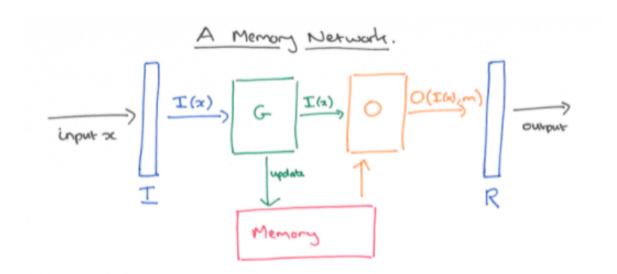


BTP Report

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Memory Network (Reimplementation)



Overview

Memory networks have hidden states and weights. The aim is to make machines learn the meaning and inference of sentences by training them with fully supervised learning. After the training, the machine starts to infer meaning from a given story and becomes able to answer questions about that particular story. The machine memory can be read and written.

Memory networks have larger storage compared to other approaches which try to solve the same kind of Question Answering problems. RNNs are also used for solving question answering problems. They are trained and then they predict the words for a given question. However, the memory of RNNs is not enough to remember facts from the past accurately. On the other hand, memory networks are able to write to and read from external memory to remember the past. The model is trained in fully supervised techniques in order to use its memory effectively.

Model

Model Architecture

A memory network has 4 components I, G, O, and R as follows:

- I (input feature map): Takes the input, in this case, lines in a text, and returns its feature
- representation.
- G (generalization): Takes the feature representation of a line and adds it to the memory.
- O (output feature map): Takes the feature representation of a question, finds all of the related
- memory states with the question, and returns them.
- R (response): Takes both the feature representation of the question and the related memory
- states and finds the answer to this question among the dictionaries.

The following is the flow of a given a sentence x inside the model:

- 1. $x_{\text{feature}} = I(x)$ converts x into the internal feature representation.
- 2. $G(x_{feature}, memory)$ updates the memory by adding the $x_{feature}$ to the given memory.
- 3. output_feature = O(question_feature, memory) computes the output features which are highly related to question_feature.
- 4. 4. R(output_feature, vocabulary) produces the textual response according to the output_feature.

During training and testing, the flow above is followed. The only difference between training and testing is the parameter update. Parameters are updated while training, however; they are not updated during the test time. The detailed role of each component is as follows:

I component: This component uses the Bag of Words approach to convert a sentence into an inner feature representation. Firstly, the dictionary is created by assigning a number beginning from 1 to each unseen word in the text. This component uses this dictionary to find which vocabularies are present in a given sentence x. It returns a matrix of dimension $V \times 1$ where V is the length of the dictionary. The values in this matrix are all 0 except the indexes which correspond to a word in the sentence x (these are 1).

G component: This component adds the feature representation of a new sentence to the next available slot in the memory. In this case, the memory is reset at the beginning of each story, and in this way, the memory does not become huge and unrelated memory slots are removed. If larger memory is necessary for problems such as large-scale QA dataset cases where there are 14M statements, a better approach would be fixing the size of the memory and forgetting the least used memory slot when the memory is full. O component: This component scores each memory slot according to a given question and finds the most relevant memory slots. The dataset consists of 20 different types of QA tasks and Each requires a different number of supporting facts to answer them. For instance, if the following story is told, it is enough to remember the third and six sentences in order to answer the question:

- 1 Mary moved to the bathroom.
- 2 Sandra journeyed to the bedroom.

- 3 Mary got the football there.
- 4 John went to the kitchen.
- 5 Mary went back to the kitchen.
- 6 Mary went back to the garden.
- 7 Where is the football? garden 3 6

For the above QA problem, each line is added to the memory and when the seventh line is the input, x = "Where is the football?", this O component finds the first relevant fact from the story using the following calculation:

$$O(x, m) = \underset{i=1,...,N}{\operatorname{argmax}} s_o(x, m_i)$$

where so is a function that scores the match between x and mi. After the first supporting fact is found as o1, the second supporting fact is found using the following calculation:

$$s(x, y) = \phi_x(x)^T U^T U \phi_y(y)$$

In this way, the O component finds both the first and the second supporting facts and the final output of this component is [x, mo1, mo2], and this is the input for the R component.

R component: This component scores each vocabulary in the dictionary according to the input which contains the question feature representation and the supporting facts, and finds the most likely answer to the question. The following formula ranks the words:

In this way, the R component returns the answer from the vocabulary by ranking them. Both so and sr functions use a fundamental scoring function called s which scores x according

to y:

 $s(x, y) = \phi x(x)TU$

 $TU \varphi y(y)$

where U is a n × D matrix where D is the number of features and n is the embedding dimension. Both ϕx and ϕy are responsible for mapping the feature representation of the text into a D-dimensional feature space where D = (number of supporting facts + 1)|W| where W is the length of the dictionary. O and R components use different weight matrices and are trained separately.

The above example explanation about the O and R components is for just 2 supporting facts. If the necessary number of supporting facts changes then D changes.

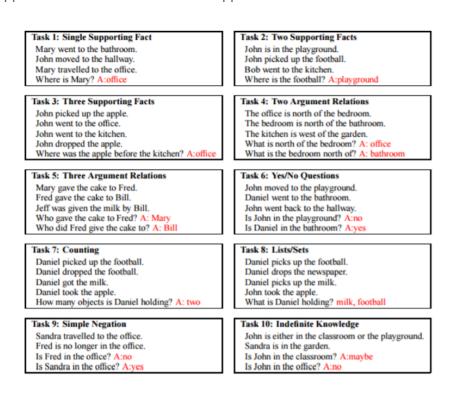
Training

Memory networks are trained using fully supervised settings. During the training, the O component finds the supporting facts, and its loss and accuracy are calculated; then the correct supporting facts with the question are given to the R component as its parameters in order to calculate its loss and accuracy accurately. O component weight and R component weight are trained separately using softmax. In this way, they become experts on their tasks and the risk of misleading the R component due to the incorrect supporting fact choice of the O component is eliminated. However; during the test time, the correct supporting facts and responses are used only for accuracy calculation. The output of the O component is directly given to the R component. Knet.Adam is used in order to update the weights of O and R components.

The model is tested with 20 different tasks which each tests the different aspects of understanding of the story, question, and the ability to relate the question to the sentences in the story.

The 20 different toy tasks can be summarized as follows:

- 1. Single Supporting Fact: In order to answer a question, 1 supporting fact along with the question is enough. These questions generally ask the location of a person in a given story.
- 2. Two or Three Supporting Facts: These tasks are more difficult in comparison with the single supporting fact case. This time two or three supporting facts are necessary in order to answer the question. Two supporting fact questions are usually about the location of an object in the story and three supporting fact questions ask the location of an object before or after something happened such as "Where was the apple before the kitchen?"



Example stories for tasks 1 to 10.

3. Two or Three Argument Relations: In the case of the two-argument relations, the order of the words in a sentence is important in order to answer the question, because two sentences with distinct meanings may have the same words. In the three-argument relations, there are two

people and one object and the goal is to understand which object is involved, who is the giver and the receiver.

- 4. Yes/No Questions: This is the simplest task among others. The questions are true/false questions and answers are either "yes" or "no".
- 5. Counting and List/Set: Task 7 tests the model in terms of the ability to count objects. Questions about counting tasks are usually "How many objects is Daniel holding? ". Task 8 is an advanced version of the questions in Task 7, like "What is Daniel holding? ".
- 6. Simple Negation and Indefinite Knowledge: Task 9 tests one of the simplest forms of negation, some supporting facts imply a statement is false. Task 10 asks questions which are about uncertainty such as giving a fact that "John is either in the classroom or the playground." and asked "Is John in the classroom?", then the correct answer is neither yes nor no, the correct answer is "maybe".
- 7. Basic Coreference, Conjunctions, and Compound Coreference: Task 11 tests the model whether it can find the nearest referent or not. Task 12 asks questions about the conjunction of subjects in a sentence. Task 13 is a combination of Task 11 and task 12, this time one sentence has a referent and the referent has a reference to at least two subjects in another sentence.
- 8. Time Reasoning: Task 14 tests understanding the use of time expressions within the statements and answering questions according to their time order.
- 9. Basic Deduction and Induction: These tasks test the model's ability to understand the basic inheritance of properties between sentences.

Task 11: Basic Coreference Task 12: Conjunction Daniel was in the kitchen. Then he went to the studio. Mary and Jeff went to the kitchen. Then Jeff went to the park. Where is Mary? A: kitchen Where is Jeff? A: park Sandra was in the office. Where is Daniel? A:stu Task 13: Compound Coreference Daniel and Sandra journeyed to the office. Task 14: Time Reasoning In the afternoon Julie went to the park. Then they went to the garden. Yesterday Julie was at school Sandra and John travelled to the kitchen. After that they moved to the hallway. Where is Daniel? A: garden Julie went to the cinema this evening Where did Julie go after the park? A Where was Julie before the park? A Task 16: Basic Induction Task 15: Basic Deduction Sheep are afraid of wolves. Cats are afraid of dogs. Mice are afraid of cats. Bernhard is green. Gertrude is a sheep. Greg is a swan. What is Gertrude afraid of? A:v What color is Greg? A:white Task 17: Positional Reasoning Task 18: Size Reasoning The triangle is to the right of the blue square. The red square is on top of the blue square. The football fits in the suitcase. The suitcase fits in the cupboard. The red sphere is to the right of the blue square. Is the red sphere to the right of the blue square? A:yes Is the red square to the left of the triangle? A:yes The box is smaller than the football. Will the box fit in the suitcase? A:yes Will the cupboard fit in the box? A:no Task 19: Path Finding Task 20: Agent's Motivation John is hungry. John goes to the kitchen. John grabbed the apple there. Daniel is hungry. Where does Daniel go? A:kitchen Why did John go to the kitchen? A:hu The kitchen is north of the hallway. The bathroom is west of the bedroom. The den is east of the hallway. The office is south of the bedroom. How do you go from den to kitchen? A: west, north How do you go from office to bathroom? A: north,

Example stories for tasks 11 to 20.

- 10. Positional and Size Reasoning: Task 17 tests whether the model is able to find relative positions of colored blocks. Task 18 requires a general understanding of the relative size of different objects.
- 11. Path Finding: The problem in Task 19 is about going somewhere to somewhere. It asks how do you get from one to another.
- 12. Agent's Motivation: Task 20 questions test the model understanding of the relation between certain actions and states. For example, the model learns that thirsty people tend to go to the kitchen.

Result

Task	Vocabulary Size	Task	Vocabulary Size
1	19	11	26
2	33	12	20
3	34	13	26
4	15	14	27
5	39	15	25
6	35	16	17
7	43	17	20
8	45	18	20
9	24	19	32
10	25	20	39

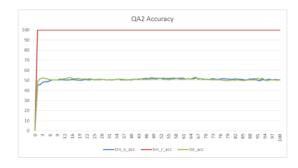
Table 1: Vocabulary size in each task

Task 1: Single Supporting Fact Mary went to the bathroom. John moved to the hallway. Mary travelled to the office. Where is Mary? A:office

100

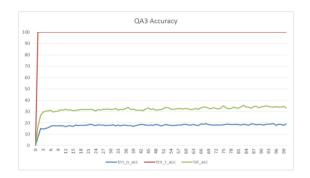
QA1 Accuracy

Task 2: Two Supporting Facts John is in the playground. John picked up the football. Bob went to the kitchen. Where is the football? A:playground



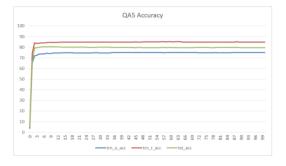
Task 3: Three Supporting Facts

John picked up the apple.
John went to the office.
John went to the kitchen.
John dropped the apple.
Where was the apple before the kitchen? A:office



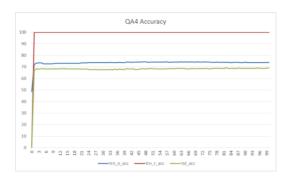
Task 5: Three Argument Relations

Mary gave the cake to Fred.
Fred gave the cake to Bill.
Jeff was given the milk by Bill.
Who gave the cake to Fred? A: Mary
Who did Fred give the cake to? A: Bill



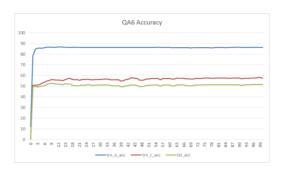
Task 4: Two Argument Relations

The office is north of the bedroom.
The bedroom is north of the bathroom.
The kitchen is west of the garden.
What is north of the bedroom? A: office
What is the bedroom north of? A: bathroom



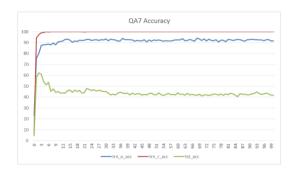
Task 6: Yes/No Questions

John moved to the playground.
Daniel went to the bathroom.
John went back to the hallway.
Is John in the playground? A:no
Is Daniel in the bathroom? A:yes



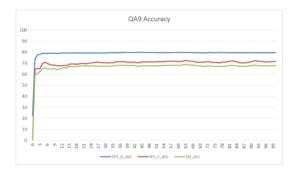
Task 7: Counting

Daniel picked up the football.
Daniel dropped the football.
Daniel got the milk.
Daniel took the apple.
How many objects is Daniel holding? A: two



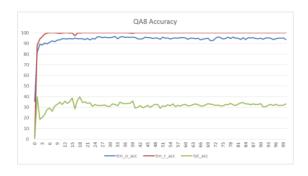
Task 9: Simple Negation

Sandra travelled to the office. Fred is no longer in the office. Is Fred in the office? A:no Is Sandra in the office? A:yes



Task 8: Lists/Sets

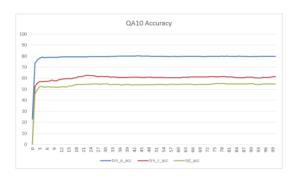
Daniel picks up the football.
Daniel drops the newspaper.
Daniel picks up the milk.
John took the apple.
What is Daniel holding? milk, football



Task 10: Indefinite Knowledge

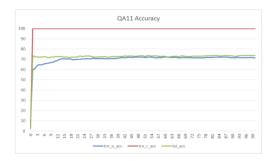
John is either in the classroom or the playground. Sandra is in the garden. Is John in the classroom? A:maybe

Is John in the classroom? A: Is John in the office? A:no



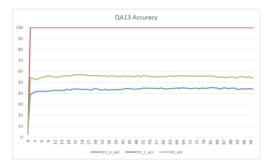
Task 11: Basic Coreference

Daniel was in the kitchen. Then he went to the studio. Sandra was in the office. Where is Daniel? A:studio



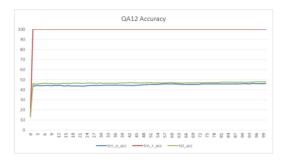
Task 13: Compound Coreference

Daniel and Sandra journeyed to the office. Then they went to the garden. Sandra and John travelled to the kitchen. After that they moved to the hallway. Where is Daniel? A: garden



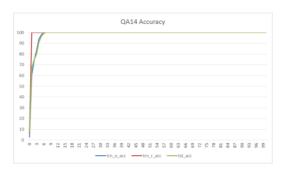
Task 12: Conjunction

Mary and Jeff went to the kitchen. Then Jeff went to the park. Where is Mary? A: kitchen Where is Jeff? A: park



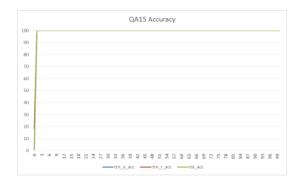
Task 14: Time Reasoning

In the afternoon Julie went to the park. Yesterday Julie was at school. Julie went to the cinema this evening. Where did Julie go after the park? A:cinema Where was Julie before the park? A:school



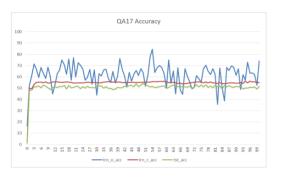
Task 15: Basic Deduction

Sheep are afraid of wolves. Cats are afraid of dogs. Mice are afraid of cats. Gertrude is a sheep. What is Gertrude afraid of? A:wolves



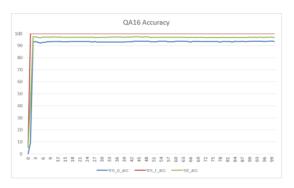
Task 17: Positional Reasoning

The triangle is to the right of the blue square.
The red square is on top of the blue square.
The red sphere is to the right of the blue square.
Is the red sphere to the right of the blue square? A:yes
Is the red square to the left of the triangle? A:yes



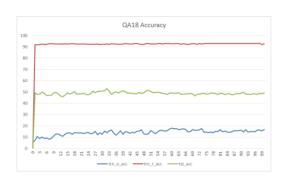
Task 16: Basic Induction

Lily is a swan.
Lily is white.
Bernhard is green.
Greg is a swan.
What color is Greg? A:white



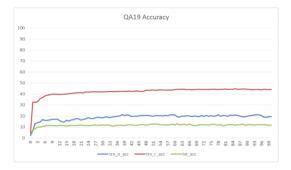
Task 18: Size Reasoning

The football fits in the suitcase.
The suitcase fits in the cupboard.
The box is smaller than the football.
Will the box fit in the suitcase? A yes
Will the cupboard fit in the box? A:no

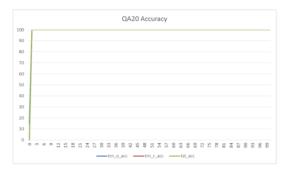


Task 19: Path Finding

The kitchen is north of the hallway.
The bathroom is west of the bedroom.
The den is east of the hallway.
The office is south of the bedroom.
How do you go from den to kitchen? A: west, north
How do you go from office to bathroom? A: north, west



Task 20: Agent's Motivations
John is hungry.
John goes to the kitchen.
John grabbed the apple there.
Daniel is hungry.
Where does Daniel go? A:kitchen
Why did John go to the kitchen? A:hungry



Achieved better accuracy on 7 tasks and similar accuracy on 7 tasks, however; got poor accuracy on 6 tasks. The following can cause getting poor accuracy in these tasks:

- 1. Normally each task other than task 7 and task 8 requires a constant number of supporting facts to answer questions correctly. In tasks 7 and 8, the number of supporting facts needed changes for different questions, but my model assumes that each question requires the same number of supporting facts at the beginning of training, so it uses D = maximum number of supporting facts needed and updates it during the training. This also causes overfitting after a few epochs.
- 2. The inner feature representation of sentences could be another reason. I could not find the code of authors' implementation on the Internet, thus I implemented my memory network by understanding the paper and watching video lectures online about memory networks.
- 3. The paper suggests an improvement for memory networks which is called the time feature. A memory network with a time feature knows which sentence comes first and scores memories according to this fact while finding supporting facts, therefore relative time is used. Adding a time feature requires using hinge loss as the loss function, however; hinge loss is not suitable in

my implementation since it is not differentiable. Thus, I used absolute time with soft loss instead of relative time with hinge loss.

4. In the original paper, margin ranking loss and SGD is used for training, but I used soft loss and Knet.Adam for training.

The implementation has achieved higher accuracy on the following 7 tasks:

- Task 3: Three Supporting Facts 35.5 %
- Task 6: Yes/No Questions 53 %
- Task 9: Simple Negation 68 %
- Task 14: Time Reasoning 100 %
- Task 15: Basic Deduction 100 %
- Task 16: Basic Induction 97.4 %
- Task 19: Path Finding 12.9 %

Higher accuracies are obtained on especially Task 3, Task 15, Task 16, and Task 19 in the implementation. The implementation is not good enough to answer all questions on Task 3 as well, but it has achieved higher accuracy. The paper's implementation is not good enough to answer all of the questions on Task 15 and Task 16, and obtains poor accuracy on these tasks; on the other hand, implementation of memory networks is better on these tasks and answers approximately all of the questions correctly. If Task 19 is considered, the paper's implementation obtained 0 % accuracy on this task and the agent never finds the correct path. My implementation also does not always give the correct path to the agent, but at least 12.9 % of the time the agent reaches wherever it wants. Possible reasons why my implementation has achieved higher accuracies on these tasks:

- 1. used Knet.Adam as the parameter optimizer.
- 2. used soft loss instead of margin ranking loss.

3. internal representation might be more appropriate for these tasks

Related Work

Question Answering is one of the fundamental problems in machine learning. There have been various approaches in the past which include using RNN, LSTM, SVM, etc. Memory Networks work of Weston et al. introduced a new approach called MemNN which is the abbreviation of Memory Neural Networks. They tested their approach with large-scale and toy QA tasks and proved MemNN is able to solve these problems. I used their approach to solve their QA problems with absolute time and tested my model with the dataset which is used for Towards Al-Complete Question Answering: A Set of Prerequisite Toy Tasks. The dataset is more generalizable, comprehensive, and splits the tasks into categories.

Conclusion

In conclusion, recently introduced new classes of learning models memory networks are re-implemented and tested. As stated in the Analysis and Contributions section, the model has achieved similar and better accuracies on most of the tasks (14 out of 20 tasks) In the other tasks, obtained poor accuracies according to the stated accuracies in Towards Al-Complete Question Answering: A Set of Prerequisite Toy Tasks. The possible reasons for these results are explained in the Analysis and Contributions section. In summary, the internal representation, absolute time vs. relative time, and different numbers of supporting facts cause the accuracy difference. The results in the Towards Al-Complete Question Answering: A Set of Prerequisite Toy Tasks is for another implementation of memory networks that uses relative time. Overall the model achieved nearly the same accuracy in terms of mean performance and fulfilled the task of reimplementation of MemNN.

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