

## The Effects of Attention Mechanism

### Before Attention

All inputs treated equally

	$x_1$	$x_2$	$x_3$	$x_4$
$y_1$	0.25	0.25	0.25	0.25
$y_2$	0.25	0.25	0.25	0.25
$y_3$	0.25	0.25	0.25	0.25

**Static context:** Same fixed representation used for all outputs. No selective focus on relevant information.

### With Attention

Dynamic attention weights

	$x_1$	$x_2$	$x_3$	$x_4$
$y_1$	0.7	0.2	0.05	0.05
$y_2$	0.1	0.6	0.25	0.05
$y_3$	0.05	0.15	0.3	0.5

**Dynamic context:** Different attention distribution for each output. Focuses on most relevant inputs at each step.



#### Better Alignment

Learns input-output correspondence automatically. Essential for translation tasks.



#### Long Sequences

Handles long inputs effectively without information bottleneck. No gradient vanishing.



#### Interpretability

Attention weights visualize what model focuses on. Debugging and understanding easier.

Translation Quality

↑ 15-20%

BLEU Score Improvement

Long Sequences

↑ 30-40%

Performance on 50+ tokens

Gradient Flow

✓ Stable

Direct path to each input