End-To-End Memory Networks

Sainbayar Sukhbaatar, Arthur Szlam, Jason Wetson, Rob Fergus
Dept. Of Computer Science
Courant Institute, NYU

&

Facebook AI Research New York

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Motivation

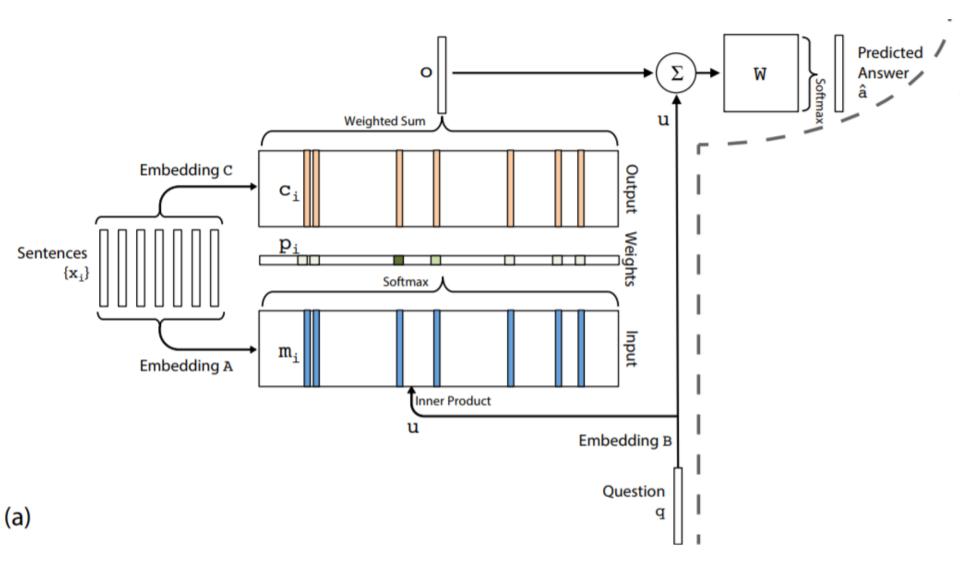
- Make a model that can perform many computational steps to answer a question.
- Make a model that describes dependencies in sequential data.
- I.E. sequential reasoning
- Lightweight & easily trainable

Motivation over MemNN

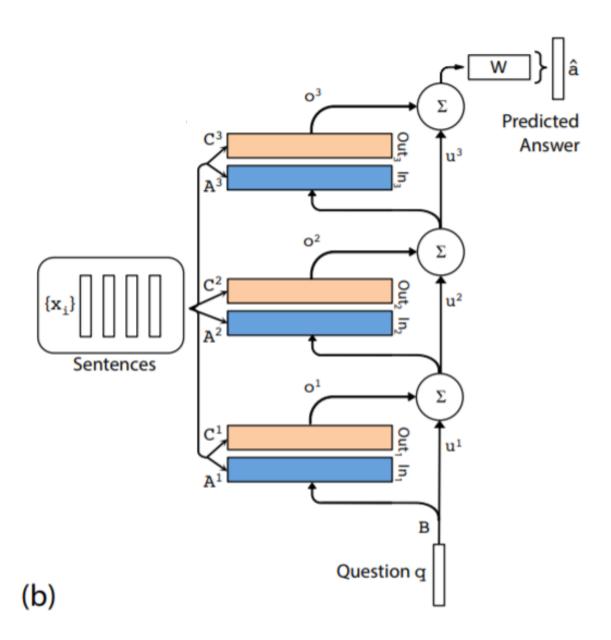
- End-To-End Trainable
- Far less supervision
- More generalizable

Overview of Model

- Variables:
 - Discrete set of inputs (x_i)
 - A query (q)
 - Produce an answer (a)
- Static Memory Bank
- Multiple Hops

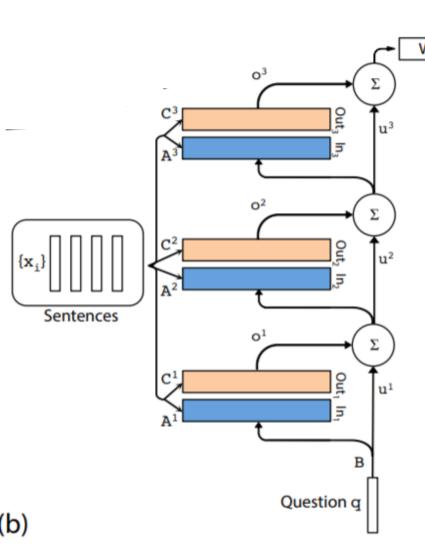


$$p_i = \text{Softmax}(u^T m_i)$$
 $o = \sum_i p_i c_i$ $\hat{a} = \text{Softmax}(W(o + u))$



Weight Tying

Predicted Answer



Adjacent:

$$-A^{k+1}=C^k$$

$$-W^T = C^K$$

$$- B = A^{1}$$

Layer-wise (RNN-like):

$$-A^1 = ... = A^k$$
, $C^1 = ... = C^k$

$$-u^{k+1} = Hu^{k} + o^{k}$$

Sentence Representation

Bag-of-words

$$-m_i = \sum_j Ax_{ij}$$

Position Encoding (PE)

$$-m_i = \sum_j l_j \cdot Ax_{ij}$$

Temporal Encoding (TE)

$$-m_i = \sum_j Ax_{ij} + T_A(i)$$

Synthetic QA Experiments

Sam walks into the kitchen.

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Sam picks up an apple.
Sam walks into the bedroom.
Sam drops the apple.
Q: Where is the apple?
A. Bedroom
Brian is a lion.
Julius is a lion.
Julius is white.
Bernhard is green.
O: What color is Brian?
A. White
```

Similarity to Attention

| Story (16: basic induction) | Support | Hop 1 | Hop 2 | Hop 3 | |
|------------------------------------|--------------------|-------|-------|-------|--|
| Brian is a frog. | yes | 0.00 | 0.98 | 0.00 | |
| Lily is gray. | | 0.07 | 0.00 | 0.00 | |
| Brian is yellow. | yes | 0.07 | 0.00 | 1.00 | |
| Julius is green. | | 0.06 | 0.00 | 0.00 | |
| Greg is a frog. | yes | 0.76 | 0.02 | 0.00 | |
| What color is Greg? Answer: yellow | Prediction: yellow | | | | |

| Story (18: size reasoning) | Support | Hop 1 | Hop 2 | Hop 3 | | | |
|---|---------|-------|-------|-------|--|--|--|
| The suitcase is bigger than the chest. | yes | 0.00 | 0.88 | 0.00 | | | |
| The box is bigger than the chocolate. | A 2.04 | 0.04 | 0.05 | 0.10 | | | |
| The chest is bigger than the chocolate. | yes | 0.17 | 0.07 | 0.90 | | | |
| The chest fits inside the container. | | 0.00 | 0.00 | 0.00 | | | |
| The chest fits inside the box. | | 0.00 | 0.00 | 0.00 | | | |
| Does the suitcase fit in the chocolate? Answer: no Prediction: no | | | | | | | |

NOTE: This model does not use the "support" label during training

Results

| | E | MemN | I2N | | | | | |
|------------------------------|------------|------|-------|-------|--------|--------|-------|-------|
| | Strongly | | | 1 hop | 2 hops | 3 hops | PE | PE LS |
| | Supervised | LSTM | MemNN | PE LS | PE LS | PE LS | LS RN | LW |
| Task | MemNN [22] | [22] | WSH | joint | joint | joint | joint | joint |
| 1: 1 supporting fact | 0.0 | 50.0 | 0.1 | 0.8 | 0.0 | 0.1 | 0.0 | 0.1 |
| 2: 2 supporting facts | 0.0 | 80.0 | 42.8 | 62.0 | 15.6 | 14.0 | 11.4 | 18.8 |
| 3: 3 supporting facts | 0.0 | 80.0 | 76.4 | 76.9 | 31.6 | 33.1 | 21.9 | 31.7 |
| 4: 2 argument relations | 0.0 | 39.0 | 40.3 | 22.8 | 2.2 | 5.7 | 13.4 | 17.5 |
| 5: 3 argument relations | 2.0 | 30.0 | 16.3 | 11.0 | 13.4 | 14.8 | 14.4 | 12.9 |
| 6: yes/no questions | 0.0 | 52.0 | 51.0 | 7.2 | 2.3 | 3.3 | 2.8 | 2.0 |
| 7: counting | 15.0 | 51.0 | 36.1 | 15.9 | 25.4 | 17.9 | 18.3 | 10.1 |
| 8: lists/sets | 9.0 | 55.0 | 37.8 | 13.2 | 11.7 | 10.1 | 9.3 | 6.1 |
| 9: simple negation | 0.0 | 36.0 | 35.9 | 5.1 | 2.0 | 3.1 | 1.9 | 1.5 |
| 10: indefinite knowledge | 2.0 | 56.0 | 68.7 | 10.6 | 5.0 | 6.6 | 6.5 | 2.6 |
| 11: basic coreference | 0.0 | 38.0 | 30.0 | 8.4 | 1.2 | 0.9 | 0.3 | 3.3 |
| 12: conjunction | 0.0 | 26.0 | 10.1 | 0.4 | 0.0 | 0.3 | 0.1 | 0.0 |
| 13: compound coreference | 0.0 | 6.0 | 19.7 | 6.3 | 0.2 | 1.4 | 0.2 | 0.5 |
| 14: time reasoning | 1.0 | 73.0 | 18.3 | 36.9 | 8.1 | 8.2 | 6.9 | 2.0 |
| 15: basic deduction | 0.0 | 79.0 | 64.8 | 46.4 | 0.5 | 0.0 | 0.0 | 1.8 |
| 16: basic induction | 0.0 | 77.0 | 50.5 | 47.4 | 51.3 | 3.5 | 2.7 | 51.0 |
| 17: positional reasoning | 35.0 | 49.0 | 50.9 | 44.4 | 41.2 | 44.5 | 40.4 | 42.6 |
| 18: size reasoning | 5.0 | 48.0 | 51.3 | 9.6 | 10.3 | 9.2 | 9.4 | 9.2 |
| 19: path finding | 64.0 | 92.0 | 100.0 | 90.7 | 89.9 | 90.2 | 88.0 | 90.6 |
| 20: agent's motivation | 0.0 | 9.0 | 3.6 | 0.0 | 0.1 | 0.0 | 0.0 | 0.2 |
| Mean error (%) | 6.7 | 51.3 | 40.2 | 25.8 | 15.6 | 13.3 | 12.4 | 15.2 |
| Failed tasks (err. $> 5\%$) | 4 | 20 | 18 | 17 | 11 | 11 | 11 | 10 |
| On 10k training data | | | | | | | | |
| Mean error (%) | 3.2 | 36.4 | 39.2 | 24.5 | 10.9 | 7.9 | 7.5 | 11.0 |
| Failed tasks (err. $> 5\%$) | 2 | 16 | 17 | 16 | 7 | 6 | 6 | 6 |

Language Modeling

| | Penn Treebank | | | | Text8 | | | | | |
|-----------|---------------|------|--------|--------|-------|--------|------|--------|--------|-------|
| | # of | # of | memory | Valid. | Test | # of | # of | memory | Valid. | Test |
| Model | hidden | hops | size | perp. | perp. | hidden | hops | size | perp. | perp. |
| RNN [15] | 300 | - | - | 133 | 129 | 500 | - | - | - | 184 |
| LSTM [15] | 100 | - | - | 120 | 115 | 500 | - | - | 122 | 154 |
| SCRN [15] | 100 | - | - | 120 | 115 | 500 | - | - | - | 161 |
| MemN2N | 150 | 2 | 100 | 128 | 121 | 500 | 2 | 100 | 152 | 187 |
| | 150 | 3 | 100 | 129 | 122 | 500 | 3 | 100 | 142 | 178 |
| | 150 | 4 | 100 | 127 | 120 | 500 | 4 | 100 | 129 | 162 |
| | 150 | 5 | 100 | 127 | 118 | 500 | 5 | 100 | 123 | 154 |
| | 150 | 6 | 100 | 122 | 115 | 500 | 6 | 100 | 124 | 155 |
| | 150 | 7 | 100 | 120 | 114 | 500 | 7 | 100 | 118 | 147 |
| | 150 | 6 | 25 | 125 | 118 | 500 | 6 | 25 | 131 | 163 |
| | 150 | 6 | 50 | 121 | 114 | 500 | 6 | 50 | 132 | 166 |
| | 150 | 6 | 75 | 122 | 114 | 500 | 6 | 75 | 126 | 158 |
| | 150 | 6 | 100 | 122 | 115 | 500 | 6 | 100 | 124 | 155 |
| | 150 | 6 | 125 | 120 | 112 | 500 | 6 | 125 | 125 | 157 |
| | 150 | 6 | 150 | 121 | 114 | 500 | 6 | 150 | 123 | 154 |
| | 150 | 7 | 200 | 118 | 111 | - | - | - | - | - |

Conclusion

- Outperforms all baselines with the same level of supervision (LSTMs etc.)
- Slightly worse than a strongly supervised
 Memory Network, but it was trained without
 supporting facts, so it can be easily trained in
 more general settings.
- On language modeling, outperforms RNNs and LSTMs