



UE21CS343BB2

Topics in Deep Learning

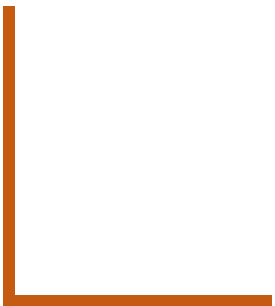
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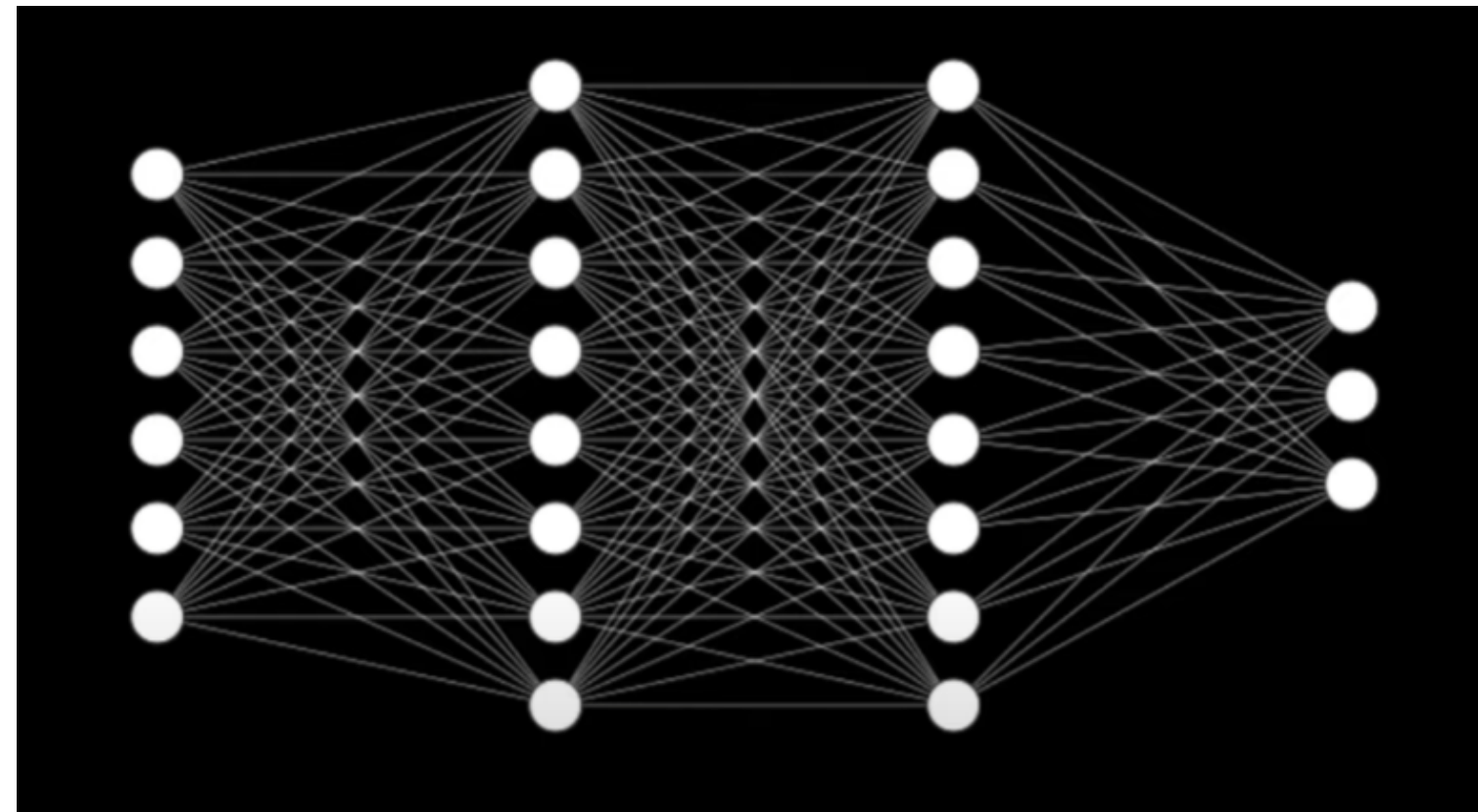
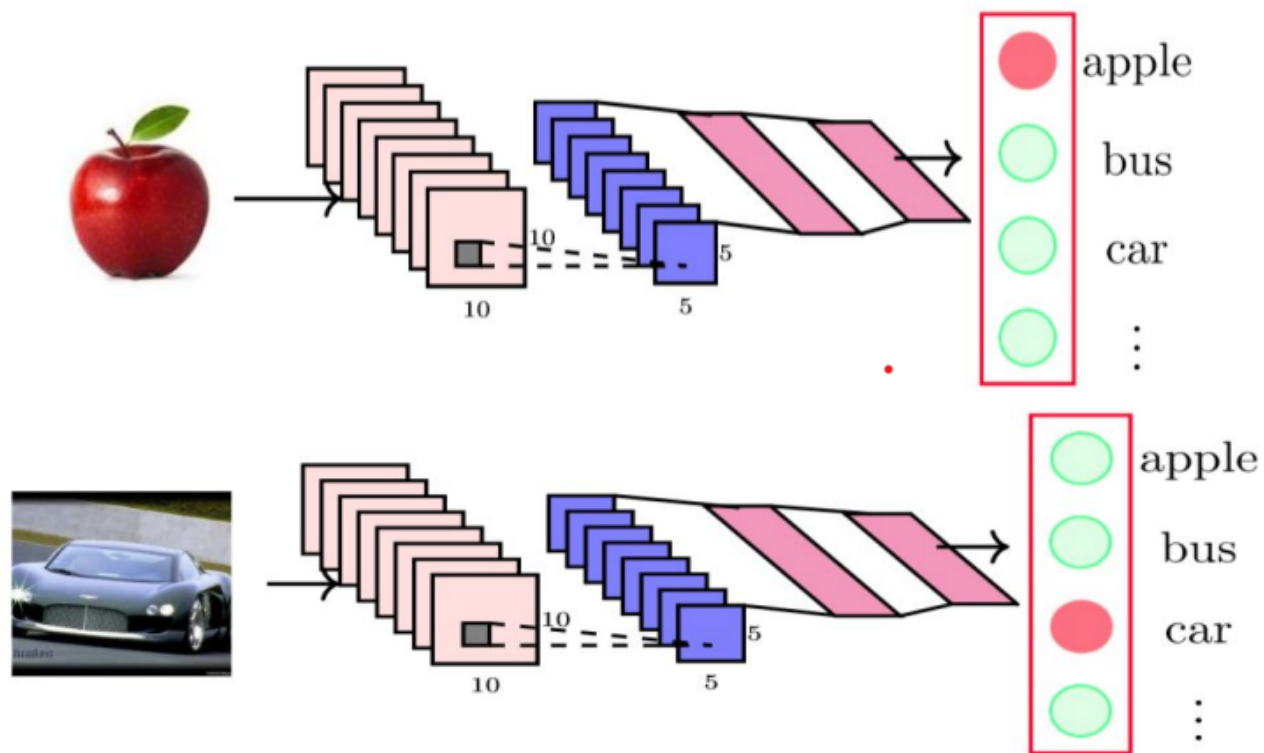
**Ack:Aryan Sharma,
Teaching Assistant**

UE21CS343BB2: Topics in Deep Learning

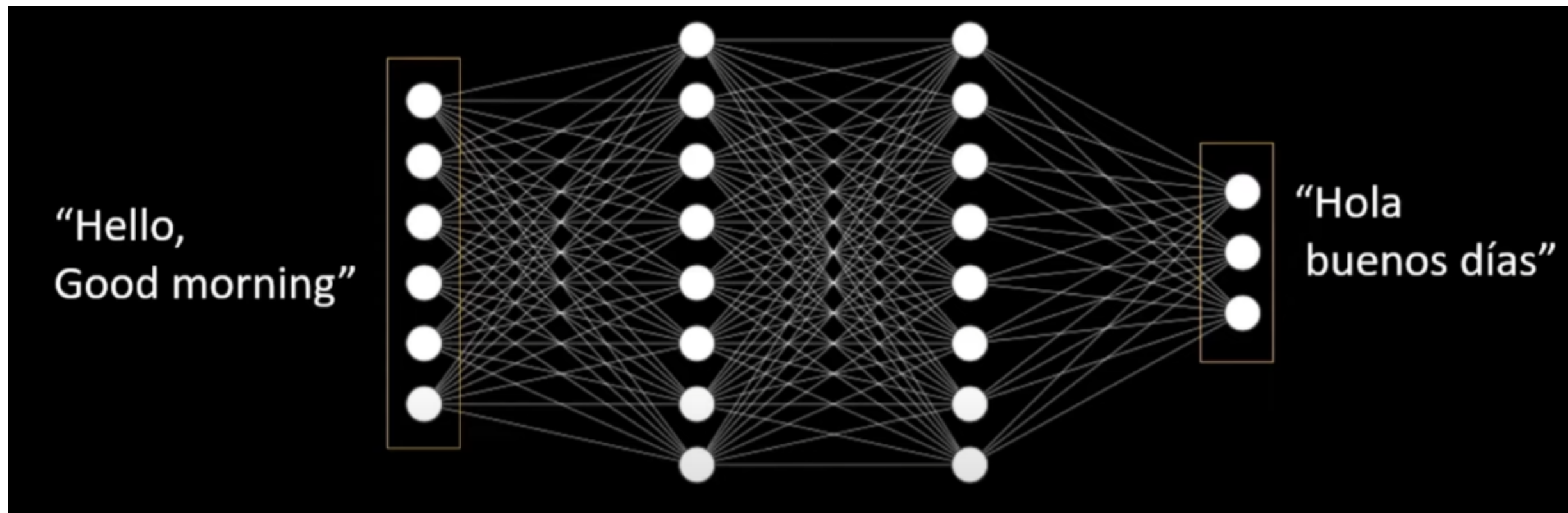
Introduction to RNNs



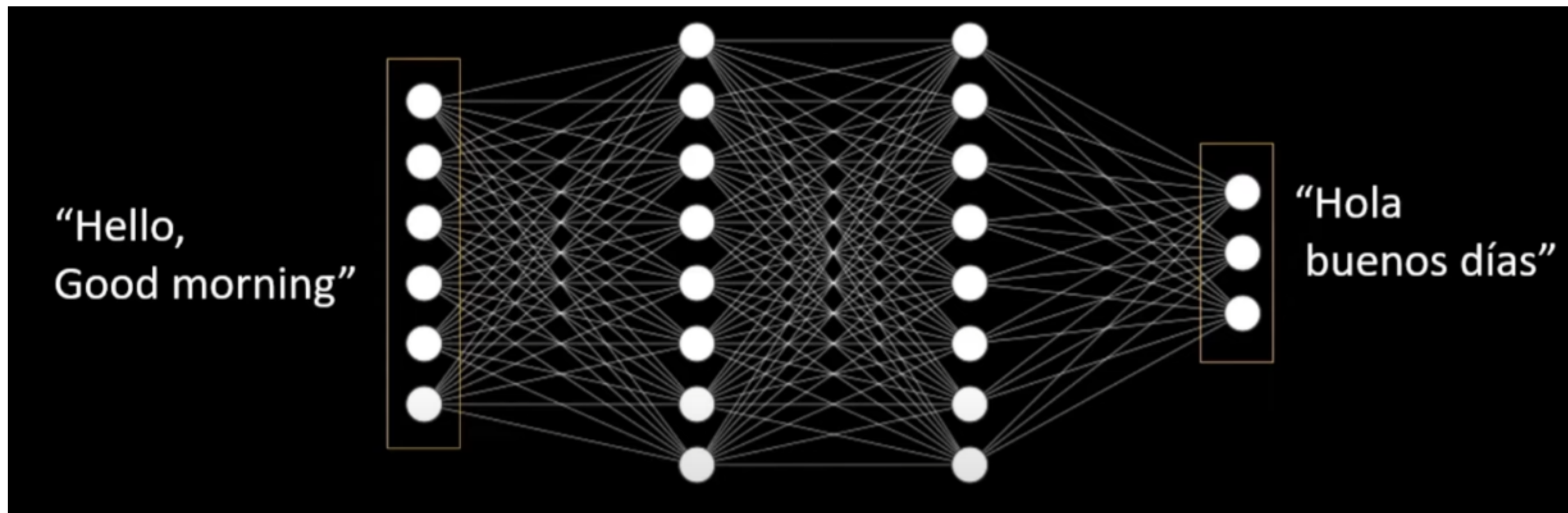
- Artificial Neural Networks and Convolutional Neural Networks work very well in predicting outputs to specific sized inputs.
- Here the inputs being certain images resized to fit a particular ratio and a fixed set of classes or outputs.



- To understand its limitations, let's take an example of a use case.
- Let's say we want to translate a sentence from English to another language.
- Consider the sentence - “Hello good morning” must be translated to spanish.

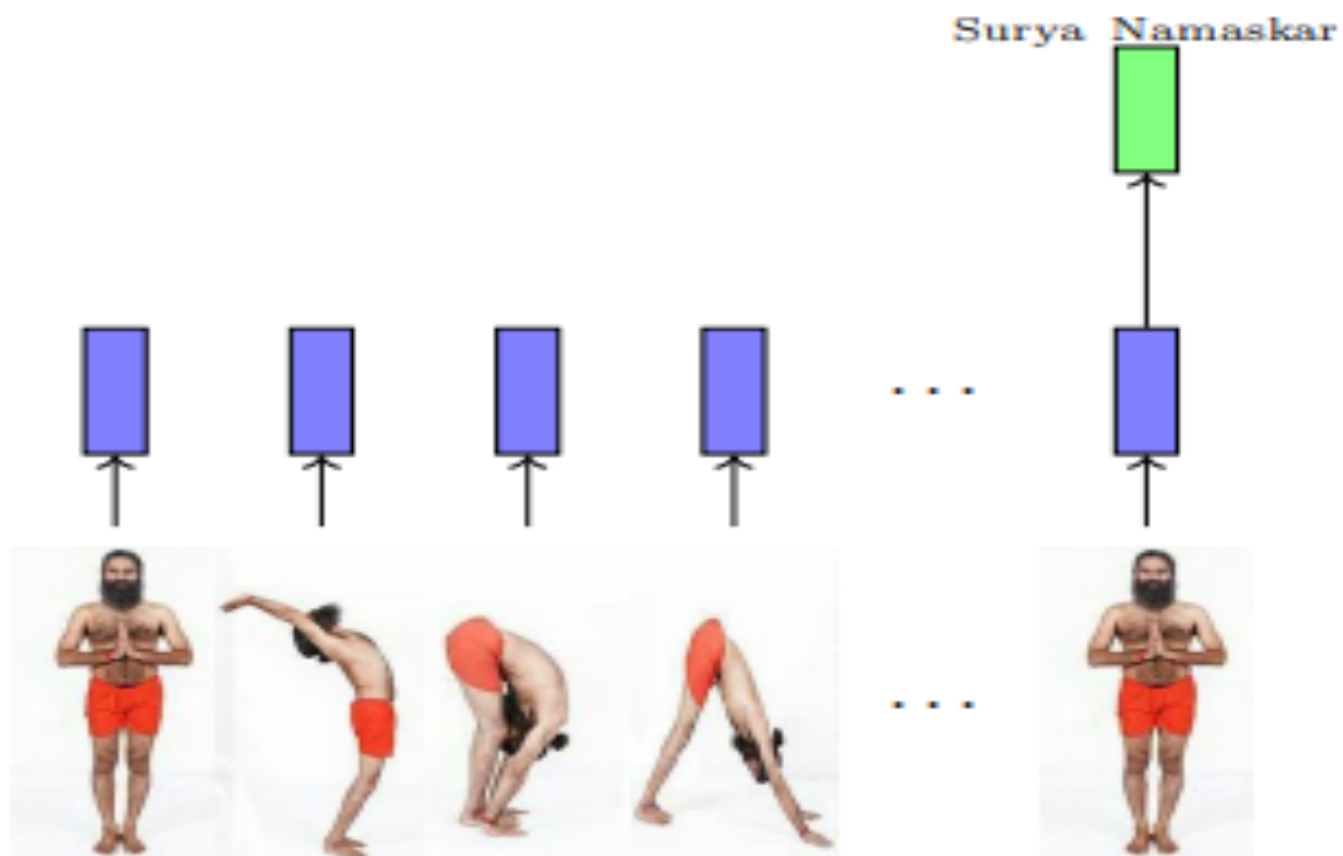


- The first step would be to convert the words into numbers, specifically vectors which is a format the computer can understand.
- For example - Let's use One Hot Encoding.
Hello = [1,0,0]; Good = [0,1,0]; Morning = [0,0,1].
- But in the case of there being a longer sentence we have taken an ANN with 6 input nodes. What would the rest of the nodes get as an input?

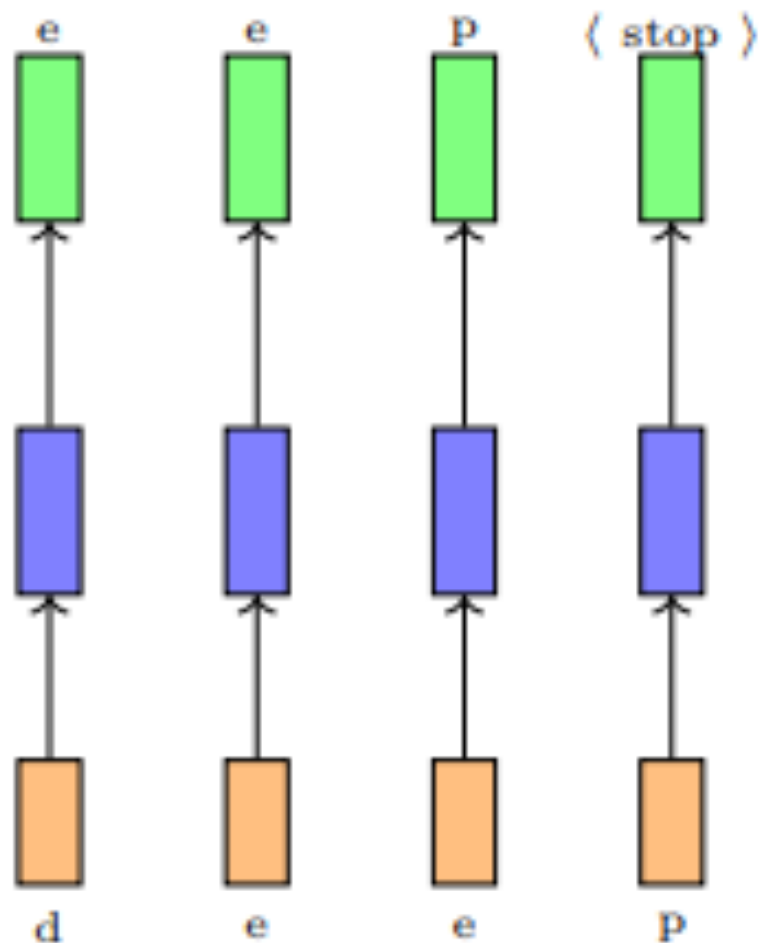


- We could zero encode it in case of a shorter input. This is okay considering our vocabulary is only 3 words.
- In real cases, if each root word in the language or at least the vocabulary in some specific case would need a vector, that would mean a vector whose dimensions are 10,000 or more!
- But, in cases of a large vocabulary (words and their respective vectors), the size of the vectors would lead to unnecessary overhead.
- **Problem #1- As you can see above, ANNs cannot accept variable length inputs.**

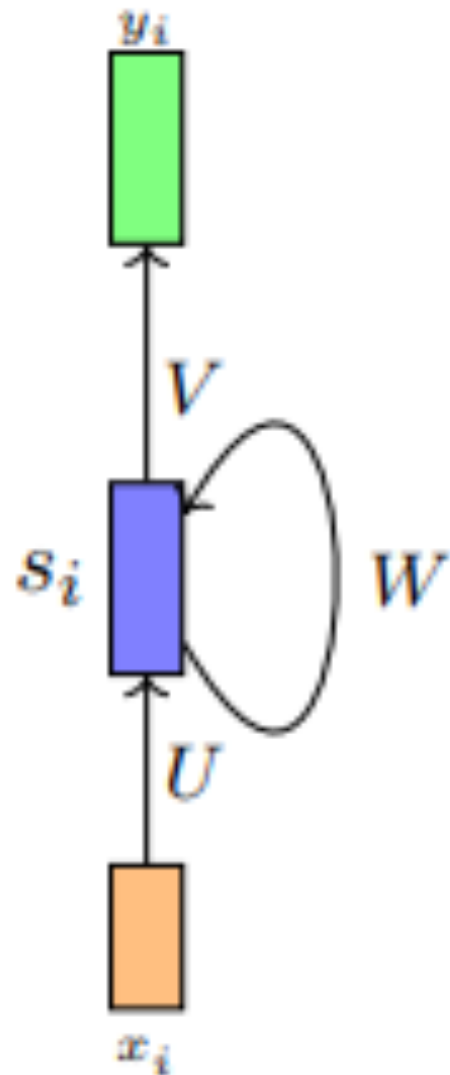
- Now let's take another example, video captioning.
- The video will be broken into individual images which are passed to the network.
- The order here becomes very important for the caption to make sense.



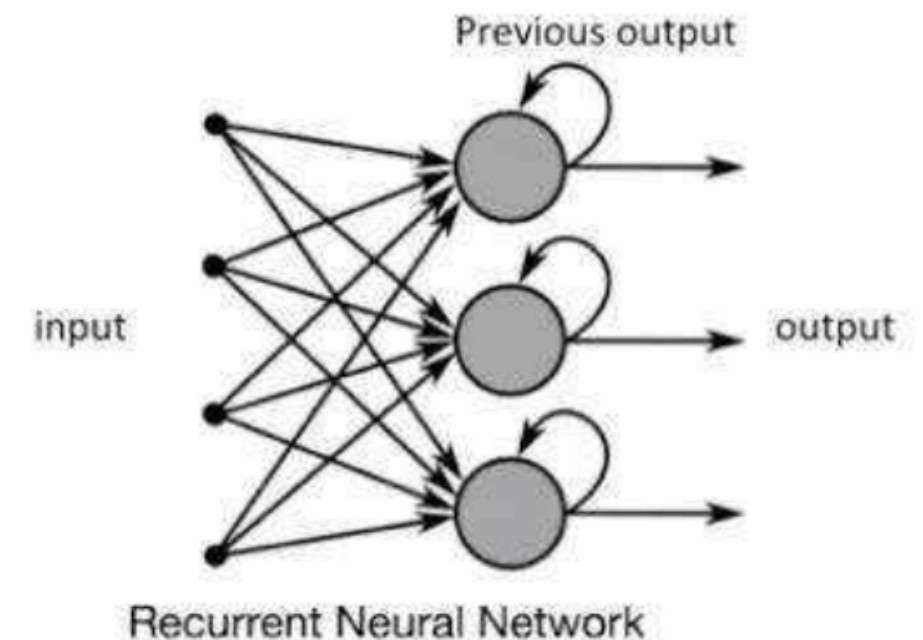
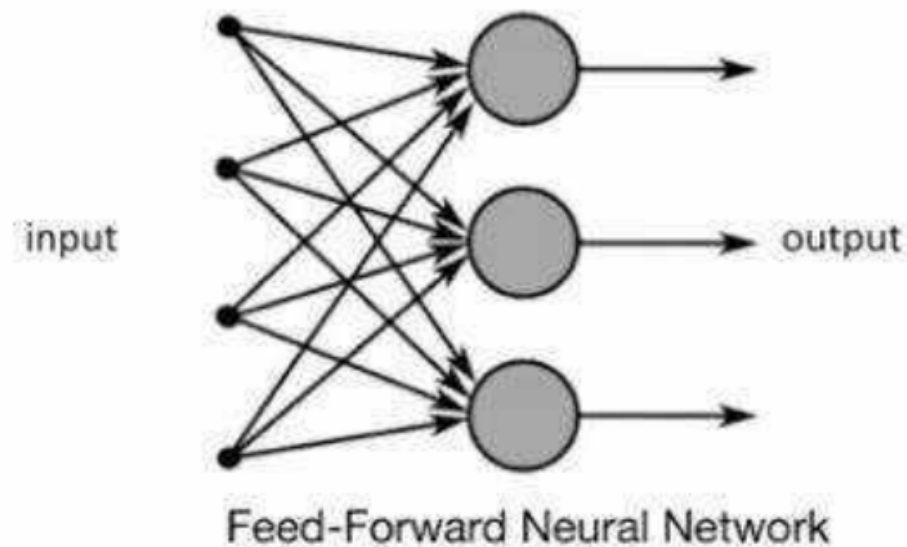
- Let's take more examples, consider the task of auto completion
- Given the first character 'd' you want to predict the next character 'e' and so on.

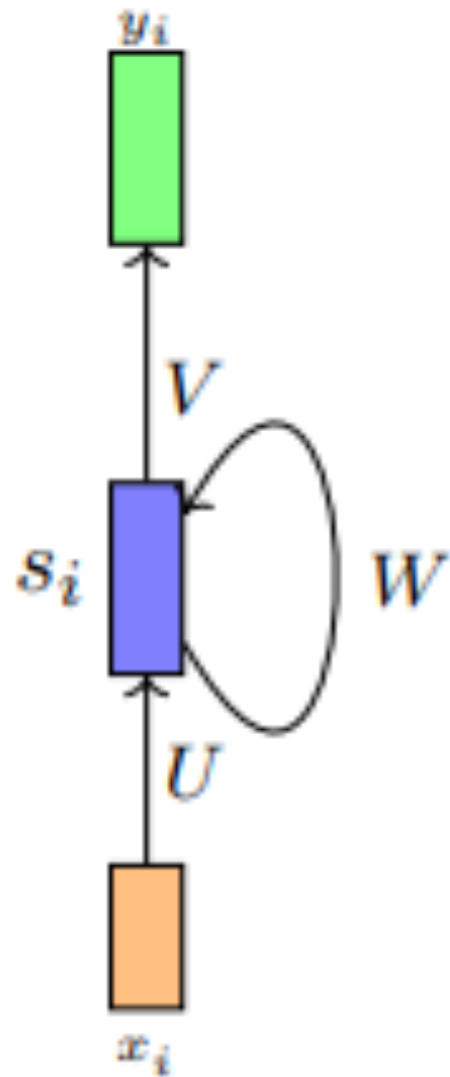


- As seen here, the sequence matters but also, the word can be anything other deep too! It can say detail, desperate, destroy or anything else after the 'd' and 'e'.
- The ANN would not be able to handle such cases.
- In ANNs, the order the input is passed into the network is arbitrary.
- **Problem #2- Thus, in ANNs the sequence is not taken into consideration.**



- In the given diagram, a chunk of neural network, A, looks at some input x_t and outputs a value y_t . A loop allows information to be passed from one step of the network to the next.

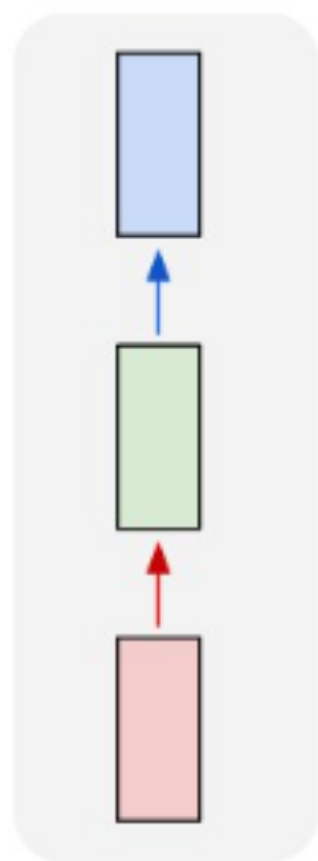




- This information is stored in the 'hidden state' which constitutes its memory and passed in the 'hidden layer'.
- A recurrent neural network can be thought of as multiple copies of the same network(A), each passing a message to a successor.

- Each rectangle is a vector and arrows represent functions (e.g. matrix multiply). Input vectors are in red, output vectors are in blue and green vectors hold the RNN's state (more on this soon).

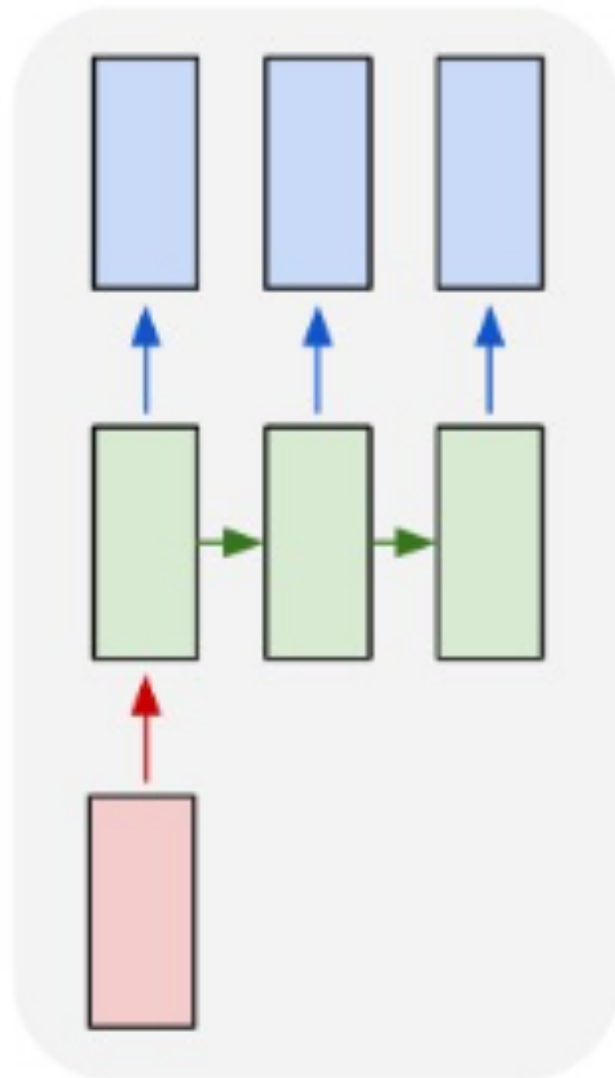
one to one



- Vanilla mode of processing without RNN, from fixed-sized input to fixed-sized output (e.g. image classification).

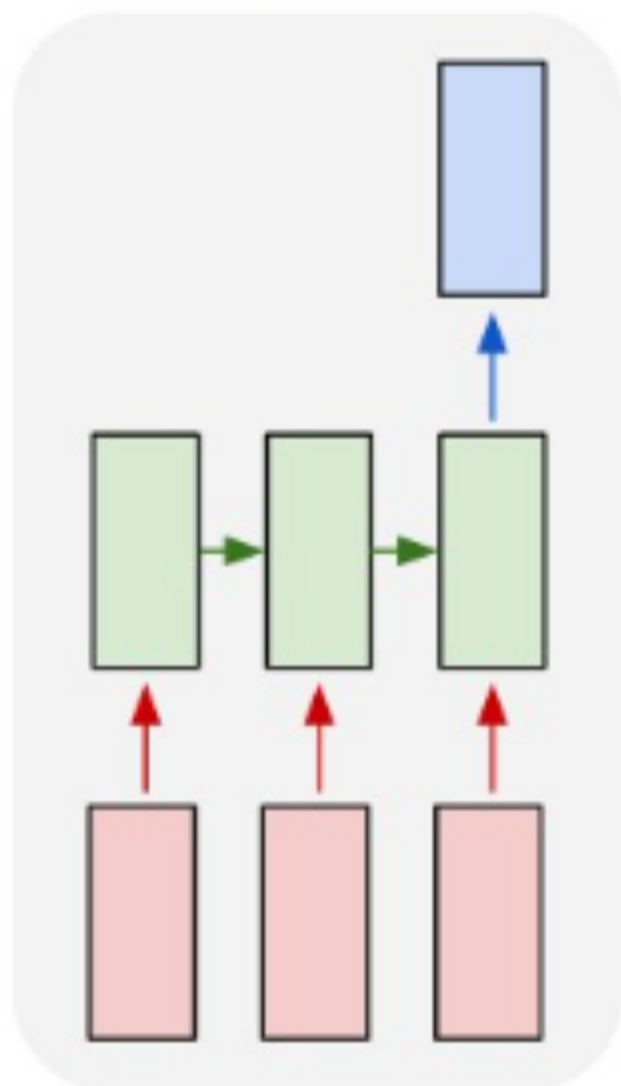
- Sequence output (e.g. image captioning takes an image and outputs a sentence of words).

one to many



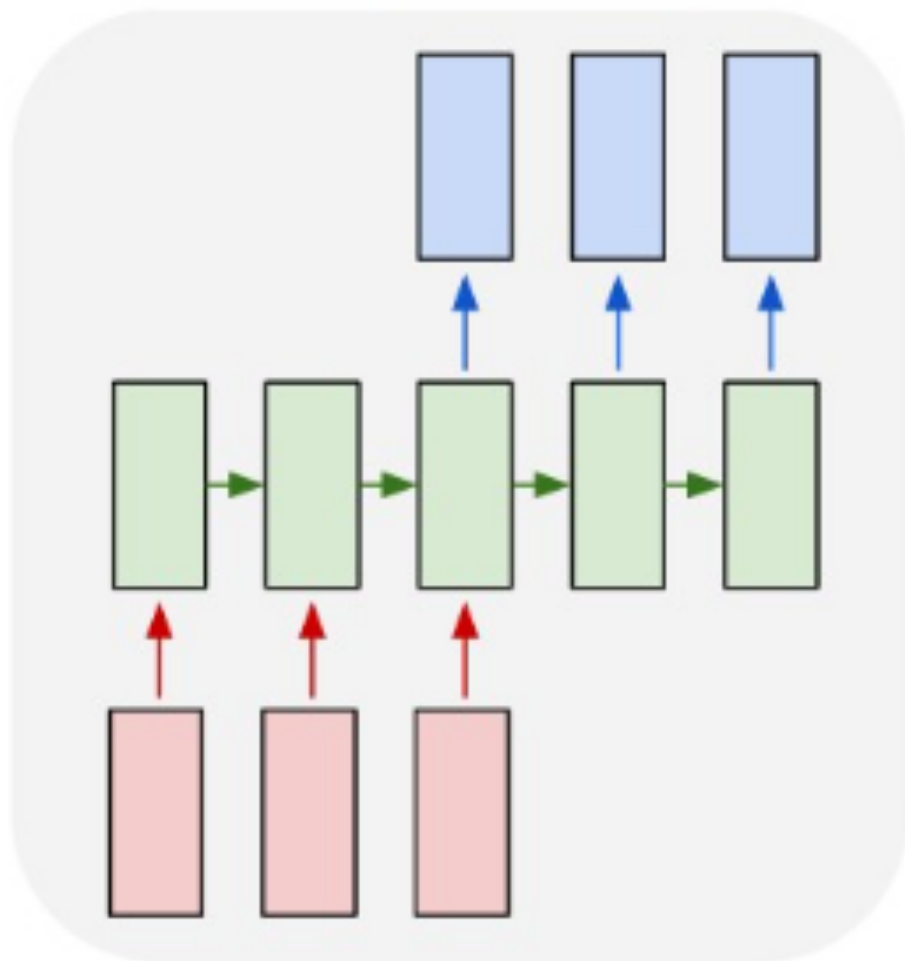
- Sequence input (e.g. sentiment analysis where a given sentence is classified as expressing positive or negative sentiment).

many to one



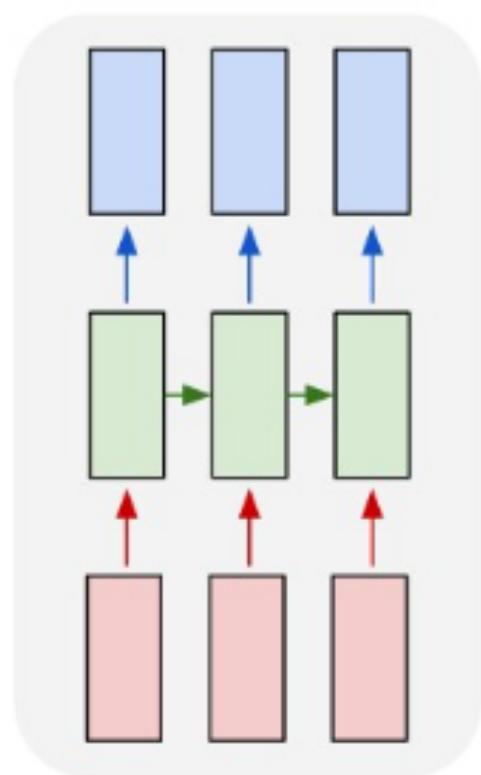
- Sequence input and sequence output (e.g. Machine Translation: an RNN reads a sentence in English and then outputs a sentence in French).

many to many



- Synced sequence input and output (e.g. video classification where we wish to label each frame of the video). Notice that in every case are no pre-specified constraints on the lengths sequences because the recurrent transformation (green) is fixed and can be applied as many times as we like.

many to many



- <http://www.cse.iitm.ac.in/~miteshk/CS7015/Slides/Handout/Lecture14.pdf>
- https://www.youtube.com/watch?v=7ZQgK_MV_t0
- <https://karpathy.github.io/2015/05/21/rnn-effectiveness/>



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