



Broadcasting and `np.newaxis` in NumPy — A Visual, Head-First Guide

Why do we need `np.newaxis`?

Imagine you have:

- An array of data (say IQ samples) shaped `(2, 3)` — 2 rows, 3 columns.
- A **gain** array shaped `(2,)` — just 2 elements, one gain value per row.

You want to multiply each row of IQ data by its gain. But shapes don't match!

The Problem: Shape Mismatch

Array	Shape	What it means
IQ data	<code>(2, 3)</code>	2 rows, 3 columns
Gain vector	<code>(2,)</code>	1 dimension, 2 elements

Question: How do we multiply row-wise when the gain vector has no "columns"?

Enter `np.newaxis`

Using `np.newaxis` adds a **new dimension**:

```
gain[:, np.newaxis]
```

Changes shape from `(2,)` to `(2, 1)` — 2 rows, 1 column.

Visualizing the Shapes

Gain vector original: shape (2,)	After <code>np.newaxis</code> : shape (2,1)
<code>[1, 0.5]</code>	<code>[[1], [0.5]]</code>
(just two numbers in a flat list)	(two rows, one column)

How does multiplication work now?

You want to multiply:

- IQ data: `(2, 3)`
- Gain vector with new axis: `(2, 1)`

Can they multiply elementwise?

NumPy Broadcasting Rules Simplified:

- Compare shapes **from the right** (end of shape tuple).
 - Dimensions must either be equal or one must be 1.
 - If one dimension is 1, it **stretches** to match the other.
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Compare shapes side by side:

Dimension (from right)	IQ data size	Gain vector size	Action
-1 (last dimension)	3	1	Gain is stretched to 3
-2	2	2	Matches exactly

What does “stretching” mean?

Logical stretching, not actual copying!

Gain vector	(2,1)	Broadcasted to	(2,3) logically:
[[1], [0.5]]		[[1, 1, 1], [0.5, 0.5, 0.5]]	

Now you can multiply elementwise:

IQ data	Gain broadcasted	Result
[[a, b, c], [d, e, f]]	[[1, 1, 1], [0.5, 0.5, 0.5]]	[[a*1, b*1, c*1], [d*0.5, e*0.5, f*0.5]]

Why is this useful?

Without adding the new axis:

- NumPy can't multiply (2,3) and (2,) directly.
- You get an error or wrong results.

Adding np.newaxis lets you:

- Keep your gain as a vector of length 2.
- Multiply it row-wise with your 2D IQ data easily.

Summary Table

Concept	What it means	Example
gain.shape = (2,)	1D array with 2 elements	[1, 0.5]
gain[:, np.newaxis]	Add a new dimension,	[[1], [0.5]]

Concept	What it means	Example
	shape becomes (2,1)	
Broadcasting	Dimensions with 1 stretch to match other	<code>[[1,1,1], [0.5,0.5,0.5]]</code>
Final multiplication	Multiply each IQ row by corresponding gain	<code>iq * gain[:, np.newaxis]</code>

Visual Diagram

Before:

iq shape: (2,3) gain shape: (2,)

`[[a, b, c],` `[g0, g1]`
`[d, e, f]]`

After `gain[:, np.newaxis]`:

gain shape: (2,1)

`[[g0],`
`[g1]]`

Broadcasted for multiplication:

`[[g0, g0, g0],`
`[g1, g1, g1]]`

Multiply elementwise:

`[[a*g0, b*g0, c*g0],`
`[d*g1, e*g1, f*g1]]`

Quick code snippet you can run:

```
import numpy as np

iq = np.array([[1, 2, 3], [4, 5, 6]])
gain = np.array([1, 0.5])

print("iq shape:", iq.shape)
print("gain shape:", gain.shape)

gain_expanded = gain[:, np.newaxis]
print("gain expanded shape:", gain_expanded.shape)

result = iq * gain_expanded
print("result:\n", result)
```
