

# A Convolutional Neural Network for Modelling Sentences

## Link

[arXiv](#)

## Summary

- They propose a CNN model for learning representations that can handle sentences of varying length and can capture short and long range relations explicitly.
- They replace the usual narrow type of convolution with wide convolution which ensures all weights in the filter reach the entire sentence. The narrow convolution produces output sequence of length  $s - m + 1$  where  $s$  is input sequence length,  $m$  is filter size. The wide convolution produces output sequence of length  $s + m - 1$ . This convolution gives equal importance to words at the margin.

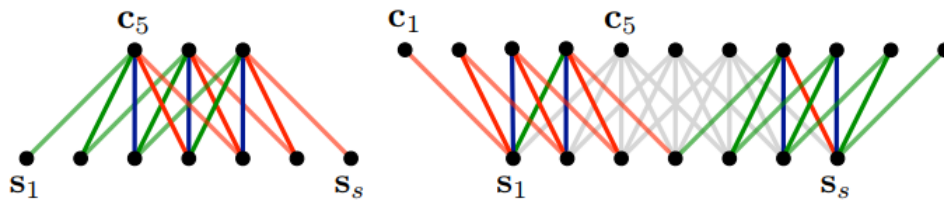


Figure 1: Narrow and wide types of convolution

- Instead of using max pool they use dynamic  $k$ -max pooling operation. This returns a sub-sequence of  $k$  maximum values in the sequence, instead of a single maximum value. Max pooling operation cannot distinguish whether a relevant feature occurs once or multiple times and it forgets the order in which the features occur. Moreover the pooling factor can be too excessive. The  $k$ -max pooling operation mitigates these problems.  $k$  can be also be a function of length of sentence and the depth of the network. The value of  $k$  can be seen as a model of number of values needed to describe a feature of corresponding order. For example first order features such as positive words occur at most  $k - 1$  times in the sentence, second order features such as negated phrase or clause occur at most  $k_2$  times etc.
- Filters in first layer can learn to recognize specific  $n$ -grams with size less than or equal to the filter size  $m$ . Filters in the higher layers can capture syntactic or semantic relations between non-continuous phrases that are far apart in the input sentence.
- The dynamic  $k$ -max pooling operation combined with wide convolution performs better than max pool time delayed convolutional neural network across variety of tasks.