How Grammarly Teaches Machines

by Oleksii Sliusarenko

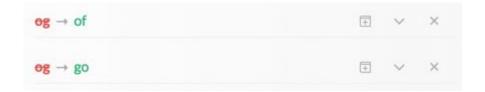




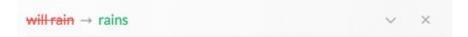


Problem statement

It is og course a good idea to og for a walk.



If it will rain tomorrow, I won't go for a run.





Problem statement

Tom is resembled by Andy.

Intransitive verb in passive voice Andy resembles Tom

If I cannot understand the material, the teacher will me.





Available data

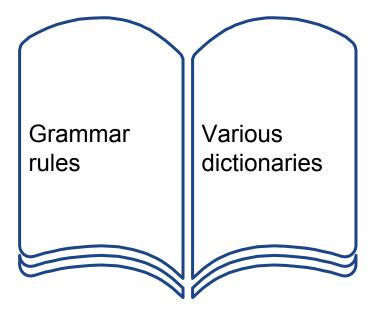


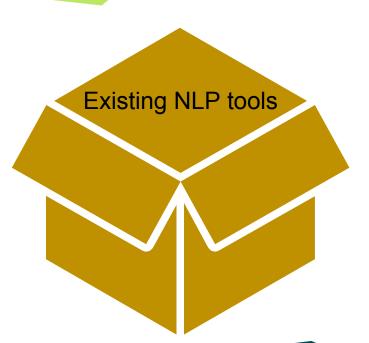
~0.1 Petabytes of plain texts



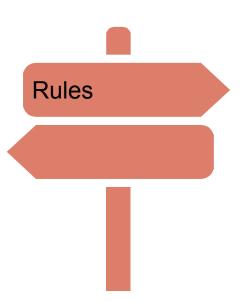
10³-10⁶ marked sentences

Available data





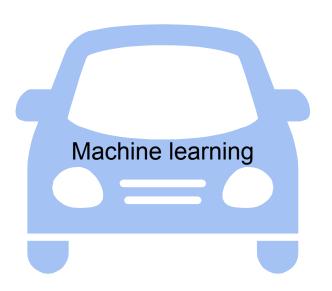
Grammarly uses all possible approaches

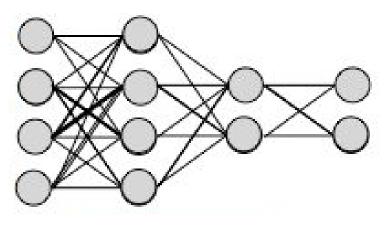




N-grams: roses are red \rightarrow 5082 roses is red \rightarrow 28 roses am red \rightarrow 0

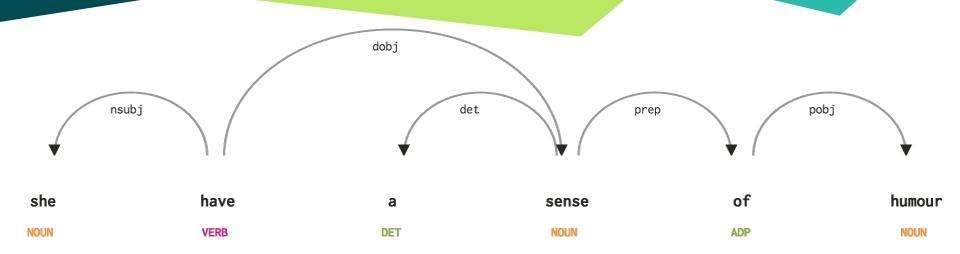
Grammarly uses all possible approaches





Deep learning

Rule based approach





she → third-person pronoun

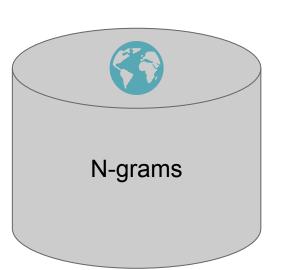
have → first-person verb

N-gram approach

Shane Bergsma, Dekang Lin, and Randy Goebel. 2009. Web-scale N-gram models for lexical disambiguation.

Of course it helps that the **whether** is almost always sunny and dry.

Of course it helps that the **weather** is almost always sunny and dry.



N-gram approach

```
a(5,0) = freq(it helps that the whether)
```

a(5,1) = freq(helps that the **whether** is)

a(5,2) = freq(that the **whether** is almost)

a(5,3) = freq(the whether is almost always)

a(5,4) = freq(whether is almost always sunny)

score = sum log(a(i, j))



N-grams

ML approach

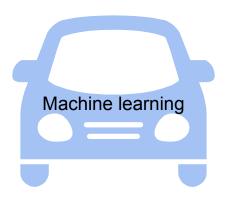
Matthew Shardlow. 2013. A Comparison of Techniques to Automatically Identify Complex Words

Complex:

Mike got spam today that contained gibberish.

Non-complex:

The book tells about people's irrationality.

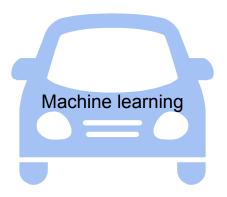


ML approach

Features:

- x_1 = word frequency
- x_2 = word length
- x_3 = syllable count
- \star x_4 = sense count
- $x_5 = synonym count$

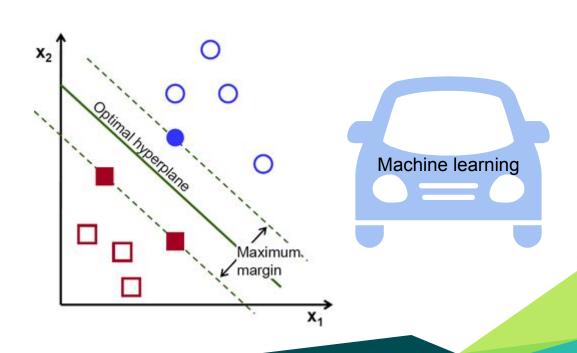




ML approach

Support vector machine:

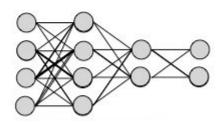
$$w_1^*x_1 + w_2^*x_2 + \dots + w_5^*x_5 > 0$$



Chengjie Sun, Xiaoqiang Jin, Lei Lin, Yuming Zhao, and Xiaolong Wang. 2015. Convolutional Neural Networks for Correcting English Article Errors



I watched **a** film you were talking about yesterday.



Deep learning

correct:

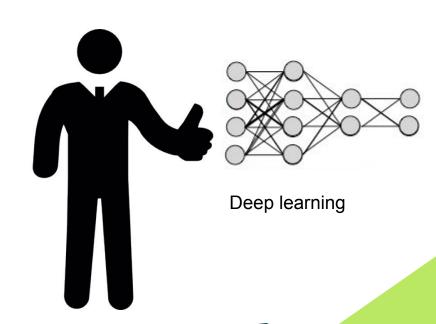
I watched **the** film you were talking about yesterday.

No features! One-hot encoding:

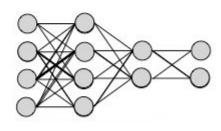
```
rose = (0,0,0,0,1,0,0,0)
red = (0,1,0,0,0,0,0,0)
violet = (0,0,0,0,0,0,0,1)
...
```

(in fact, word embeddings are used)

- Better quality
- Works for all: NLP, computer vision...

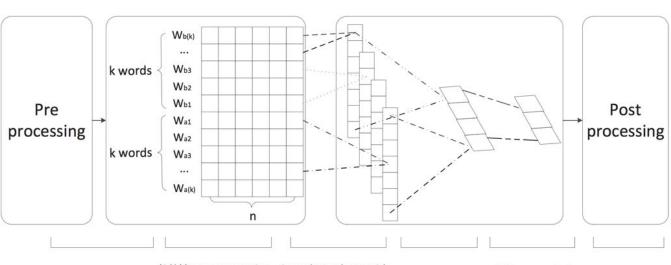


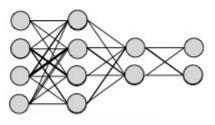
What's the catch?



Deep learning

Much harder algorithm:





Deep learning

Pre-processing

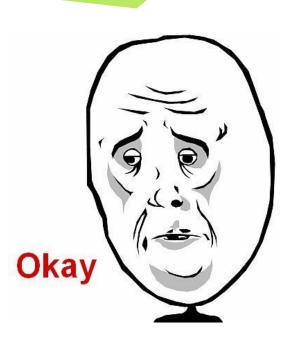
(2*k)*n representation of article context with non-static channel

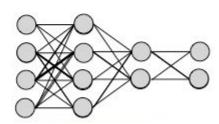
Convolution layer with multiple filter widths and feature maps

Max-over time Pooling Fully connected layer with dropout and softmax output

Post-processing

- Much harder algorithm
- Needs much more data
- Needs more computation





Deep learning

Difference from academia



Big data



High precision requirement



High speed requirement



Thanks!

Any questions?







Credits

Special thanks to all the people who made and released these **awesome resources** for free:

- Presentation template by <u>SlidesCarnival</u>
- Photographs by <u>Unsplash</u>