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Transformers, Large Scale Pre-trained Language Models & Applications in Conversational AI

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Agenda

- 1. Background:
 - a. Neural Networks
 - b. Recurrent Neural Networks
 - c. Language Models (LMs)
 - d. Word Embeddings

2. Transformers:

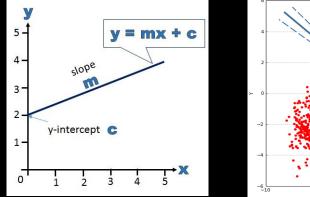
- a. Building blocks of Transformers
- b. Large Scale Pre-trained LMs
- c. From Millions to Trillions of Parameters
- d. Different kinds of Transformers: Encoders, Decoders, Seq2Seq

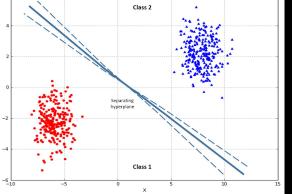
3. Applications:

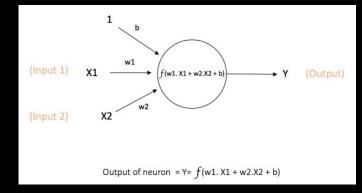
- a. Transformers for NL Technologies: State of the Art and Recommendations
- b. Evolution of Conversational AI and how Transformers are democratizing the space
- c. Transformers at Got It AI

Neural Networks Background

Fundamental Unit of Neural Networks and Deep Learning: Linear Unit





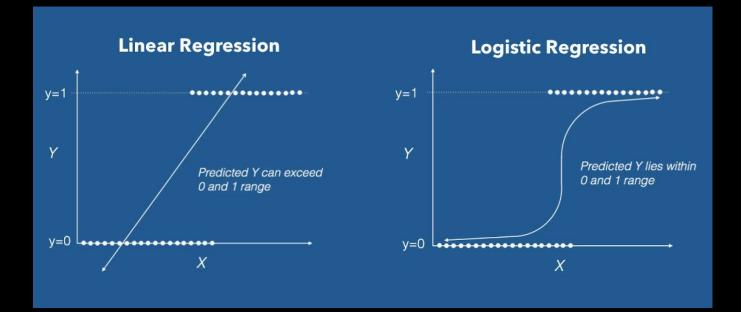


Linear Equation

Linear Classification

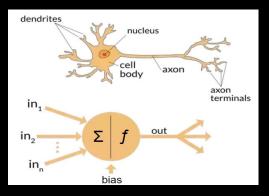
Mathematical representation of a Linear Unit / Perceptron

Linear Regression and Logistic Regression

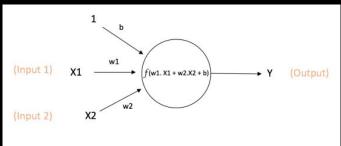


Neural Network

Neuron

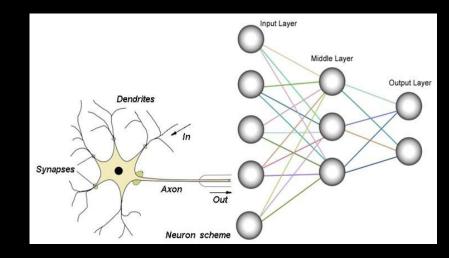


Linear Unit/Perceptron



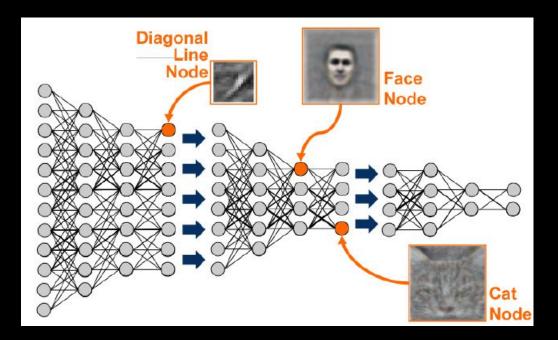
Output of neuron = Y= f(w1. X1 + w2.X2 + b)

Neural Network: Multi-Layer Perceptrons



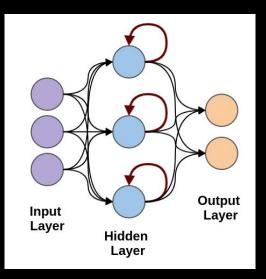
Deep Neural Networks

Fancy new term for Multi-layer Neural Networks with efficient ways of training



Recurrent Neural Network

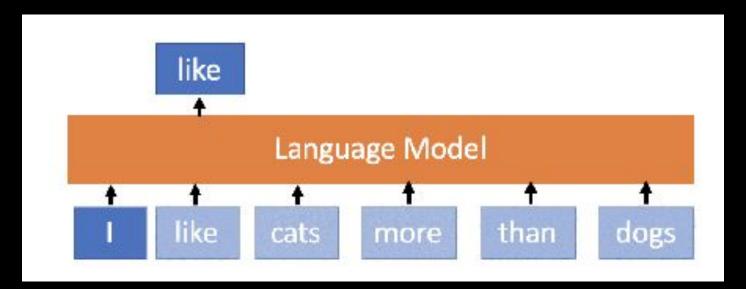
Fancy RNN Architecture



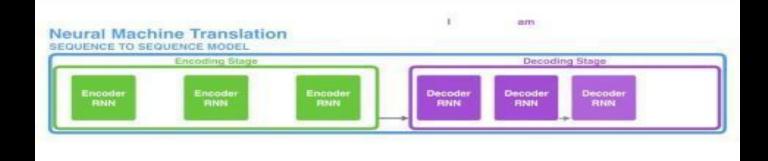
Reality: Memory, i.e. keep track of previous states for future prediction



Language Models

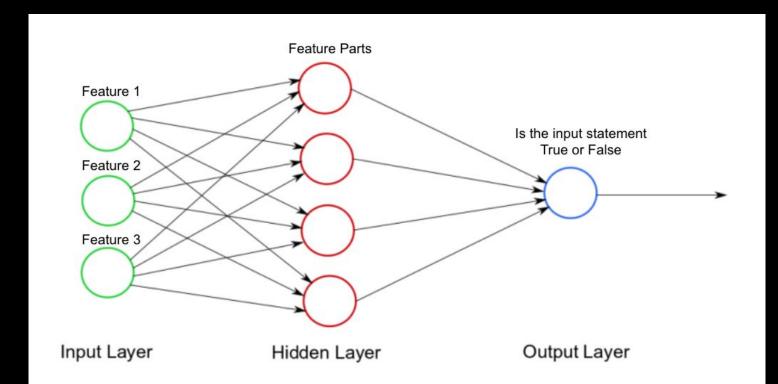


Seq2Seq Illustration



Source: https://jalammar.github.io/visualizing-neural-machine-translation-mechanics-of-seg2seg-models-with-attention/

Neural Network Features and Labels (Annotations)



Traditional NLP Featurization

Vocabulary Lookup Table

Sentence	the	is	visited		president	great	 obama
Obama is the president of the us	1	1	0	•••	. 1	0	 1
Obama visited the great wall of china	1	0	1		0	1	 1

One-hot encoding

Input/Sentence: Obama is the president of the us

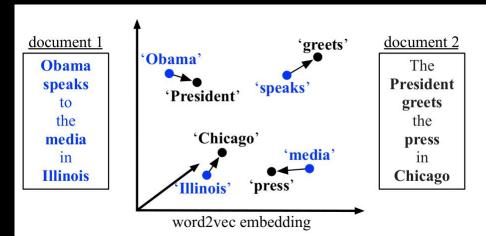
Features: Index of words appearing in the sentence

Vectorization: Initialize a vector of dimension V with all zeros except the ones appearing in the sentence [1, 1, 0, 0, 1, 0, 1, 0, 0, 1,, 1]

This vector is the feature representation of the given sentence

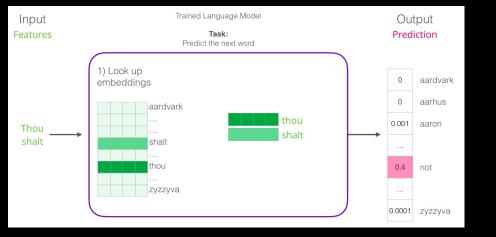
Word Embeddings: Semantic Repres. of Language Units

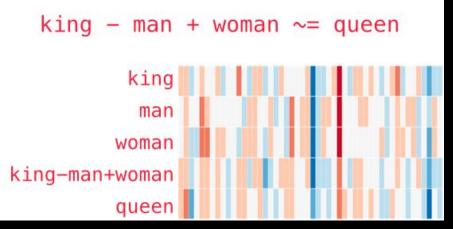
Word Embeddings



word	Dim 1	Dim 2	Dim	Dim 300
obama	0.1	0.5		0.9
president	0.15	0.4		0.85
chicago	0.3	0.7		0.6
illinois	0.25	0.72		0.58
the	0.01	0.02		0.07
press				

Word2Vec Illustration





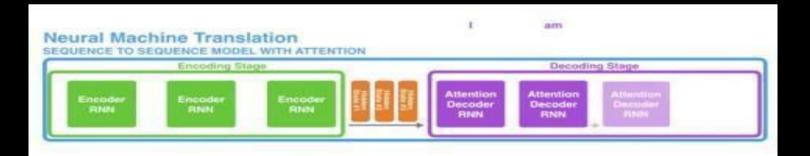
Are Word Embeddings contextual?

Transformers: Attention is all you need!

What are Transformers? Key Ideas

- Attention based Neural Networks
- Contextul Embeddings using:
 - Word position
 - Dynamic attention
- Auto-Regressive:
 - Use output generated at time "t" to generate at "t+1" time
- More Scalable:
 - Parallelizable: Vertical Architecture [RNN Horizontal]
 - Faster, Deeper,

Seq2Seq with Attention

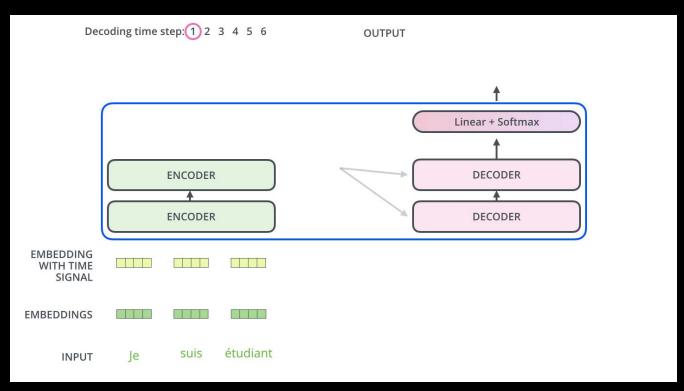


Source: https://jalammar.github.io/visualizing-neural-machine-translation-mechanics-of-seq2seq-models-with-attention/

Attention in Neural Networks

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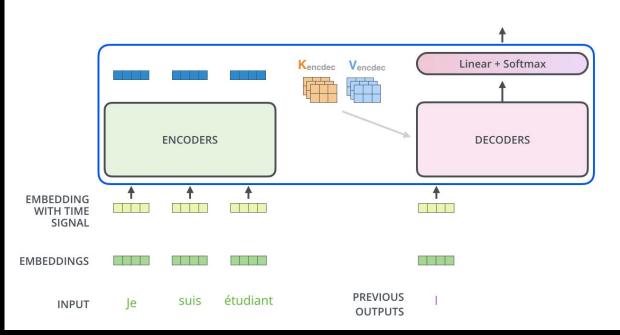
Dynamic Attention



Transformers Language Models

Decoding time step: 1 2 3 4 5 6

OUTPUT

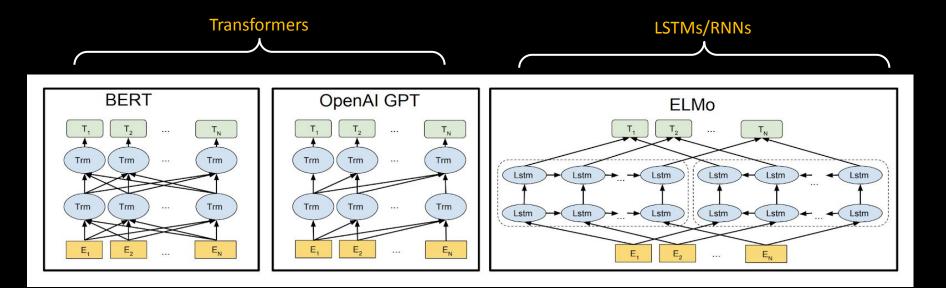


Autoregressive Language Models

There are a million ways

Source: https://eigenfoo.xvz/deep-autoregressive-models/

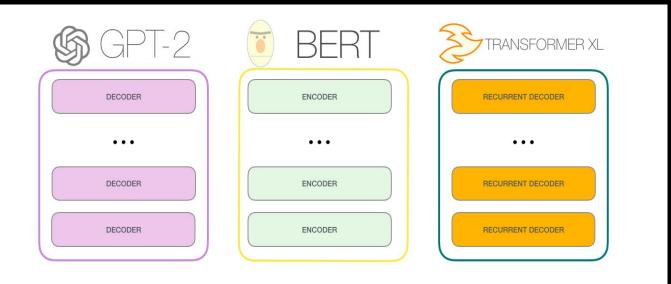
Contextual Word and Sentence Embeddings: For better NLU/NLP



Transformer, a Neural Architecture, which learns contextual word and sentence embeddings. Trained with 100s of millions, billions of parameters

Source: https://medium.com/@gauravghati/comparison-between-bert-gpt-2-and-elmo-9ad140cd1cda

BERT (RoBERTa) vs GPT (1, 2, 3) vs Transformer XL



Source: http://jalammar.github.io/illustrated-gpt2/

Encoders

- Usually bi-directional
- Great for getting good input representations
- Classification, Tagging, Extractive Summarization

Decoders:

- Uni-directional (forward)
- Great for generative tasks
- NLG, Abstractive Summarization, Translation, QA

Lots of Transformers!!

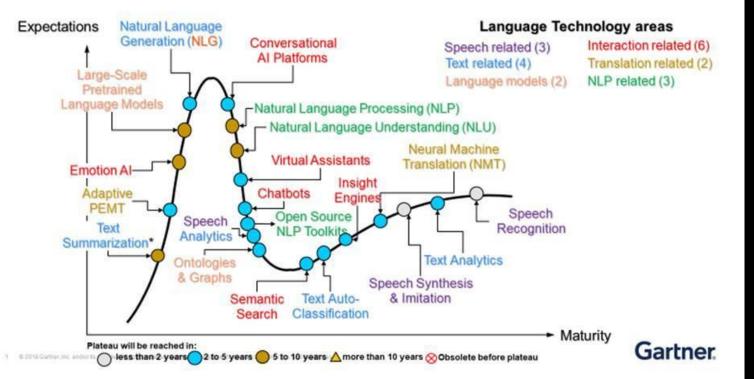
Which one to use?



Source: https://medium.com/analytics-vidhya/openai-gpt-3-language-models-are-few-shot-learners-82531b3d3122

Applications

Hype Cycle for Natural Language Technologies, 2020



Language Technology Areas

Multimodal

Image Captioning Video Understanding Multimodal Interaction OCR Voice - ASR, TTS NLU - Intent, Slot Filing Dialog Management - Action NLG - Response Selection/Generation

Conversational AI

Speech

ASR

Synthesis/TTS Voice Analytics

Translation

Text-text, Speech-speech, Text-speech, Speech-to-text

Language Models

Generation, NLG Summarization Story/Music Generation Transfer Learning (Pre-trained LMs)

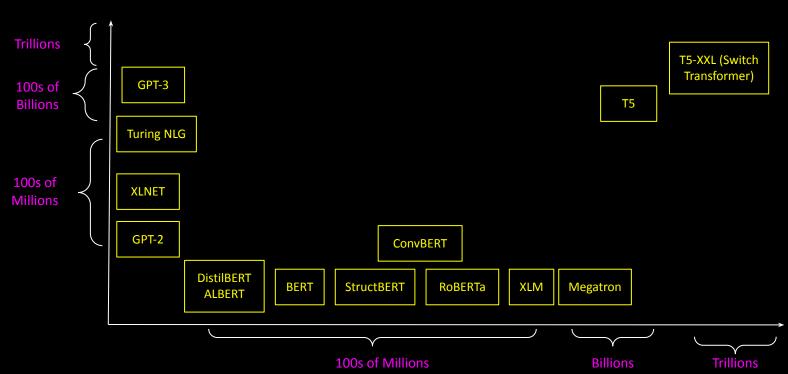
NLU

Sentiment Emotion/Empathy KB driven QA Semantic Parsing Interpretability

NLP

Topic Modeling Ranking: Search Syntactic Parsing Text Analytics Similarity POS/NER Tagging

Popular Transformer Architectures & their Applications



Decoders: Better for Generation Applications

Encoders: Better for Classification, Ranking Applications