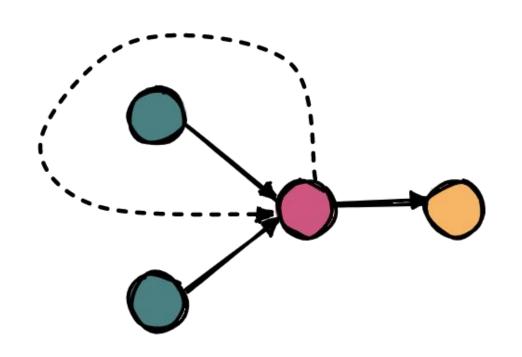
Recurrent neural networks

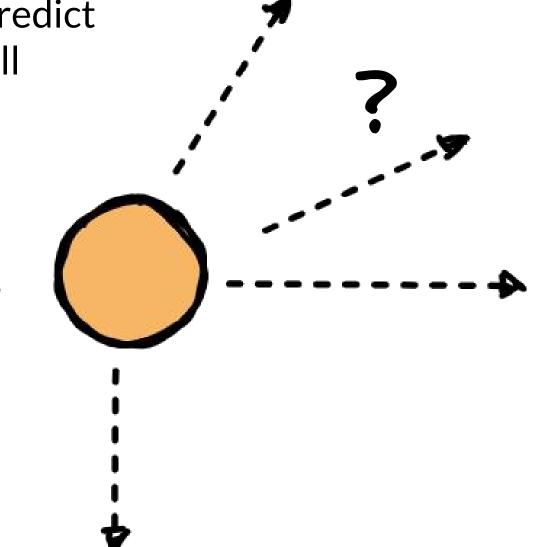


Week 19

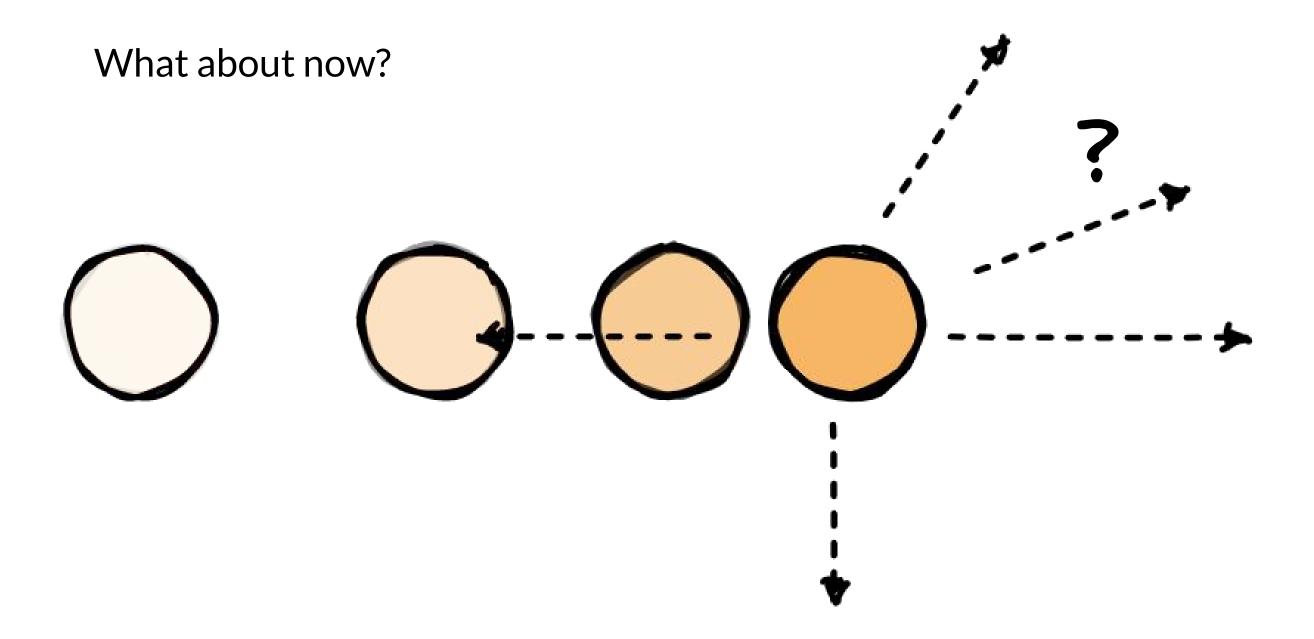
Middlesex University Dubai; CST4050 Fall21; Instructor: Dr. Ivan Reznikov

Ball position prediction

Given the image of the ball, can you predict the direction and location where it will move next?

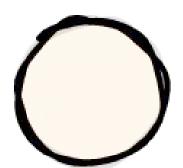


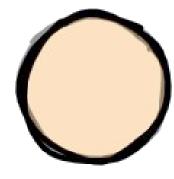
Ball position prediction

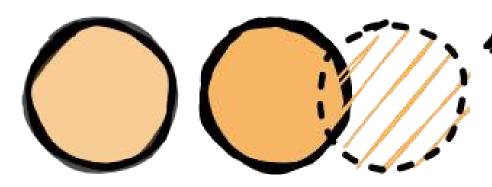


Ball position prediction

High chance that you correctly the exact location and direction. Previous ball locations gave you enough additional information to make an accurate forecast.







Try saying all the numbers in order from 0 to 11 as fast as you can.

What if we randomize the order of the numbers?

A bit slower, right?

The order matters.

What if we start from 4? Pretty much fast as the first time

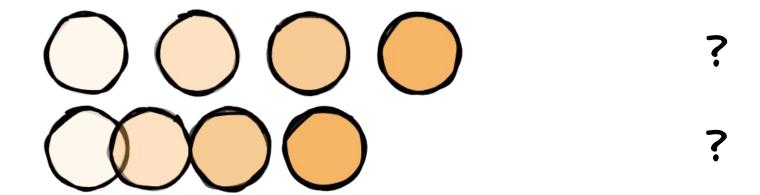
Now let's do the same exercise with the alphabet

ABCDEFGHIJKLMNOPQRSTUVWXYZ

Now start with the letter F. It takes a while to pick up the pace.

Why is that?

One of the reasons may be "more structured" sequenced:



The official website for Game of Thrones on HBO, ____ Formula1 is the highest class of international racing _____



Problem 0: How to push text to a neural network, if the length of the sequence may vary?

The fact that Batman is so reliant on tech is his ...

Solution a: Fixed small window

$$\begin{bmatrix} 00001 & 00010 \end{bmatrix} \rightarrow \text{prediction}$$
 is his

Problem a: Long-distance relationships

London is my home city. This is the reason I fluently speak English

The fact that Batman is so reliant on tech is his ...

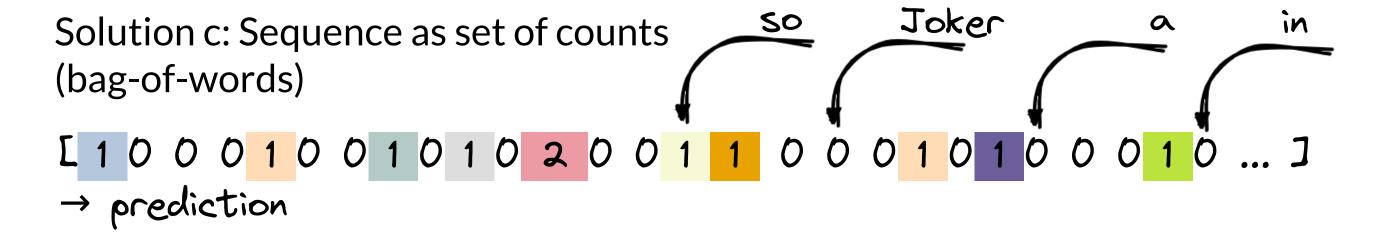
Solution b: Fixed wide window

```
L 00001 00010 01011 10110 11000 01110 11010 01110 10101 11000 11100] The fact that Batman is so reliant on tech is his \rightarrow prediction
```

Problem b: Lose of sense if met in different part of sequence 11010 01110 10101

reliant on tech

The fact that Batman is so reliant on tech is his ...



Problem c: Lose of sense if met in different part of sequence

Batman is reliant on tech == tech is reliant on Batman

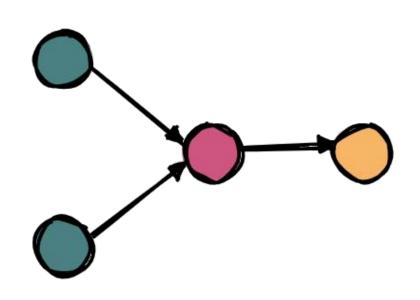
Model criteria

Requirements:

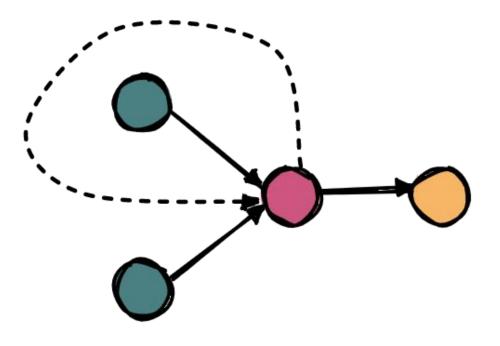
- 1. Handle sequences of different length
- 2. Track long-term dependencies
- 3. Share parameters across sequence
- 4. Save order information

Saving memories

Having memory may be pretty important when we deal with time-series events. A regular perceptron has two inputs from the input or previous layer. But what if we could pass the "memory" – the last value of the neuron? This is called a **recurrent** perceptron. The "memory" is called **hidden state**.

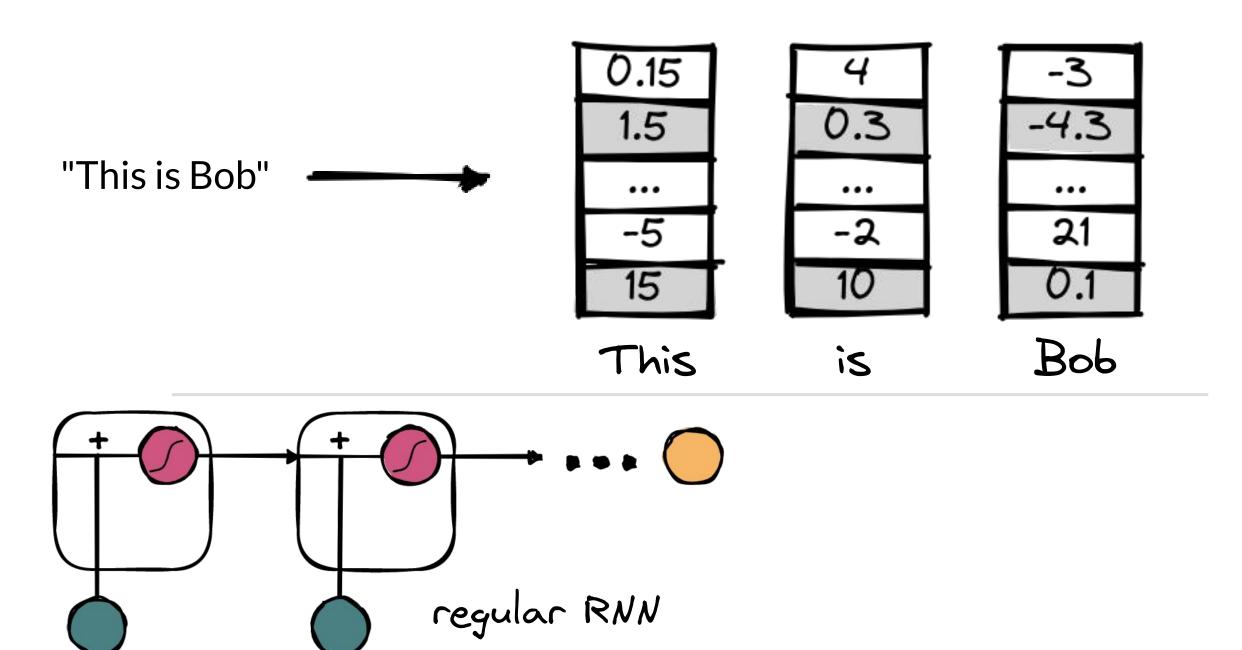


regular perceptron

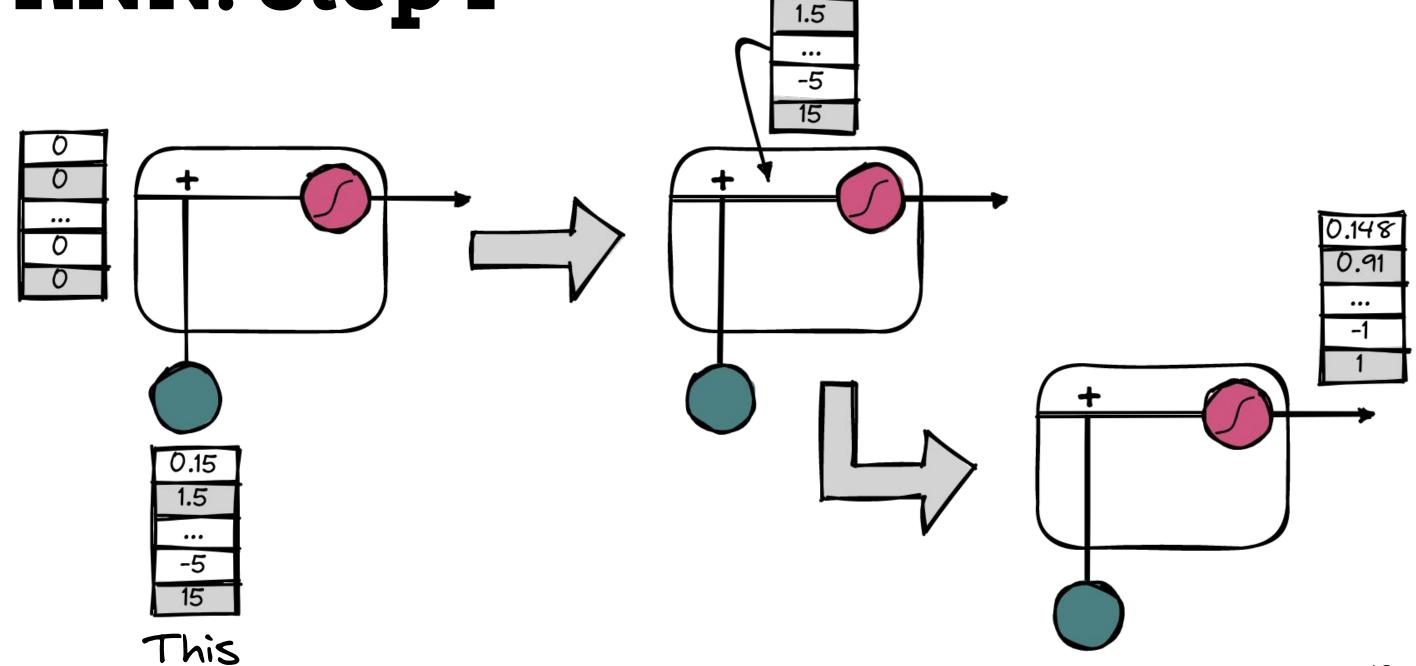


recurrent perceptron

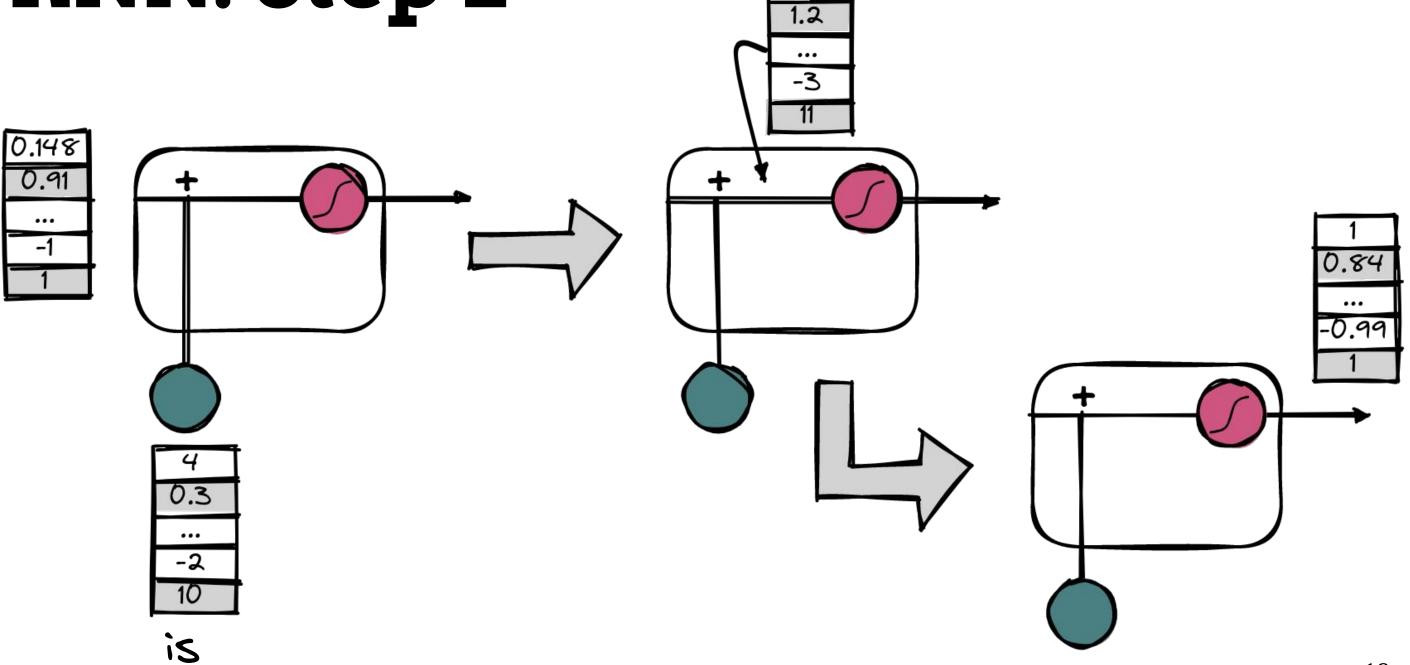
RNN: Case1



RNN: Step 1

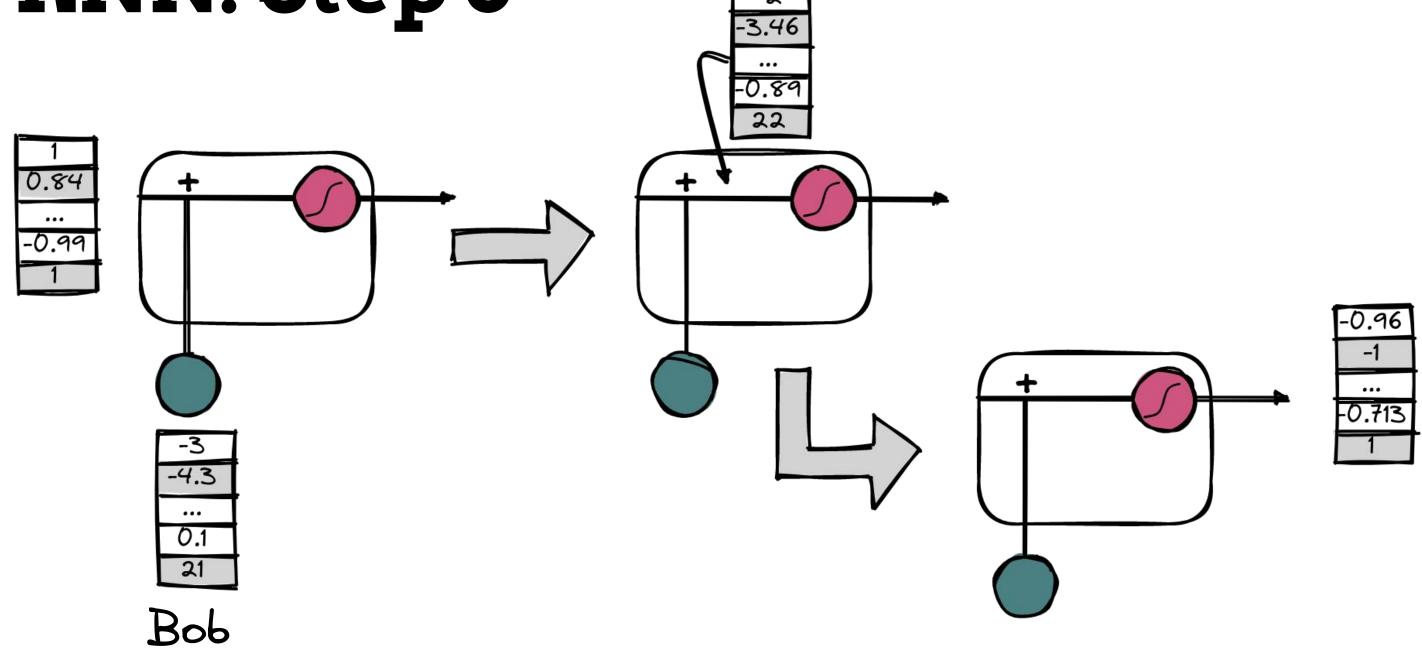


RNN: Step 2

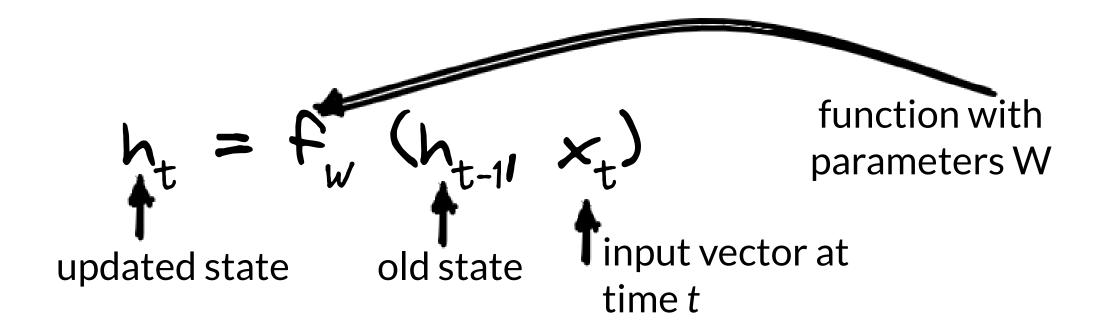


4.14

RNN: Step 3



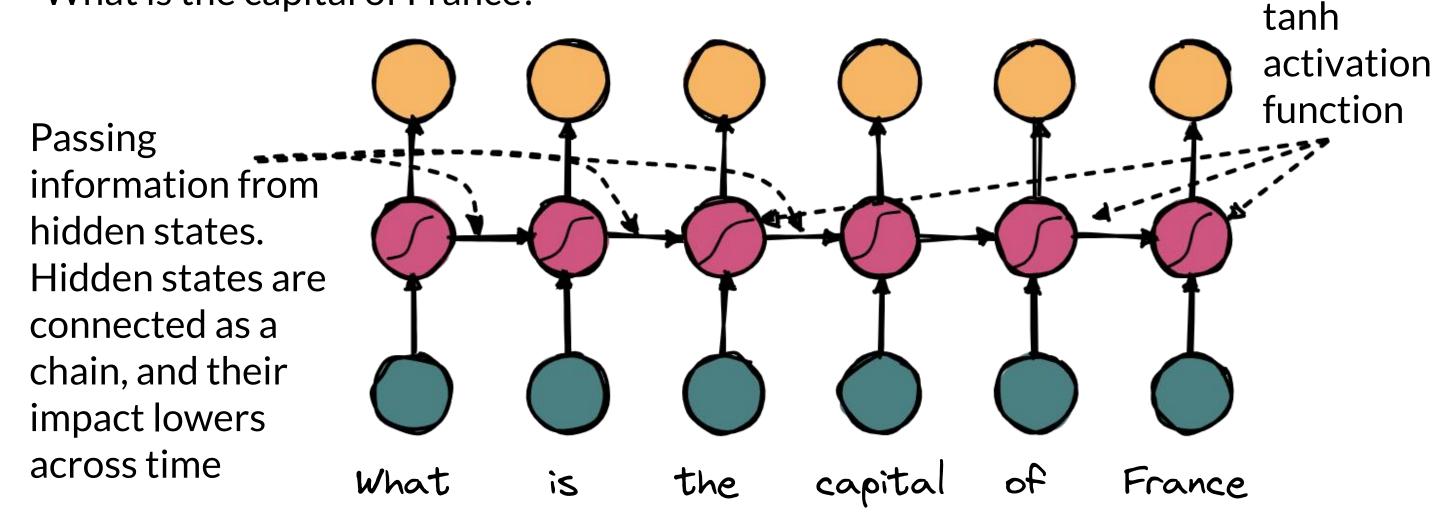
RNN: Formula



RNN: Case2

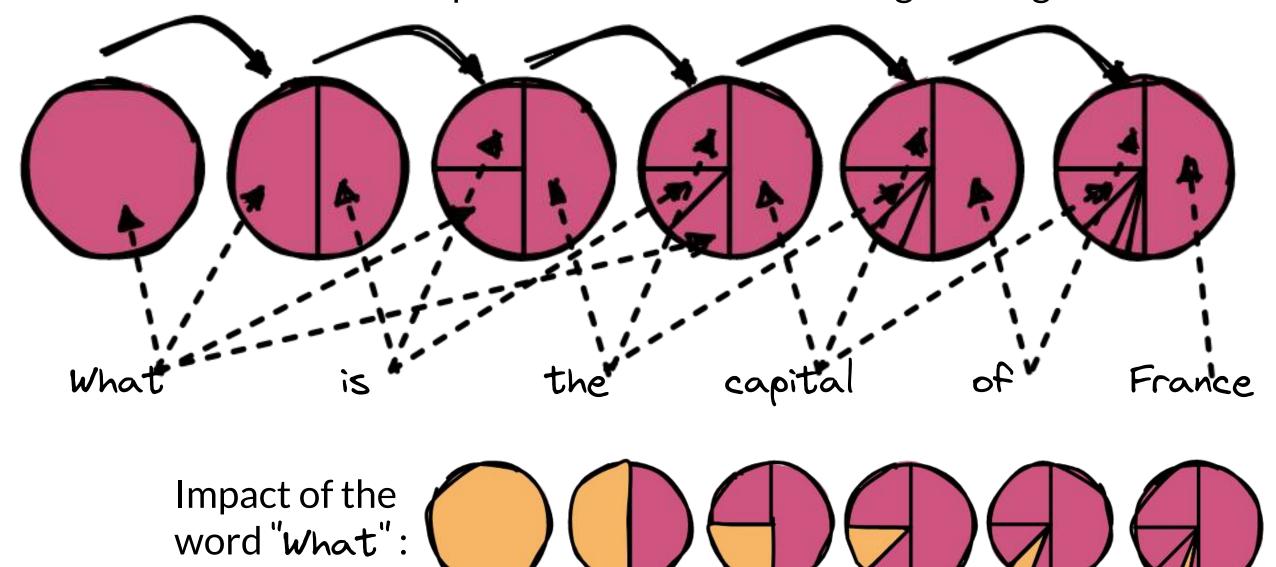
Let's take another query example:

"What is the capital of France?"



Vanishing gradient

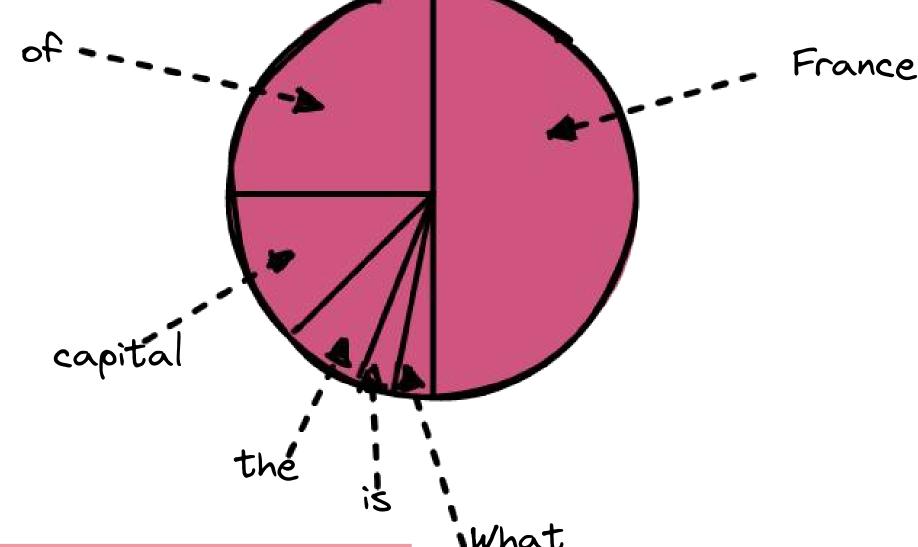
Let's take a look at how importance of the word changes along the chain:



Vanishing gradient

As one can notice, the impact of recently met words is much higher than the ones from the beginning of the sequence.

This is an example of the vanishing gradient problem.



What is the capital of France