Question Answering using Attention and HMN

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Topics

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Problem

- Question answering is crucial task
 - Needs understanding of Natural Language
 - Understanding the world knowledge
- Many solutions exist
 - Select local maxima

Document: The Rankine cycle is sometimes referred to as a practical Carnot cycle ecause, when an efficient turbine is used, the TS diagram begins to resemble the Carnot cycle.

Question: WHAT IS THE RANKINE CYCLE SOMETIMES CALLED?

Ground truth: practical Carnot cycle Prediction: practical Carnot cycle

Dataset

- Human annotated datasets had less data but high quality
- Automated datasets are large scaled but low quality
- Using Stanford Question Answering Dataset (SQuAD)
 - Reading comprehension
 - 100,000 questions
 - 500 articles
 - Sources: like Wikipedia
 - 80k trainings set, 10k dev set, 10k test set
 - SQuAD answers can contain long phrases
 - Varied length questions

Related Work

- Statistical QA
 - Rule based algo
 - Relational queries
- Neural QA
 - Encoder decoder
- Bidirectional Attention Flow Mechanism
 - Character level
 - o CNN

Our Approach

- Based on Dynamic Coattention Network
 - Ouestion & context encoded into LSTM
 - Coattention matrix is generated using normalised attention
 - Document and question multiplied with co-attention
 - Passed through Bi-LSTM to get output
 - Decoder uses Highway Maxout Networks to recover from local maxima

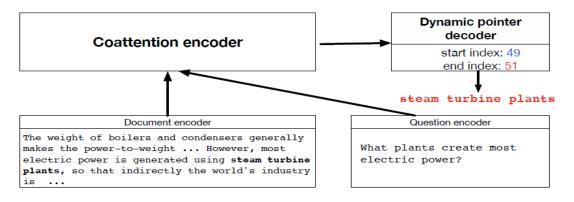
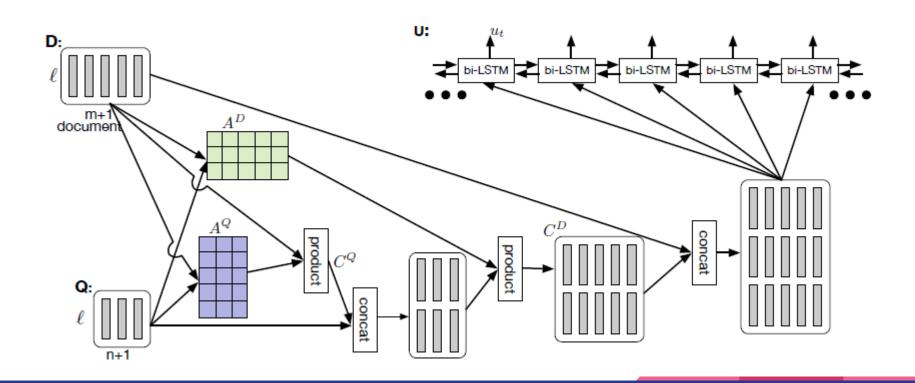


Figure used from : https://arxiv.org/pdf/1611.01604.pdf

Design

- Preprocessing
- Glove Embeddings: 300 dim
- Encoder
 - Question and document encoding
 - Co-attention encoding
- Decoder

Encoder Figure used from: https://arxiv.org/pdf/1611.01604.pdf



Decoder Figure used from: https://arxiv.org/pdf/1611.01604.pdf

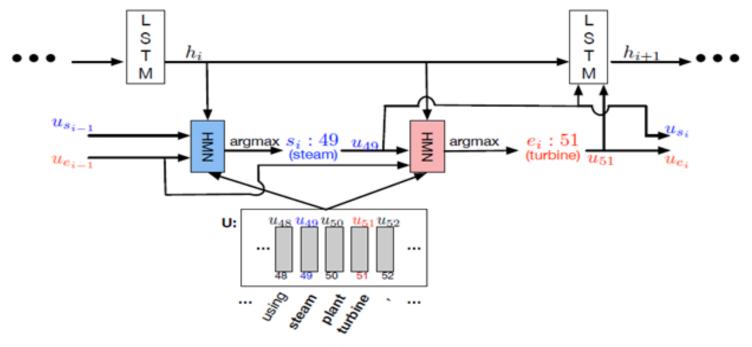


Figure 2. Decoder

Design (Encoder)

Document Matrix

$$D = [d_{1,} d_{2,} d_{3,} d_{4,...} d_{n,}] \in \mathbb{R}^{|x|(m+1)}$$

Encoded Document:

$$d_t = enc-LSTM (d_{t-1}, x_t^d)$$

Encoded Question:

$$q_t = \text{enc-LSTM} (q_{t-1}, x_t^q) \in \mathbb{R}^{|x|(m+1)}$$

Design (Encoder) contd...

Intermediate Question:

$$Q' = [q_1, q_2, ..., q_n] \in R^{|x|(n+1)}$$

Question Representation:

$$Q = tanh (W^{(Q)}Q' + b^{(Q)}) \in R^{1 \times (n+1)}$$

Coattention Encoder

$$L = D^T Q \in R^{(m+1) \times (n+1)}$$

Design (Encoder) Contd...

Affinity Matrix

$$A^Q = softmax(L) \in R^{(m+1) \times (n+1)}$$

$$A^{D}$$
 = softmax (L) $\in \mathbb{R}^{(n+1) \times (m+1)}$

Attention contexts

$$C^{Q} = D A^{Q}$$
 $\in \mathbb{R}^{|x|(m+1)}$

$$C^{D} = Q A^{D} \in \mathbb{R}^{|x|(m+1)}$$

Design (Encoder) contd...

Summeries:

$$C^{D} = [Q; C^{Q}] A^{D} \in R^{2l \times (m+1)}$$

Fusing Temporal Info:

$$U_t = Bi\text{-LSTM } (U_{t-1}, U_{t+1}, [d_t; c_t^D]) \in R^{2l}$$

Where, $U = [u_1, u_2, u_3, ..., u_m] \in R^{2l \times m}$

Design (Decoder)

```
h_i = dec-LSTM (h_{i-1}, [u_{si-1}; u_{ei-1}])
```

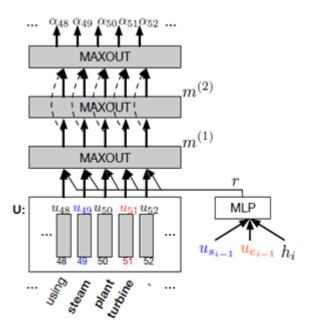
New start and end pos:

```
s_i = argmax (\alpha_{i_1} .... \alpha_{m_i})

e_i = argmax (\beta_{i_1} .... \beta_{m_i})

Where, \alpha_{t_i} = HMN_{start} (u_t, h_{i_i} u_{si-1}, u_{ei-1})
```

Highway Maxout Networks



$$\begin{split} & \text{HMN } (u_t, \, h_i, \, u_{si\text{-}1}, \, u_{ei\text{-}1}) = \text{MAX} (W^{(3)} \, [m_t^{(1)}; \, m_t^{(2)}] + b^{(3)}) \\ & r = \text{tanh } (W^{(D)} \, [h_i^{(1)}; \, u_{si\text{-}1}; \, u_{ei\text{-}1}]) \\ & m_t^{(1)} = \text{MAX} (W^{(1)} \, [u_t; \, r + b^{(1)}) \\ & m_t^{(2)} = \text{MAX} (W^{(2)} \, m_t^{(1)} + b^{(2)}) \end{split}$$

Experimental Setup

- Split word and created a dictionary
- Word is mapped to glove vectors
- Divided data into 50 epochs (1356 sized)
- Max_len = 600
- Hidden dim = 200
- Batch size = 64
- LR = 0.001
- LSTM assigned zero hidden state
- Softmax cross entropy
- Adam optimizer

Result

Set	F1 (Our Model)	F1 (Baseline)
Dev	8.6	2.3
Test	4.9	1.8

Sample Output 1

Document: British colonial rule introduced Western elements of culture to Burma. Burma 's education system is modelled after that of the United Kingdom. Colonial architectural influences are most evident in major cities such as Yangon. Many ethnic minorities, particularly the Karen in the southeast and the Kachin and Chin who populate the north and northeast, practice Christianity. According to the The World Factbook, the Burman Document 1: British colonial rule introduced Western elements of culture to Burma. Burma 's education system is modelled after that of the United Kingdom. Colonial architectural influences are most evident in major cities such as Yangon. Many ethnic minorities, particularly the Karen in the southeast and the Kachin and Chin who populate the north and northeast, practice Christianity. According to the The World Factbook, the Burman population is 68% and the ethnic groups constitute 32%. However, the exiled leaders and organisations claims that ethnic population is 40%, which is implicitly contrasted with CIA report (official US report).

Question:

Do opinions differ on the numbers for that population is considered not a part of the mainstream?

Prediction: According to the The World Factbook, the Burman population is 68 % and the ethnic groups constitute 32 %. However, the exiled leaders and and organisations claims that ethnic population is 40 %.

Groud Truth: exiled leaders and organisations claims that ethnic population is 40 %, which is implicitly contrasted with CIA report (official US report).

Sample Output 2

Document 2:

Food is passed from the kitchen to the stube, where the dining room table is placed. Some meals are communal, such as fondue, where a pot is set in the middle of the table for each person to dip into. Other meals are still served in a traditional manner on carved wooden plates. Furniture has been traditionally elaborately carved and in many Alpine countries carpentry skills are passed from generation to generation.

Question:

The food is passed from the kitchen to where?

Prediction: Food is passed from the kitchen to the stube, where the dining room table is placed

. Some meals are communal, such as fondue, where a pot is set in

Groud Truth: the stube

Conclusion

- DCN has proved good results on question answering tasks.
- Need to run more iterations to see the actual results.

References

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Thank you