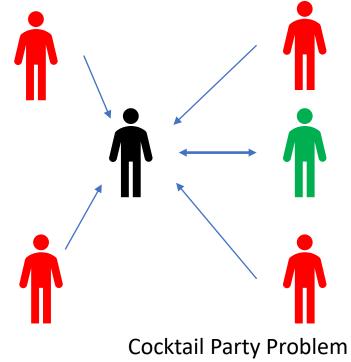
#### Problem

 need to focus on one aspect while ignoring other (distracting) aspects at the same time

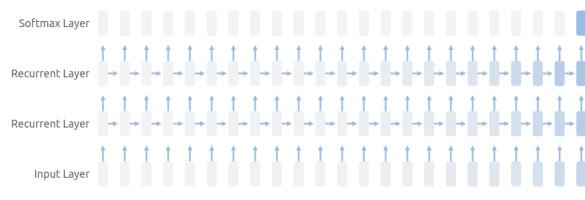
- only keep necessary information
- ignore information rather than increase information



#### Problem

- Vanishing gradient problem in RNNs
- caused by time dependency
- theoretically, every part of previous sequence know,
- BUT gradients become smaller
- farther away content has less impact
- RNNs work only well for short sequences

This is a very long sample text in which the start is forgotten.



**Vanishing Gradient:** where the contribution from the earlier steps becomes insignificant in the gradient for the vanilla RNN unit.

Source: <a href="https://distill.pub/2019/memorization-in-rnns/">https://distill.pub/2019/memorization-in-rnns/</a>

#### Intuition

- ability to focus on certain parts of input
- Jacobian matrix represents sensitivity of output to input
- partial derivatives (back propagation)

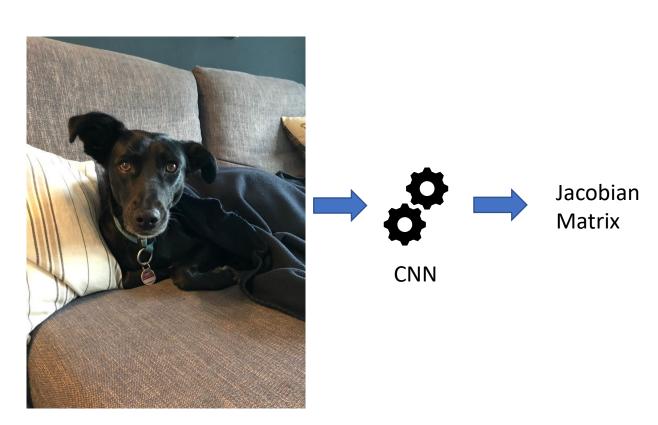
$$J = \begin{bmatrix} \frac{\delta y_1}{\delta x_1} & \dots & \frac{\delta y_1}{\delta x_k} \\ \dots & \dots & \dots \\ \frac{\delta y_m}{\delta x_1} & \dots & \frac{\delta y_k}{\delta x_n} \end{bmatrix}$$

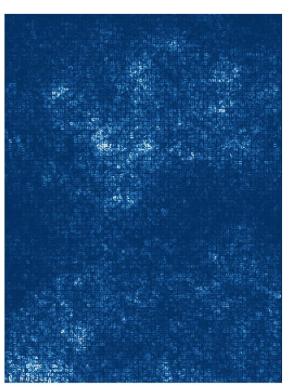


Class Activation Map

(Source: <a href="https://glassboxmedicine.com/2019/06/11/">https://glassboxmedicine.com/2019/06/11/</a> cnn-heat-maps-class-activation-mapping-cam/)

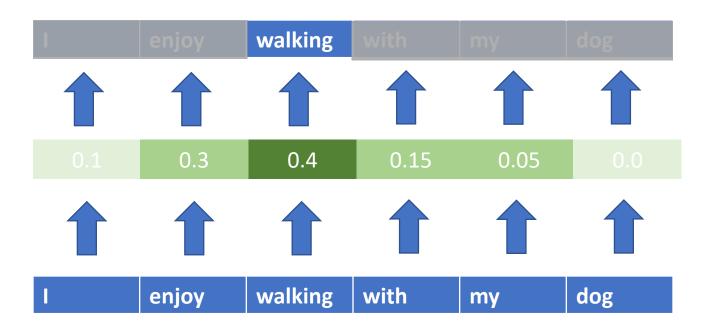
#### Intuition





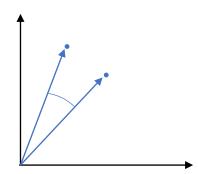
#### Introduction

Attention



### Types of Attention

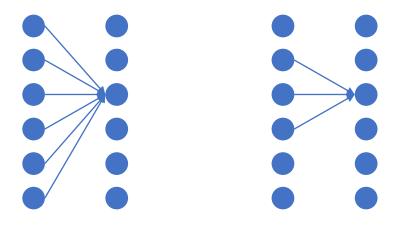
- Dot-product attention
  - calculates how close two vectors align in terms of point directions
- scaled dot-product attention
  - scales dot-product by square root of key dimension



- multi-head attention
  - splits query, key, and value vectors into multiple heads and applies dotproduct independently
- self-attention
  - input sequence applied as query and key

### Global vs. Local Attention

- global attention references to all input nodes
- local attention references to subset of input nodes

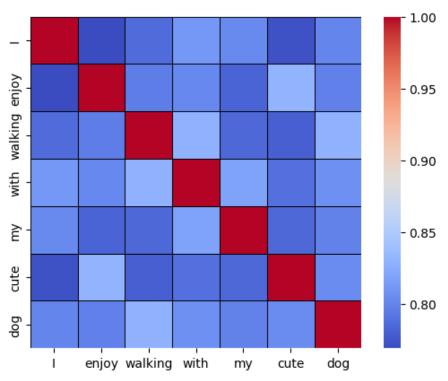


**Global Attention** 

**Local Attention** 

#### Self-Attention

score between elements of the same sequence



#### Advantages / Disadvantages



- work a bit like skip-connections
- solves vanishing gradient problem
- improved accuracy
- improved efficiency (reduced training time)
- improved explainability



- increased training difficulty
- large amount of data required
- prone to overfitting

#### **Applications**

- Natural Language Processing
- Computer Vision, e.g. Vision Transformers
- Speech recognition (cocktail party problem)
- music generation