

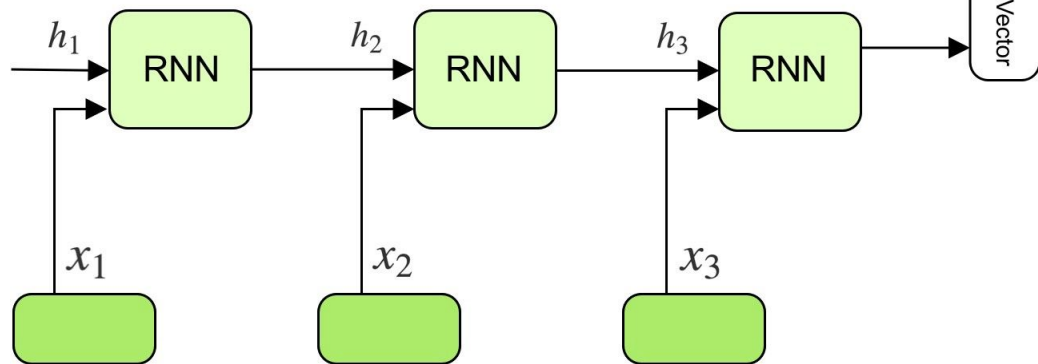
Seq2seq. Attention. Chatbots.

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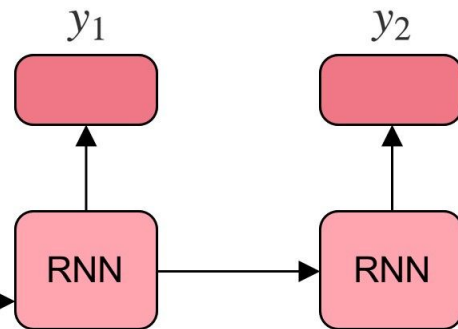
Seq2seq + attention

seq2seq (стандартная)

Encoder



Decoder

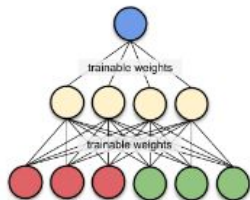


seq2seq + attention

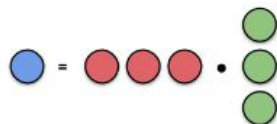
Виды attention (ИСТОЧНИК)



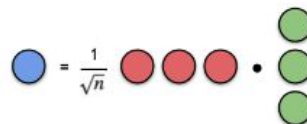
Additive / Concat



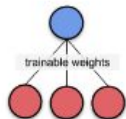
Dot product



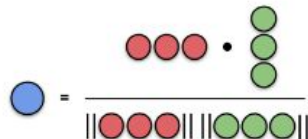
Scaled dot product



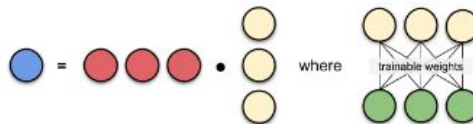
Location-based



Cosine similarity



General



Виды attention (ИСТОЧНИК)

Name	Alignment score function	Citation
Content-base attention	$\text{score}(s_t, h_i) = \text{cosine}[s_t, h_i]$	Graves2014
Additive(*)	$\text{score}(s_t, h_i) = \mathbf{v}_a^\top \tanh(\mathbf{W}_a [s_t; h_i])$	Bahdanau2015
Location-Base	$\alpha_{t,i} = \text{softmax}(\mathbf{W}_a s_t)$ Note: This simplifies the softmax alignment to only depend on the target position.	Luong2015
General	$\text{score}(s_t, h_i) = s_t^\top \mathbf{W}_a h_i$ where \mathbf{W}_a is a trainable weight matrix in the attention layer.	Luong2015
Dot-Product	$\text{score}(s_t, h_i) = s_t^\top h_i$	Luong2015
Scaled Dot-Product(^)	$\text{score}(s_t, h_i) = \frac{s_t^\top h_i}{\sqrt{n}}$ Note: very similar to the dot-product attention except for a scaling factor; where n is the dimension of the source hidden state.	Vaswani2017

Развитие идеи: Transformers

Чатботы

Какие чатботы бывают

- goal-oriented
 - попросить построить маршрут
 - попросить включить песню
 - спросить о погоде
- “болталка”
 - “мне скучно; кто себя создал; ...”
 - бот для психологической помощи
- всё и сразу :)
 - состоит из того и другого

How to train your dialog system

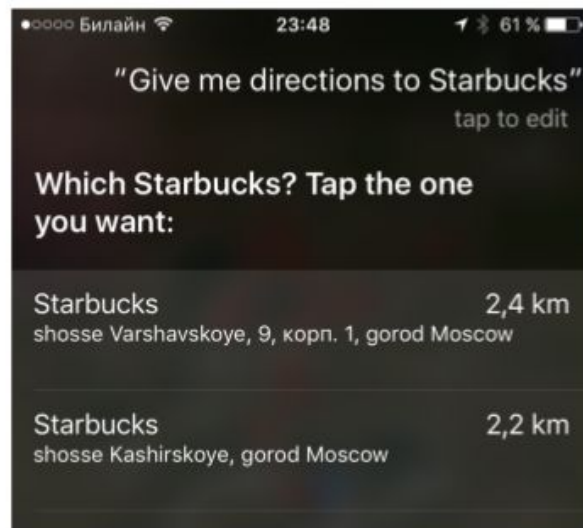
Andrey Zimovnov

Yandex



Siri

- Voice assistant from Apple
- How to make one?



Intent classification

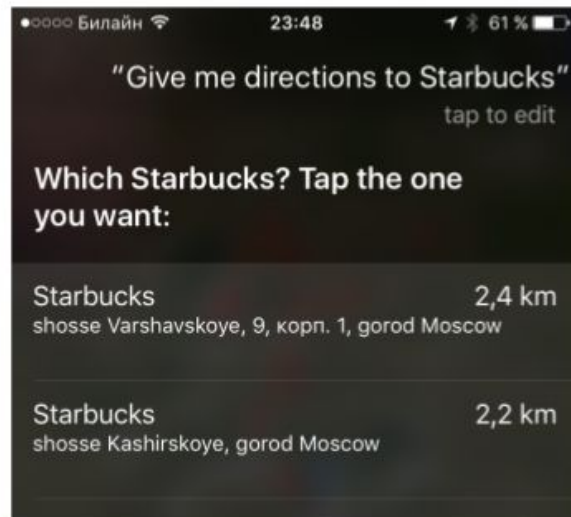
- Intent is **a class of an action** that assistant will execute
- Limited set of intents (navigation, music, notes, calendar, ... actions)



Intent: **nav.time.closest**



Intent: **nav.directions.closest**



Intent: **nav.directions**

Slot filling

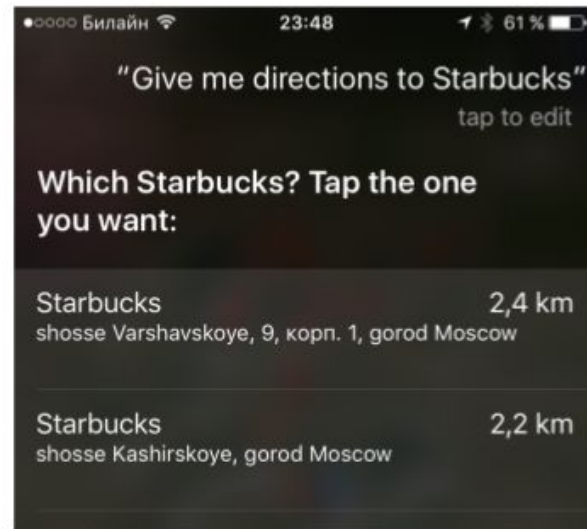
- Slot is a **parameter of an action**
- Limited set of slots (POIs, cities, streets, music artists, ... names)



Slots: @POI.NAME{Starbucks}



Slots: @POI.NAME{Starbucks}



Slots: @POI.NAME{Starbucks}

Two classification tasks

- Assume that we have an ASR result (text from speech)
- Marked up utterances (12k):
 - take me to **@NUMBER{ 1 0 8 2 }** **@STREET.NAME{ east el camino real }** **@CITY{ sunnyvale }** → **nav.route.start**
- Intent classification (80 classes):
 - **media.play, nav.find, phone.call, ...**
- Slot filling (25 slots):
 - **@MEDIA.ARTIST, @CITY, @CONTACT.NAME, ...**

Slot filling as classification

- Assign tag to each word:
 - **B-@CITY** – beginning of the slot **@CITY**
 - **I-@CITY** – continuation of the slot **@CITY**
 - **O** – this word is not used in slots

Sentence	first	class	fares	from	boston	to	denver
Slots	B-class_type	I-class_type	O	O	B-fromloc	O	B-toloc
Intent	airfare						

- Tag classification for each word

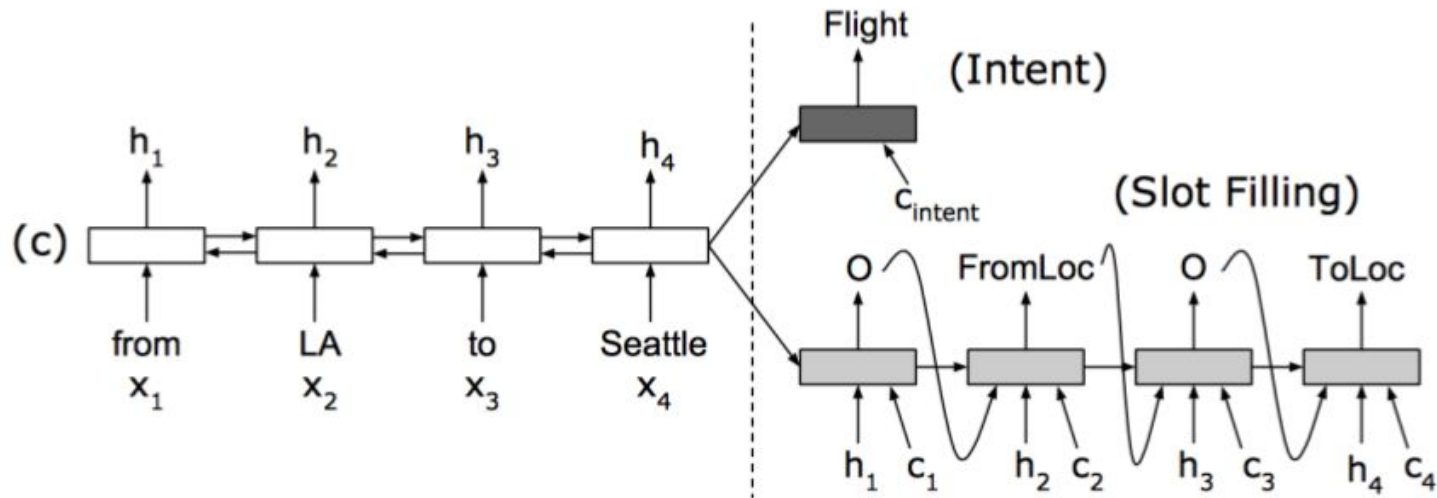
BIOES coding example

- Example for 4 lexicon dictionaries
- B, I, O, E, S are later encoded as one-hot vectors

[illegible]

Deep learning

- Encoder-decoder model for **joint** intent detection and slot filling
- Encoder is a bidirectional RNN (LSTM)
- With aligned inputs (\mathbf{h}_i on the right) and attention (\mathbf{c}_i on the right)



Joint training helps, so they say

- Paper results on ATIS (Airline Travel Information Systems) dataset:

Training	Slots F1	Intent % error
Independent training for slot filling	95.78	-
Independent training for intent detection	-	2.02
Joint training for slot filling and intent detection	95.87	1.57