

## # Pooling

Pooling operates on a features map matrix and reduces its spatial size.

### Max Pooling

For each pooling window  $R$  (e.g.  $2 \times 2$ ):

$$y = \max_{(i,j) \in R} x_{i,j}$$

This means:

- take the maximum value inside the region.
- Keep the strongest activation

### Average Pooling

For each pooling window  $R$ :

$$y = \frac{1}{|R|} \sum_{(i,j) \in R} x_{i,j}$$

This means:

- sum all values in the region
- divide by number of elements

### Output Size

If:

- input size =  $N$
- pool size =  $F$
- stride =  $S$

Then:

$$\text{Output size} = \frac{N - F}{S} + 1$$

### Key Idea

Pooling reduces feature map size while preserving important information, making CNNs faster and more robust.