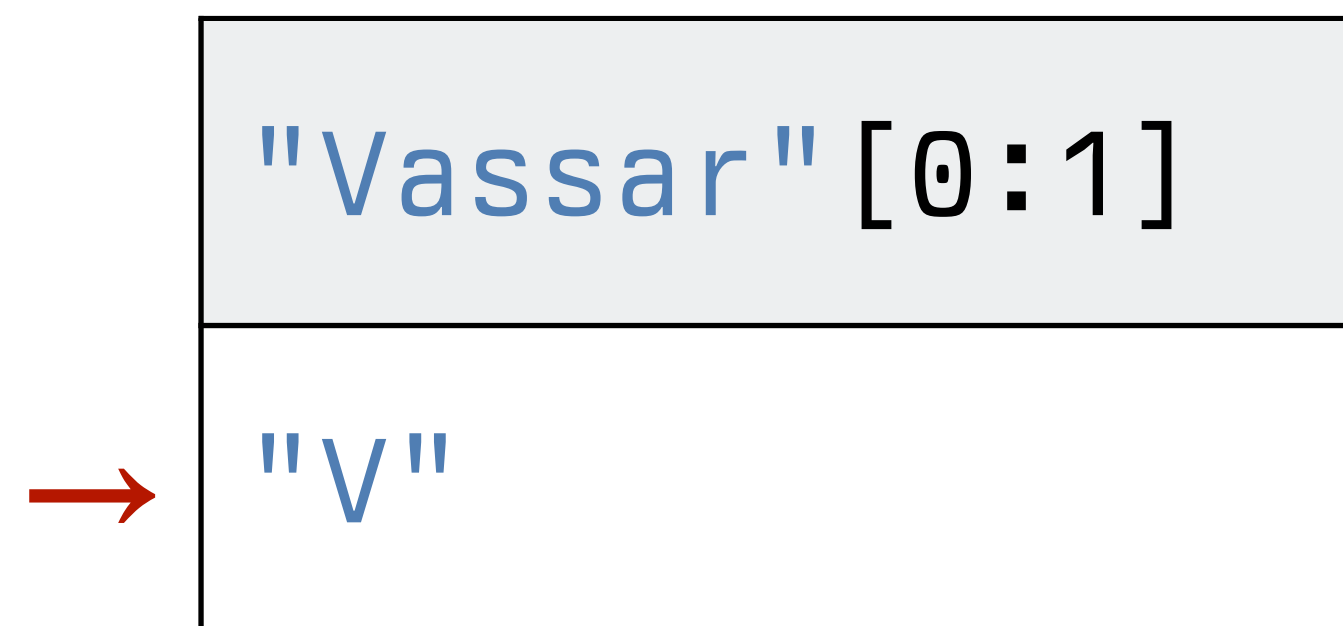
*the position to end  
before, excluding that  
character*

	0	1	2	3	4	5	
"	v	a	s	s	a	r	"

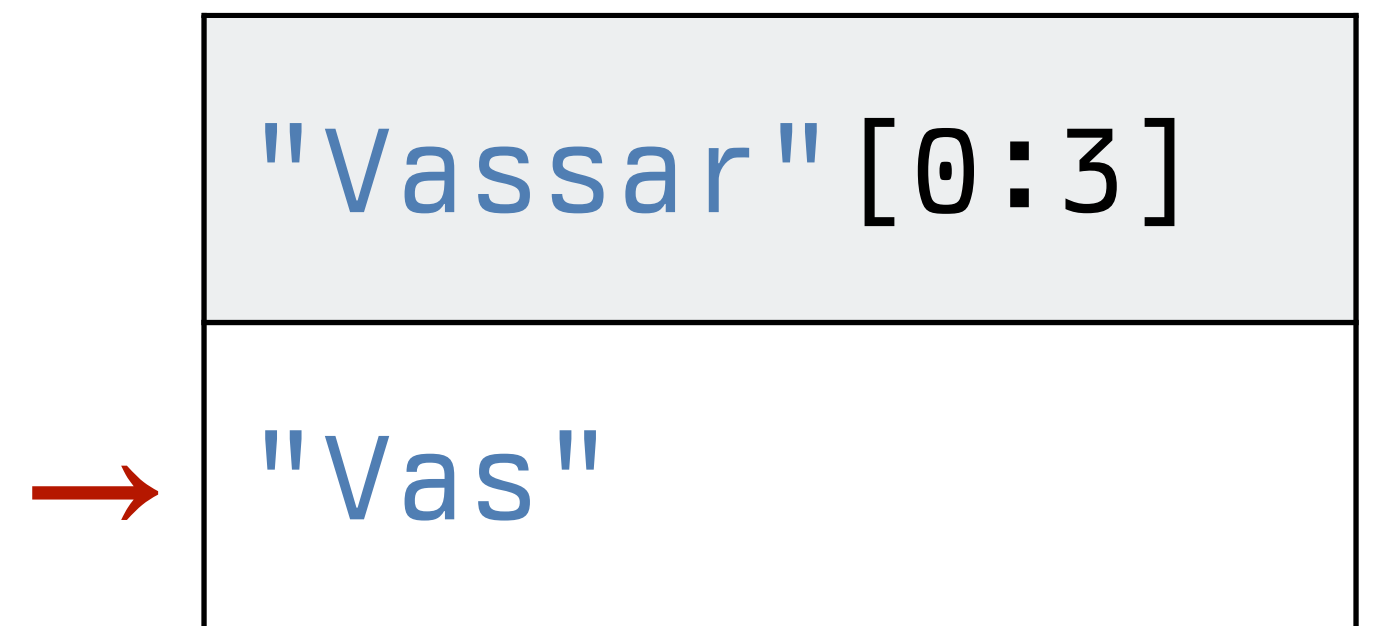
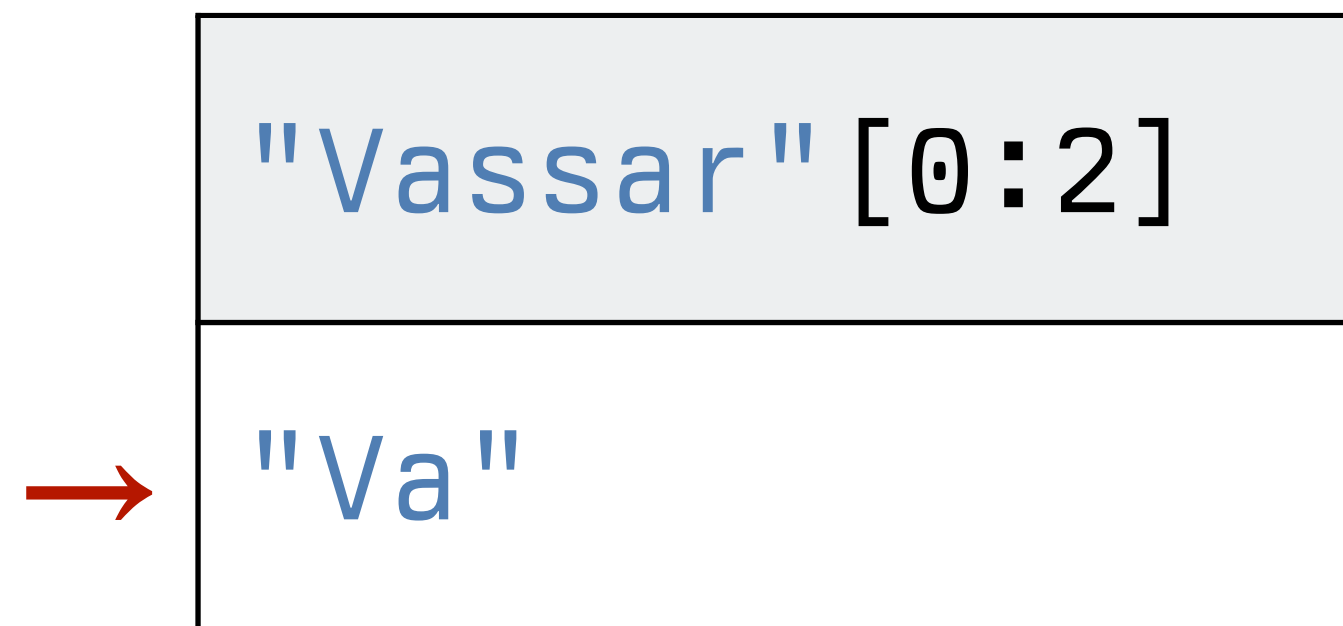
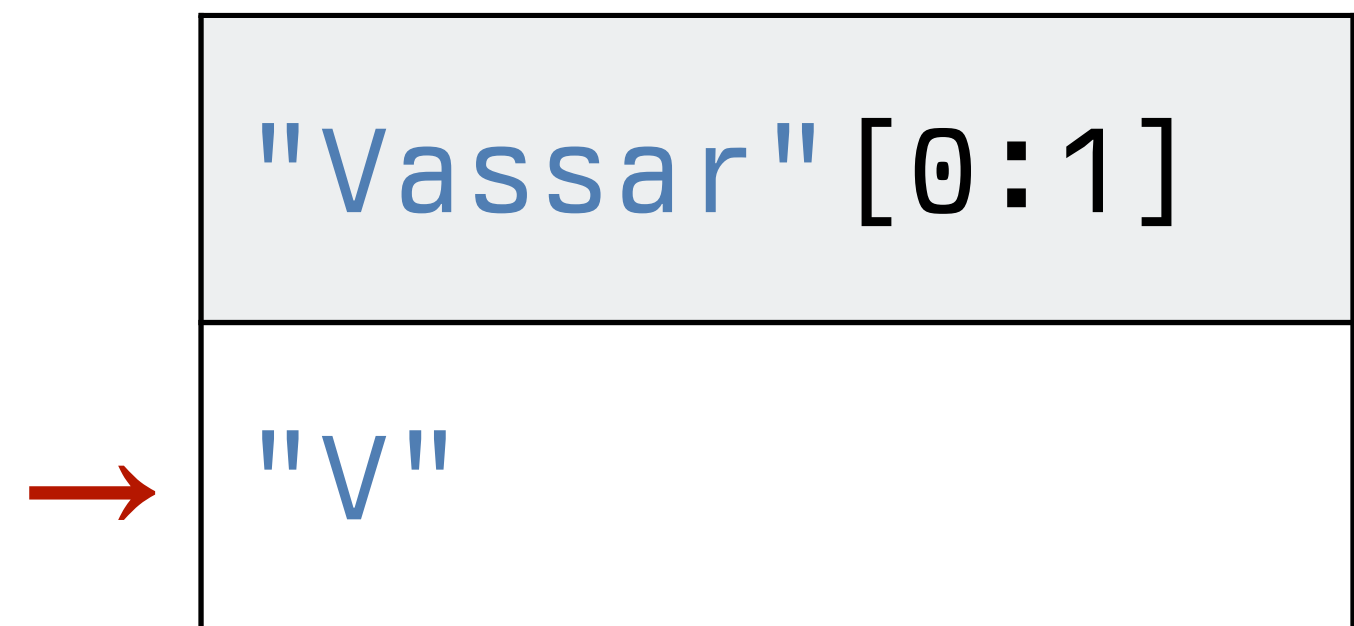
0 1 2 3 4 5  
" v a s s a r "



0 1 2 3 4 5  
" v a s s a r "



0 1 2 3 4 5  
" v a s s a r "



Let's switch to Colab to practice working with string indices.



While a string is a sequence of characters, a *list* is a sequence of any data we want!

To make a list, just enclose the contents in square brackets, e.g.,

```
["Frodo", "Sam", "Merry", "Pippin"]
```

```
[1, 2, 3]
```

Lists can hold a mix of different data types, e.g.,

```
[1.0, -3, True, "Spring"]
```

and even other lists:

```
[[1, 2, 3], [4, 5, 6]]
```

Just like with strings, we can access individual elements or subsequences, we can ask `in` questions, and we can concatenate lists using `+`.

We've seen that we can call a function on data, e.g.,

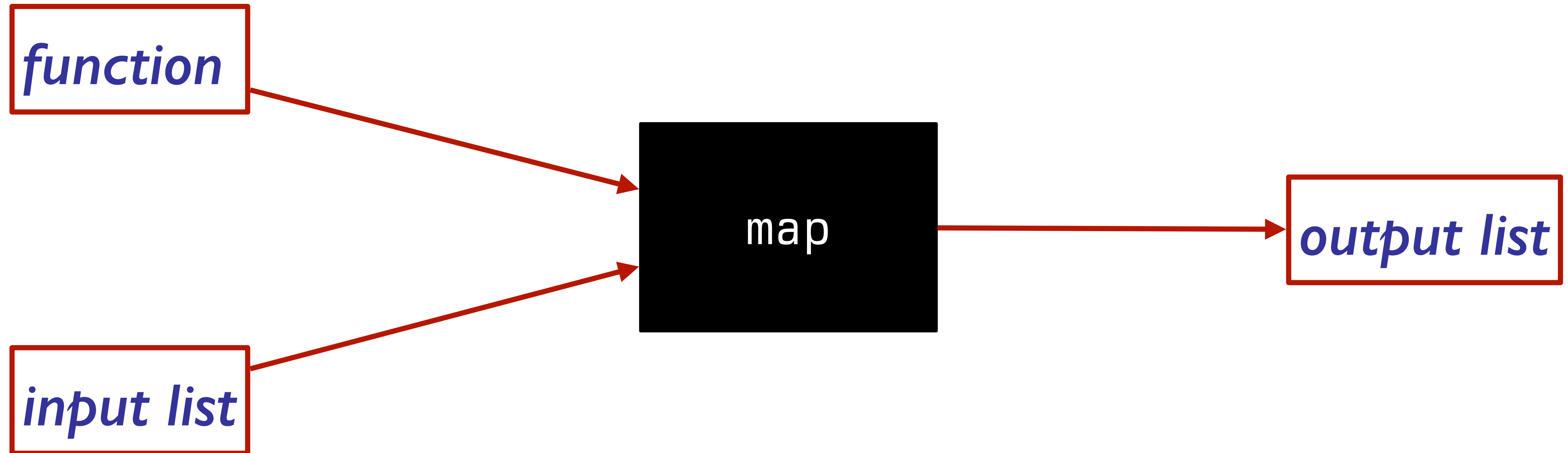
```
max(10, 12)
```

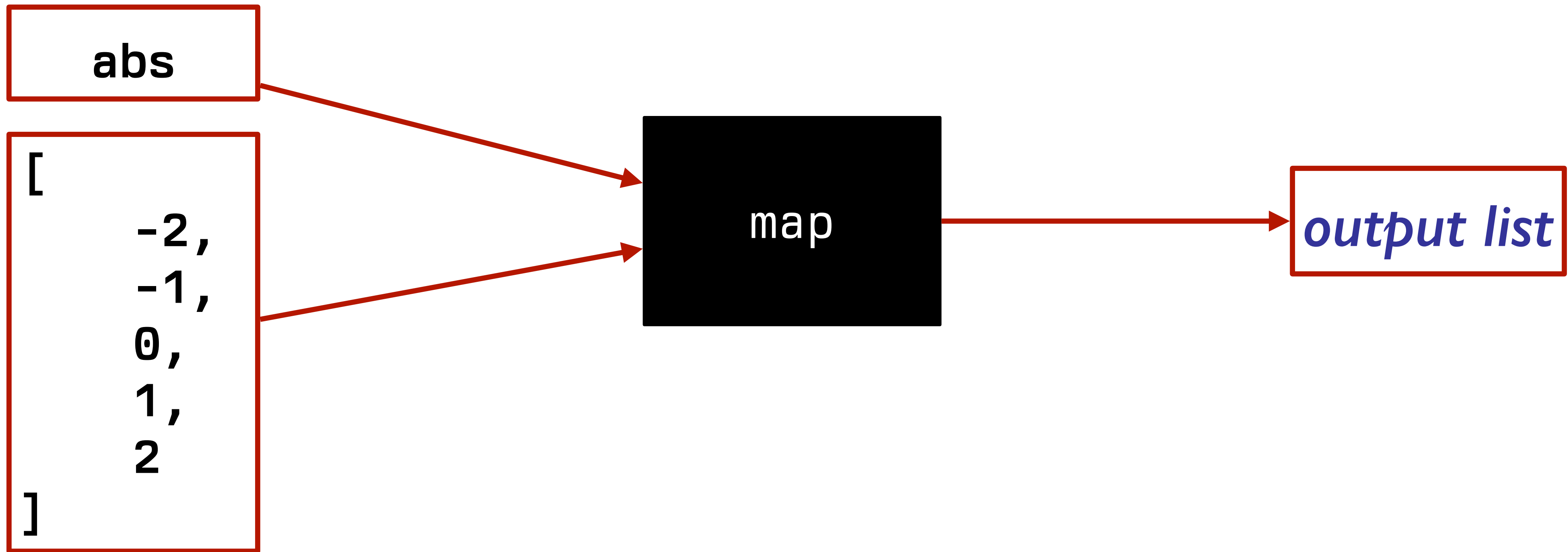
but we can also pass one function as an input to another function to tell it what to do.

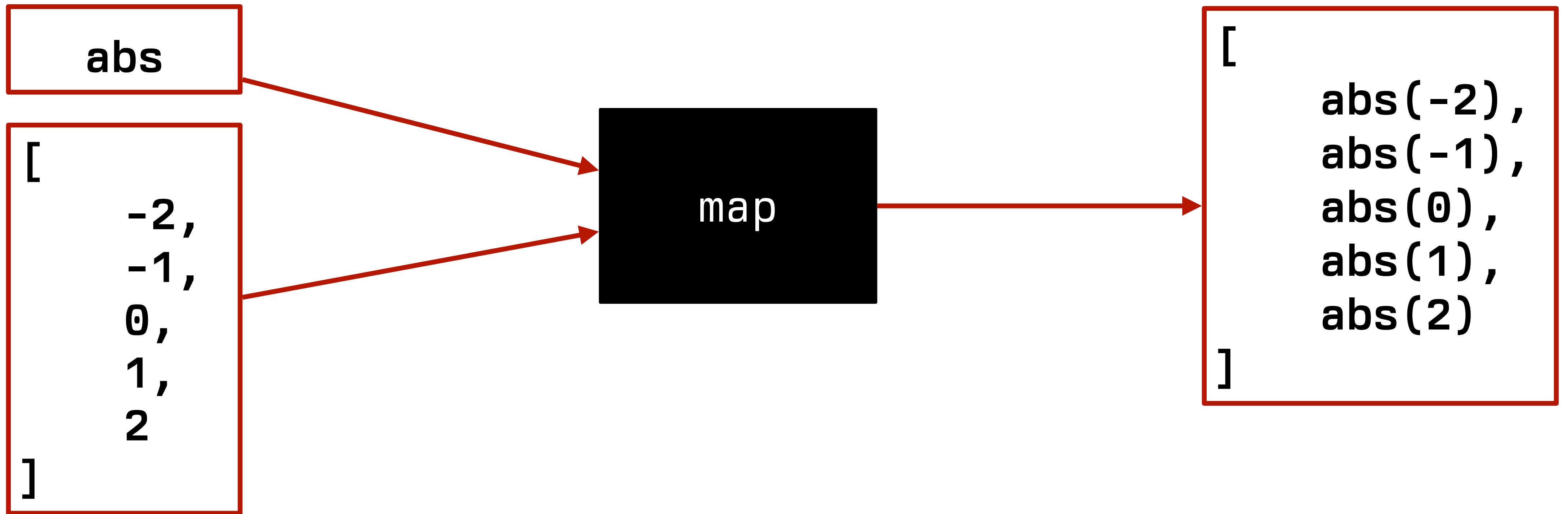
These are called *higher-order functions*, and they'll let us work with the individual elements of a list.

`map(function, input list)`

*When you call map, it returns a new list, where each item in **input list** has been transformed by running **function** on it.*





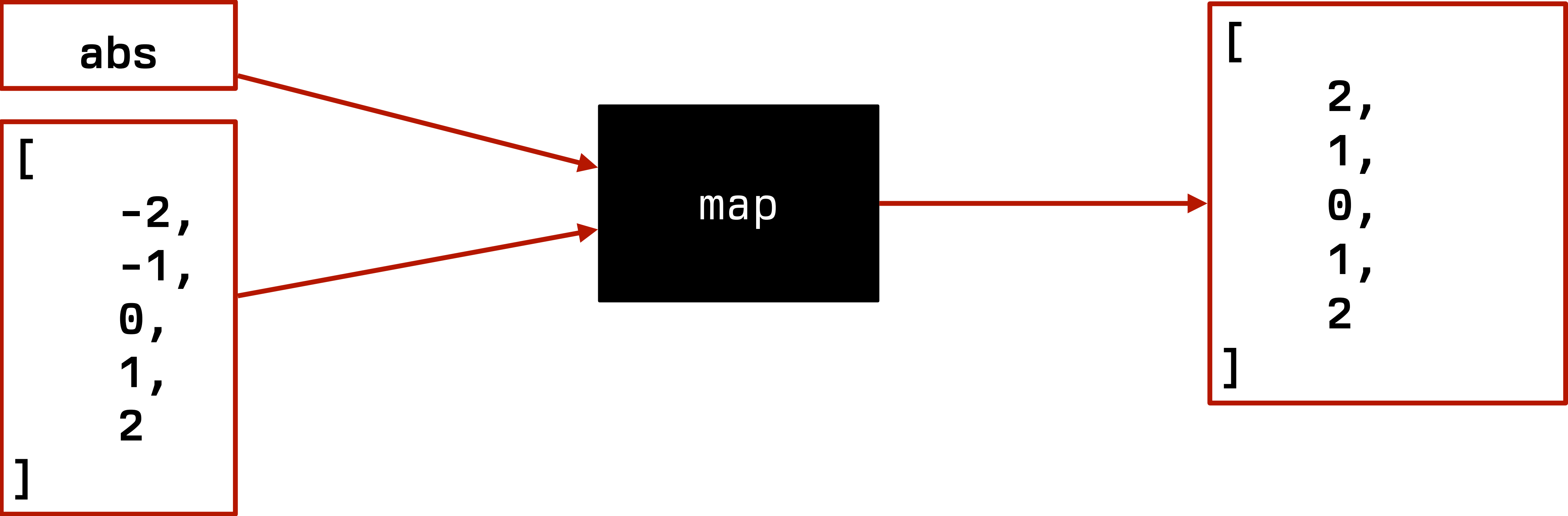


**abs**

**[  
-2,  
-1,  
0,  
1,  
2  
]**

**map**

**[  
2,  
1,  
0,  
1,  
2  
]**





An *array* is like a list, but designed for efficient computations, especially when they contain numbers.

Specifically, we'll be using the arrays from the popular NumPy library, which we load using an import statement:

```
import numpy as np
```



We can make an array out of a list by calling the `np.array` function on it:

```
np.array([1, 2, 3])
```

→ 

```
array([1, 2, 3])
```

Values in an array must all be of the same data type, and Python will attempt to convert (cast) them as appropriate:

```
np.array([5, -1, 0.3, 5])
```

→ 

```
array([5.0, -1.0, 0.3, 5.0])
```

```
np.array([4, -4.5, "not a number"])
```

→ 

```
array(["4", "-4.5", "not a number"])
```

For strings and lists, + joined two sequences together, one after another.

For arrays, it's element-wise addition:

```
np.array([1, 2, 3]) +  
np.array([1, 2, 3])
```

→ `array([2, 4, 6])`

```
np.array([-2, 1, 0]) +  
np.array([ 2, -1, 0])
```

→ `array([0, 0, 0])`

We can also easily scale the elements of an array by multiplying them by a single number:

	<code>np.array([1, 2, 3]) * 2</code>
→	<code>array([2, 4, 6])</code>

Or add a single number to each element:

	<code>np.array([1, 2, 3]) + 2</code>
→	<code>array([3, 4, 5])</code>

NumPy provides convenient built-in functions, e.g.,

```
np.mean(np.array([1, 2, 3]))
```

→ 2.0

Example



We can measure how much the radius of a tree grows in a given year by measuring the width of tree ring for that year

Suppose we have the ring widths (in mm) for a tree for five years:

```
ring_widths = np.array([3, 2, 1, 1, 3])
```

What was the total growth?

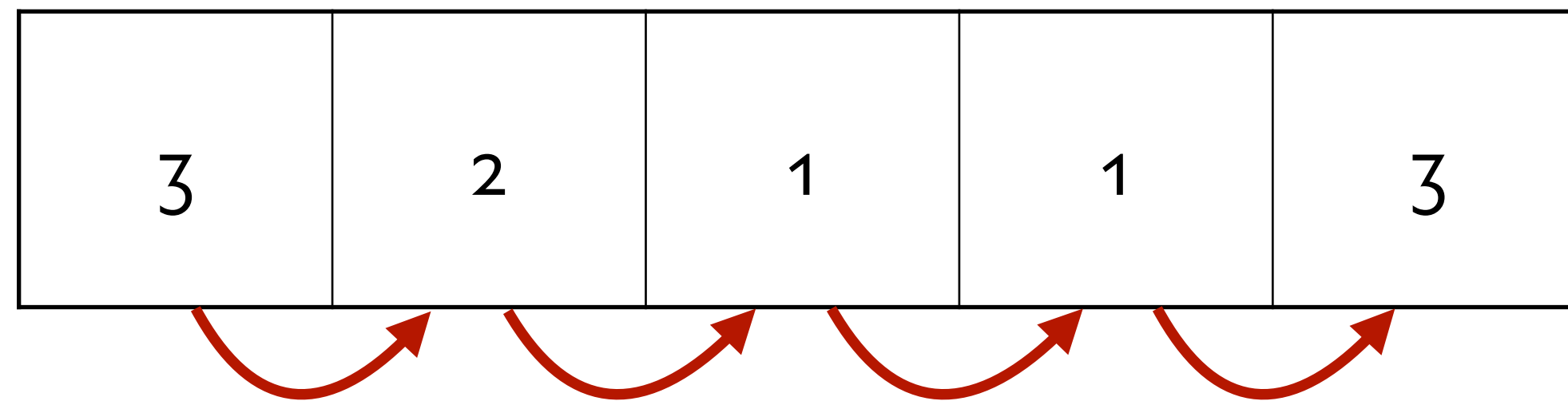
```
np.sum(ring_widths)
```

What was the average growth?

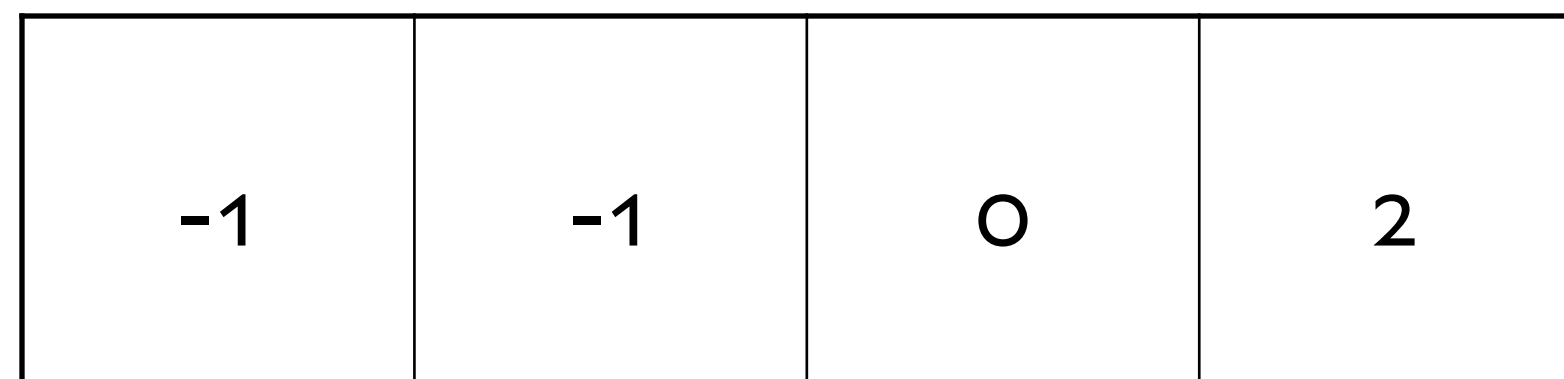
```
np.mean(ring_widths)
```

And `np.diff` produces an array of the differences between adjacent elements in the input, letting us see how the ring widths changed from year to year

```
ring_widths = np.array([3, 2, 1, 1, 3])
```



```
np.diff(ring_widths)
```



Next week, we'll see how we can build on arrays to work with tables of data!

