Today

Natural language understanding

Two tasks (intent detection and slot filling): identify speaker's intent and extract semantic constituents from the natural language query

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

Figure: ATIS corpus sample with intent and slot annotation

- Intent detection is a classification task
- Slot filling is a sequence labelling task

NLU datasets: ATIS [1], Snips [2]

Joint intent detection and slot filling [3]

1. The encoder models is a biLSTM

2. The decoder is a unidirectional LSTM

3. At each step the decoder state s_i is: $s_i = f(s_{i-1}, y_{i-1}, h_i, c_i)$, where $c_i = \sum_j^T \alpha_{i,j} h_j$, $\alpha_{i,j} = \frac{\exp(e_{i,j})}{\sum_k^T \exp(e_{i,k})}$, $e_{i,k} = g(s_{i-1}, h_k)$

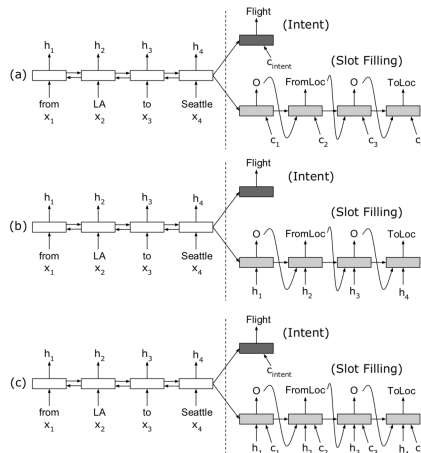The inputs are explicitly aligned. Costs from both decoders are back-propagated to the encoder.

Figure: Encoder-decoder models

Joint intent detection and slot filling [3]

- BiLSTM reads the source sequence
- forward RNN models slot label dependencies
- the hidden state h_i at each step is a concatenation of the forward state fh_i and backward state bh_i
- the hidden state is h_i combined with the context vector c_i
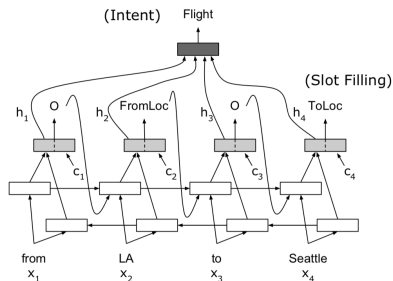- c_i is calculated as a weighted average of $h = (h_1, ..., h_T)$

Figure: RNN-based model

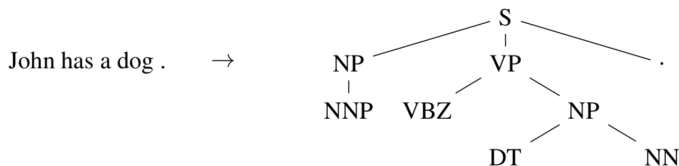| O	O	B-fromloc. city_name	O	B-toloc. city_name	O	O	B-depart_time. time_relative	B-depart_time. period_of_day
flight	from	cleveland	to	dallas	that	leves	before	noon

Figure: Attention weights

Today

1 Task oriented chat-bots

2 Constituency parsing

Grammar as a Foreign Language [4]

John has a dog . $\rightarrow$

John has a dog . $\rightarrow$ (S (NP NNP)$_{NP}$ (VP VBZ (NP DT NN)$_{NP}$)$_{VP}$.)$_S$

Figure: Example parsing task and its linearization
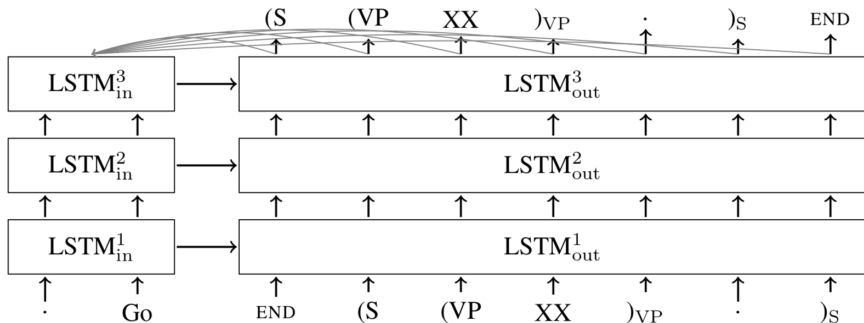
Grammar as a Foreign Language [4]

Figure: LSTM+attention encoder-decoder model for parsing

Grammar as a Foreign Language [4]

An important extension of the sequence-to-sequence model is by adding an attention mechanism. We adapted the attention model from [2] which, to produce each output symbol B_t, uses an attention mechanism over the encoder LSTM states. Similar to our sequence-to-sequence model described in the previous section, we use two separate LSTMs (one to encode the sequence of input words A_i, and another one to produce or decode the output symbols B_i). Recall that the encoder hidden states are denoted $(h_1, \ldots, h_{T_A})$ and we denote the hidden states of the decoder by $(d_1, \ldots, d_{T_B}) := (h_{T_A+1}, \ldots, h_{T_A+T_B})$.

To compute the attention vector at each output time t over the input words $(1, \ldots, T_A)$ we define:

$$
\begin{aligned}
u_i^t &= v^T \tanh(W_1' h_i + W_2' d_t) \\
a_i^t &= \mathrm{softmax}(u_i^t) \\
d_t' &= \sum_{i=1}^{T_A} a_i^t h_i
\end{aligned}
$$

The vector v and matrices W_1', W_2' are learnable parameters of the model. The vector u^t has length T_A and its i-th item contains a score of how much attention should be put on the i-th hidden encoder state h_i. These scores are normalized by softmax to create the attention mask a^t over encoder hidden states. In all our experiments, we use the same hidden dimensionality (256) at the encoder and the decoder, so v is a vector and W_1' and W_2' are square matrices. Lastly, we concatenate d_t' with d_t, which becomes the new hidden state from which we make predictions, and which is fed to the next time step in our recurrent model.

Reference I

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Reference II

📄 O. Vinyals, Ł. Kaiser, T. Koo, S. Petrov, I. Sutskever, and G. Hinton, "Grammar as a foreign language," in *Advances in neural information processing systems*, 2015, pp. 2773–2781.