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# Natural Language Processing Deep Learning — Units 5 & 6

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Slides available at jonkrohn.com/talks

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### Take-Home Exercise: VGGNet

Talk through the purpose of every line in the [VGGNet notebook], including all of the following terms:

- ReLU
   \* SGD \* input laye
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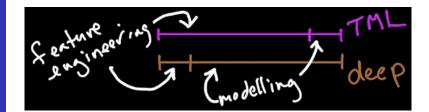
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# TML vs Deep Learning





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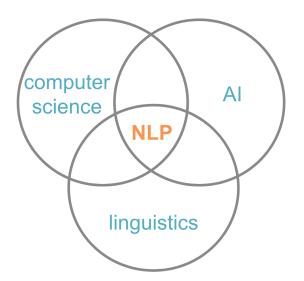
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- speech recognition (Echo, Siri, Cortana)
- search (typed into omnibox, spoken)
- classifying documents
- language translation
- chatbots



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# One-Hot Word Representations

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cat	0	0	0		Ø	
:						

Nunique\_words



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- spell checking
- synonym suggestions
- keyword search



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### Intermediate

### reading level

- extracting information
- predicting next words
- classification
- sequence generation
- time-series analysis



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- extracting information
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- extracting information
- predicting next words



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- reading level
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# Complex

- machine translation
- question-answering
- chatbots



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- question-answering
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### JR Firth (1957)

"You shall know a word by the company it keeps"



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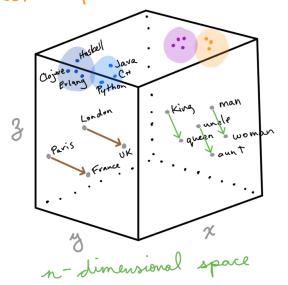
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# Vector Representations of Words





### Word Vector Arithmetic

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$$V_{\text{king}} - V_{\text{man}} + V_{\text{woman}} = V_{?}$$
 $V_{\text{jeff\_bezos}} - V_{\text{amazon}} + V_{\text{facebook}} = V_{?}$ 
 $V_{\text{windows}} - V_{\text{microsoft}} + V_{\text{google}} = V_{?}$ 
 $V_{\text{cu}} - V_{\text{copper}} + V_{\text{gold}} = V_{?}$ 



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[word2viz demo]



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### Word Representations

#### One-Hot

### **Vector-Based**

lack nuance

handle new words poorly

subjective

laborious, manual taxonomies

word similarity ignored

unwieldy with large vocabulary

extremely nuanced

seamlessly incorporate new words

driven by natural language data

fully-automatic

word similarity = closeness in space

accommodate large vocabularies



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# Word Representations

	predicts	relative strengths	
Skip-Gram (SG)	context given target	<ul><li>small data set</li><li>rare words</li></ul>	
CBOW	target given context	<ul><li>many times faster</li><li>slightly better for frequent words</li></ul>	



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# **Evaluating Word Vectors**

- 1 intrinsic
  - extrinsic



#### word2vec

# **Evaluating Word Vectors**

- intrinsic
- 2 extrinsic



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# word2vec Hyperparameters

- 1 n dimensions
- 2 window size (SG ~10, CBOW ~5)
- 3 n iterations
- data set size



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# word2vec Hyperparameters

- 1 n dimensions
- 2 window size (SG ~10, CBOW ~5)
- 3 n iterations
- 4 data set size



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# word2vec Hyperparameters

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- 3 n iterations
- 4 data set size



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[ creating word vectors notebook ]



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# Best Practices for Preprocessing NLP Data

[ NL preprocessing best practices notebook ]



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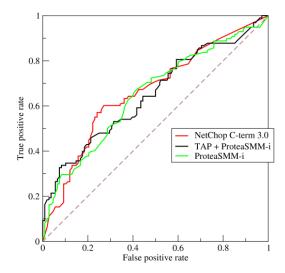
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## **Dense Net Classification**

[ dense sentiment classifier notebook ]



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### ConvNet Classification

[ convolutional sentiment classifier notebook ]



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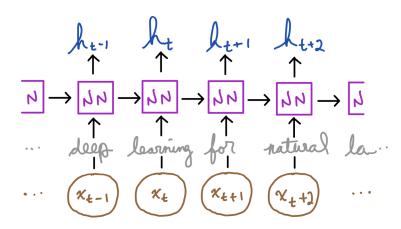
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# **RNNs** in Practice

[ rnn notebook ]



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# **LSTM Theory**

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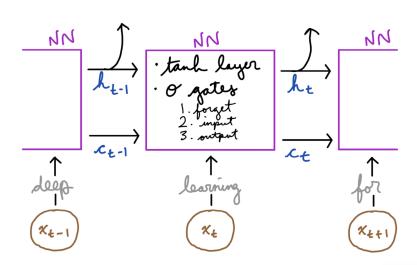
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## LSTMs in Practice

[ vanilla LSTM and GRU notebooks ]



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# **Bi-Directional LSTMs**

[ Bi-LSTM notebook ]



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### Stacked LSTMs

[ stacked LSTM and ye olde stackeroo notebooks ]



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## Parallel Network Architectures

[ multi-ConvNet notebook ]



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# Assessing Your Deep Learning Project III





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# Assessing

- 1 split your data
  - training set (80% for optimizing parameters)
  - validation set (10% for hyperparameters)
  - test set (10% don't touch yet!)
- 2 build and assess architecture
  - get above chance (simplifying problem, if necessary)
  - do existing performance benchmarks exist?
  - if not, use a simple architecture as benchmark
- (3) "teamwork makes the dream work" (?)



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- 3 "teamwork makes the dream work" (?)



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# Assessing Your Deep Learning Project III

- 1 split your data
  - training set (80% for optimizing parameters)
  - validation set (10% for hyperparameters)
  - test set (10% don't touch yet!)
- 2 build and assess architecture
  - get above chance (simplifying problem, if necessary)
  - do existing performance benchmarks exist?
  - if not, use a simple architecture as benchmark
- 3 "teamwork makes the dream work" (?)



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- 1 split your data
  - training set (80% for optimizing parameters)
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  - test set (10% don't touch yet!)
- 2 build and assess architecture
  - get above chance (simplifying problem, if necessary)
  - do existing performance benchmarks exist?
  - if not, use a simple architecture as benchmark
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