The road to AGI

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Table of contents

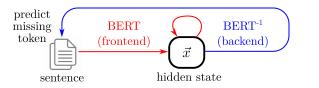
BERTBERT AGIAttention

Attention



BERT

• BERT induce representation (universality):



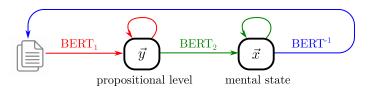
(1)

representation

- This implies that human-level AI can be *induced* from existing corpora,
- Such corpora can include items such as images, movies with dialogues / subtitles
- BERT

BERT AGI

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- 2 \ \ ·
- - BERT

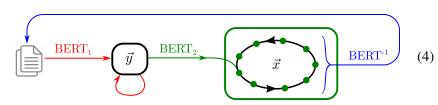


(2)

(3)

 \mathbf{r} 2 \vec{x} . permutation invariance.

- Word2Vec Word2Vec Sentence2Vec (concatenation) sentence
- \bullet \mathbb{P} $2^{\mathbb{P}}$
- \vec{x} = working memory $10 \ \vec{x} \ \mathbb{P}^{10}/\sim \sim \mathfrak{S}_{10} \ 2^{\mathbb{P}} \cong \prod_{n=0}^{\infty} \ \mathbb{P}^{n}/\mathfrak{S}_{n}$
- $\mathbb{P}^n/\mathfrak{S}_n$ \mathbb{P}^n \mathfrak{S}_n -
- \bullet \vec{x} \bullet



• BERT

Attention

- Seg2seg architecture BERT self-attention
- Seq2seq (encoder) x_i h_i

$$h_t = \text{RNN}_{encode}(x_t, h_{t-1})$$

•
$$h_i$$
 $c = q(h_1, ..., h_n)$. c

 \bullet c_t h_i

$$s_t = \text{RNN}_{decode}(y_t, s_{t-1}, c_t)$$

$$c_t = \sum_{i=1}^{n}$$

$$c_t = \sum_j \alpha_{ij} h_j$$

 $\alpha_{ij} = \operatorname{softmax}\{\langle s_i, h_i \rangle\}$

$$_{ij}h_{j}$$

(5)

(6)

(8)

Attention

- attention
- \circ s_t
- Attention $\geq 1 \sum h_j$ information retrieval
- $M N N^M$ mappings attention mappings
- "attention is all you need" inductive bias mapping
- mapping $f: \rightarrow$
- logic rules (applicable) rules
- s_t "search state" "" s_t
- •

