

## Al & Smart Tech: Words, words, words





## Thanks for having me!

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## Rasa is the leading open source machine learning toolkit for developers to extend bots beyond answering simple questions



Allianz (11)



ERGO

## Introduction Structure of this talk

- I. Why words?
- II. Teaching words
- III. Teaching meaning
- IV. Research from Rasa
- V. Where will words go?



## Why words?

How and why do we use words as data?



#### Machine learning is all about information

We have a **wealth of information** to gather from words and literature

Whether text or speech, people use words to communicate, so good word representation **facilitates AI's integration** 



RHEH

# This is not an easy task

Why words? Words are nonsense

## "Uksrukuaktak"

{*lñupiat Eskimo dictionary*}





Why words? Vocabularies are vectors





Why words? Using word vectors

# "Which flavour ice cream?"



#### 







• How do words even work anyway?

They are the smallest element of language that can be uttered in isolation with objective or practical meaning.

• Why use them?

They are convenient representations which we use to externalise our thoughts. There is a wealth of knowledge to be acquired from literature

• How do we use words as data?

Separate at whitespace and assign each word a linearly-independent vector



## **Teaching words**

# Why word vectors are bad



Teaching words
One-hot vectors are bad

## Complete the sentence: "A holiday in \_\_\_\_"





Teaching words
One-hot vectors are bad

Rome	•	[1	0	0	0]
Sofia	•	[0	1	0	0]
London	•	[0	0	0	1]
Dog	•	[0]	0	1	0]

Language	Words in the Dictionary		
Korean	1,100,373		
Japanese	500,000		
Italian	260,000		
English	171,476		

Teaching words Cosine similarity



 $sim(A, B) = cos(\theta) = \frac{A \cdot B}{||A|| ||B||}$ 

#### sim(Rome, Sofia) = 0



**Teaching words** 

Learning word embeddings



egg?





$$sim(E(Man), E(King)) = 0.7$$



**Teaching words** 

#### **Properties of word embeddings**



*Embedding space captures relations* 



# OK, but does this make things better?

Teaching Words
Intent classification with Rasa NLU

Goal: classify user messages to a list of "intents"

greet goodbye thank\_you restaurant\_search book\_table confirm

Example intent list

#### I'm hungry for sushi



Example user message



## Teaching Words Intent Classification with word vectors

Embedded words sentence representation:  $\{v_1, ..., v_s\} \rightarrow \frac{1}{s} \sum v_i$ 



This works **embarrassingly well**, and generalises from little data



#### **Teaching words**

• Why one-hot vectors are bad

Being sparse causes slow learning, related words share no association

• What are word embeddings?

Word embeddings are a mapping from one-hots to a multi-dimensional space where synonymous words occupy similar regions

• What do word embeddings do?

Capture analogous information, increase task accuracy with low amounts of data



## **Teaching Meaning**

Why word embeddings are bad



Teaching Meaning
Word embeddings are bad

"No I don't want sushi"

### "I'd go hungry before eating sushi"



## Teaching Meaning Sentence Embeddings



Universal Sentence Encoder {Google Research, 2018}



Figure 4: Hierarchical ConvNet architecture.

Infersent encoder architecture {Facebook Research, 2017}



#### How good are sentence embeddings?

**Teaching Meaning** 



Universal Sentence Encoder {Google Research, 2018}

#### They don't fall into the **saliency trap**

# They capture meaning rather than word similarity

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#### How good are sentence embeddings?

# They have interesting **task** specialisation

They can capture information on a larger scope than word embeddings alone

premise	hypothesis	label
"Two women are embracing while	"Two woman are holding packages."	entailment
holding to go packages."		
A man is typing on a machine used	The man isn't operating a steno-	contradiction
for stenography	graph	
"A man is singing a song and play-	"A man is opening a package that	1.6
ing the guitar"	contains headphones"	
"Liquid ammonia leak kills 15 in	"Liquid ammonia leak kills at least	4.6
Shanghai"	15 in Shanghai"	

Infersent embedding tasks {Facebook Research, 2017}



#### **Teaching meaning**

• Why word embeddings are bad

Don't capture information from sentence structure like negation. Spelling errors and out-of-vocabulary words still cause problems

• How do we embed whole sentences?

Google use a deep averaging network, Facebook AI use multi-level representations, Skipthought takes the hidden representation of a whole sentence

• What do sentence embeddings do?

Allow for sentiments to be captured rather than synonymous relevance



# Time for a break

RHEF

## **Research from Rasa**

Embedding sentences for FAQ detection



## Research from Rasa **FAQ detection: hypothesis**

#### Can we use sentence embeddings to classify duplicate questions?

Question 1	Question 2	is_duplicate
How has your life changed since you became HIV positive?	How did your life change after you were diagnosed with HIV positive?	1
What is the defence mechanism in plants?	Is there any immune system in plants?	1
How long does WhatsApp keep the delivered data like images or etc. in the server?	How do I share GIF images in WhatsApp without converting the GIF into a video?	0
Within the context of dispute resolution, what is natural justice?	You earn a living to live well not to kill yourself with stress. What thoughts went through your mind after reading this sentence?	0

*QUORA duplicate questions dataset* 



1. Embed the questions using Google's pre-trained Universal Sentence Encoder

2. Measure **cosine similarity** between duplicate and non-duplicate questions

3. Determine if classification is possible



## Research from Rasa FAQ detection: results



Similarity between word embeddings

Similarity between sentence embeddings



# If you try 'em out, let us

# know



#### **Scalable minds**



**TLDR:** We built a custom <u>Rasa</u> Featurizer which appends a <u>Universal Sentence</u> <u>Encoding</u>, improving intent classification performance by overcoming inherent limitations of bag-of-words models. See this Gist for the code.

When building chatbots, the first step is usually to build an *intent classifier* in order to put arbitrary user messages (*"Can you reserve me a restaurant table?"*) to a pre-defined *intent* ( request\_restaurant\_reservation ). These are usually bag-of-words (BoW) models because they require little training data and are a surprisingly effective baseline.



#### Rasa is open source





## Where will words go?

#### ...are words bad?



#### Where will words go? Uksrukuaktak





Where will words go?

#### The Eskimo library



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#### Where will words go? The Eskimo library



#### Emily M. Bender (@emilymbender)

I am not interesting LM for LM sake --- I'm saying that if the "embeddings" that are sold as a representation of sentence (meaning) are learned in the context of an LM task they can't possibly represent meaning.  $\rightarrow$ 



Where will words go?

Grounding words in reality

## "Go grab my lunch"



#### **Untapped languages**

Where will words go?



Deception Detection in Videos {University of Maryland, 2018}



DensePose: Dense Human Pose Estimation in the Wild {Facebook, 2018}

RHSH

## Words are beautiful



#### Where will words go? Words are beautiful



HEN in the Courfe of human Events, it becomes needlay, for one People to diffave the Policial lands which have con-rest the Lank, the Experiment of any Mari-di Ruth, the Experiment of any Mari-di Ruth, the Experiment of any Mari-tic Policy of the Carlo station to which the Carlo Station of Alkauer Schernstein, and a station to which the Carlo Station of Alkauer Schernstein, and a station to which the Carlo Station of Alkauer Schernstein, and a station to which the Carlo Station of Station Schernstein, and a station to which the Carlo Station of Station Schernstein, and a station to which the Carlo Station of Station Schernstein, and a station to which the Station of Station Schernstein, and a station to station the Station of Station Schernstein, and a station to station the Station of Station Schernstein, and a station to station the Station of Station Schernstein, and a station to station the Station Schernstein Schernstein, and a station to station the Station Schernstein Schernstein, and a station the Station Schernstein Schernstein Schernstein Schernstein, and a station to station the Station Schernstein Scherenstein Schernstein Schernstein Schernstein Schernstein Sc

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## Words are disastrous



#### Where will words go? Words are disastrous

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#### Where will words go?

#### Text, type, talk



## Words are emissaries



## Thanks for listening to my words!



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