



Faculté
des Sciences

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MacNetwork

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Outline

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 - *The task at hand*
- 2 The Datasets
- 3 End to End Memory Network
 - *How to solve this?*
- 4 MAC Cell
- 5 Some Results
 - *What we did*
 - *CLEVR*
 - *Babi*

Reasoning

What is relational reasoning?

- Finding who is the murder on an Agatha Christie novel
- Figure out spacial relation from textual clues
- Understanding which number comes next on a sequence of numbers

Reasoning

What is relational reasoning?

- Manipulating previous knowledge in order to answer a new question
- Composition rules to guide and address new questions

A collection of 3D geometric shapes including cylinders, spheres, and cubes in various colors and materials (matte, metallic, reflective). The shapes are arranged on a light gray surface. The objects include: a small brown cylinder, a small red sphere, a small gold sphere, a large red cylinder, a large red sphere, a large gold cube, a small gold sphere, a small green cylinder, a small blue cube, and a small cyan cube.

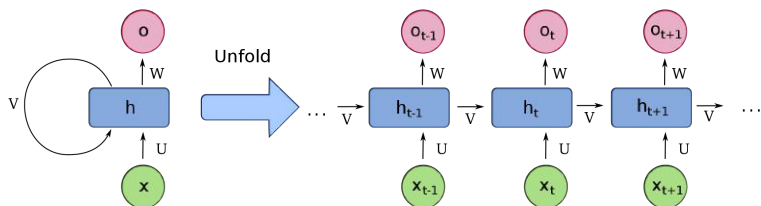
Q: How many objects are either small cylinders or red things?

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 office? A: no

23-01-2019

How to solve this?

Let's start with RNN



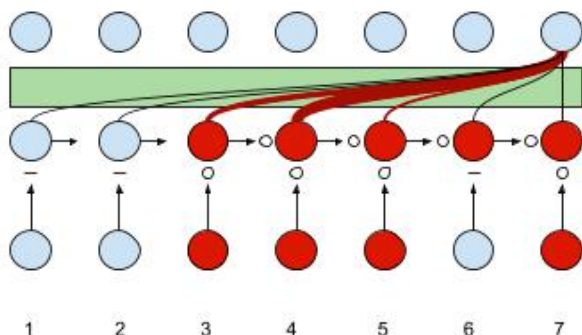
- Recurrent Neural Networks are good for temporal data (like audio, text, etc)
- Each node receives the output of the previous node alongside with the current data

Problems

- What about long term dependencies?
- And what if we have out of order sequences?

How to solve this?

Attention Model (Bahdanau et al. 2014)



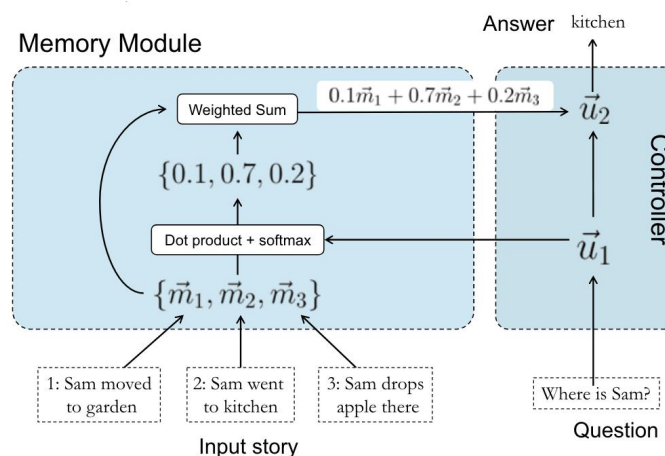
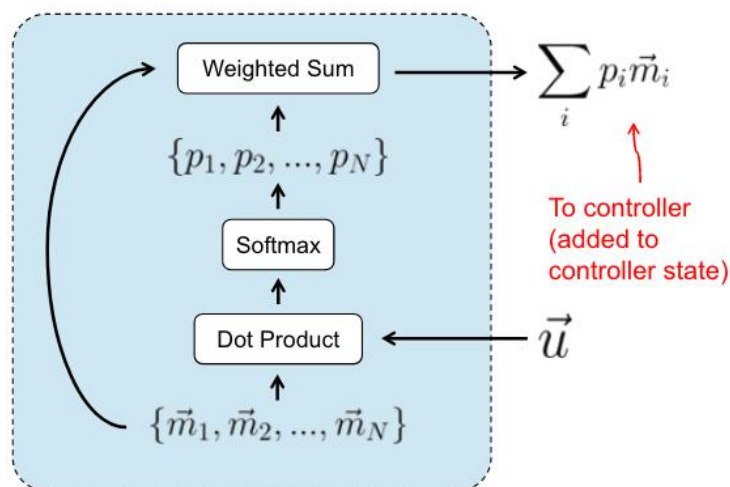
- We will have a new layer that tries to grasp the dependencies (attention) between the output and the previous states
- Started as a NLP solution but soon *invaded* other areas

What does this solve?

- It allows the network to use the previous hidden states and the representation of the input to figure out what it should pay attention to
- Can help with long term dependencies

End to End Memory Network

Sukhbaatar et al. 2015



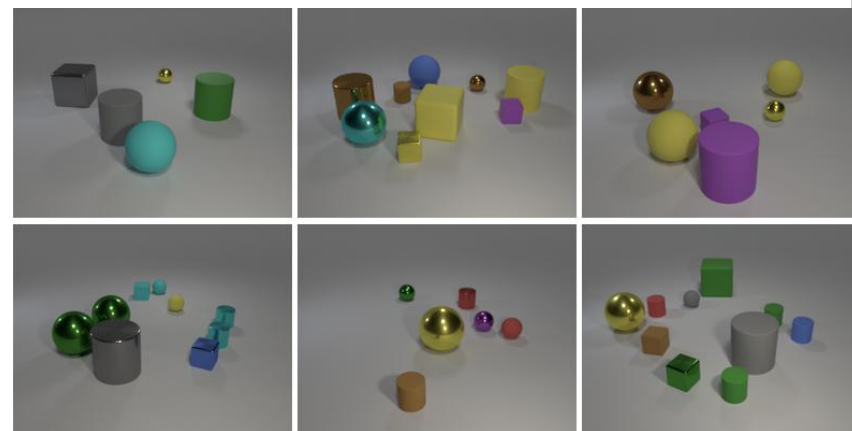
Main concepts

- Two different Modules: Memory and Control
- Memory controls what will be fed into our Controller
- Controller will help us guide the query until we reach the answer

Intuitions and Task at Hand

VQA and CLEVR

- Querying attributes
 - Two different sizes
 - Two materials
 - Three Shapes
 - Eight Colors
- Comparing
- Existence
- Counting
- Integer comparison



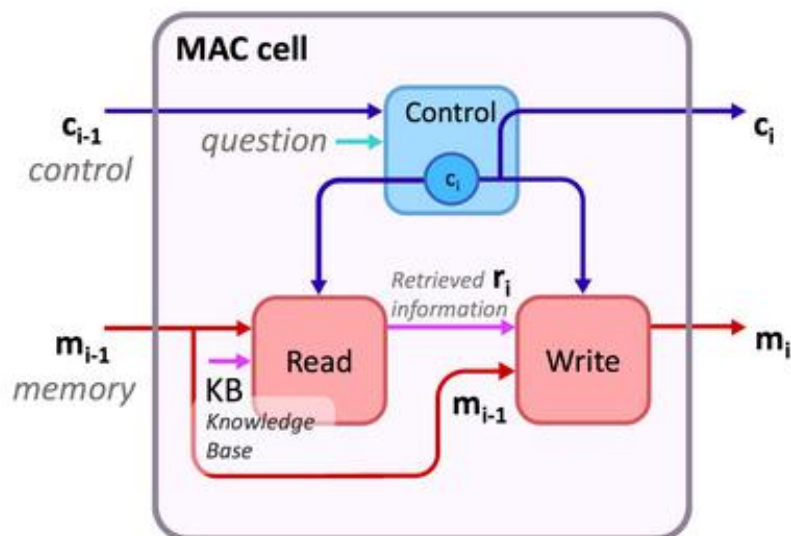
Intuitions and Task at Hand

Why was CLEVR created?

- The idea behind CLEVR is to have a Visual Questioning dataset with reduced bias and superficial reasoning
- By doing so it allows a deeper analysis of the performance and the system itself
- Artificial Dataset: images and questions are generated

Mac Cell

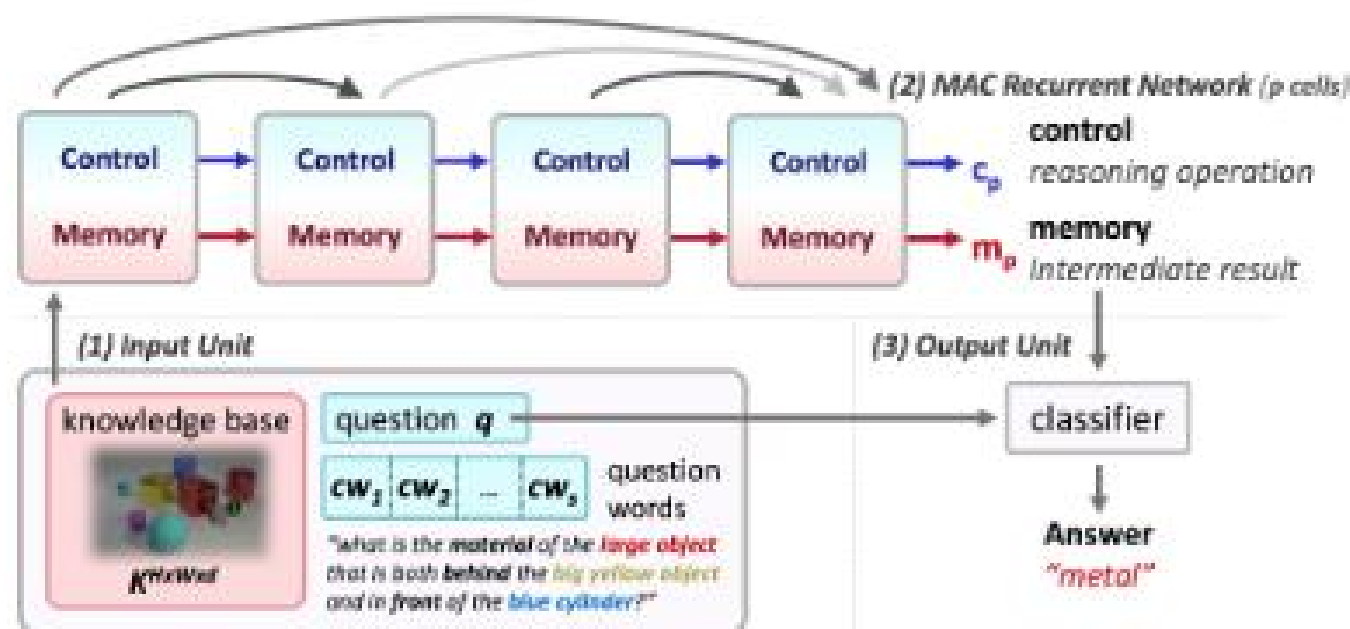
Mac Cell (Hudson et al, 2018)



- Three modules: Control, Read and Write
- Input
 - String of length S that will be processed by a d -dimensional *biLSTM* which will give the question representation $q = [\overleftarrow{cw_1}, \overrightarrow{cw_S}]$
 - Image is processed by a feature extractor trained on ResNet101 (outputs conv4) and passed through two CNN Layers with d output channels yielding a $KB_v = [H, W, d]$
- It has two recurrent states: Control and Memory (sounds familiar?)

Mac Cell

Example of Use

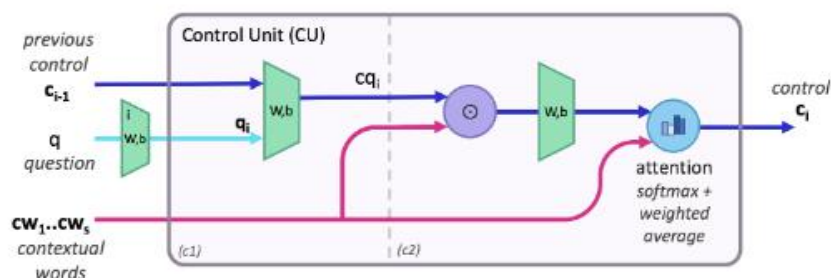


How does it work

- The Mac cell is constructed to work in a sequential manner, i.e., several cells connected together to add extra steps of reasoning
- We actually have 28 possible answers (from colors to numbers, etc), for the classification we use a softmax over the last memory and the question itself

Control Unit

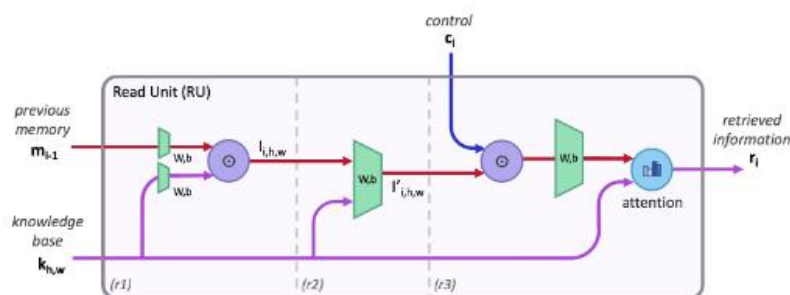
The Structure



- Computes a control state, extracting an instruction that focuses on some aspect of the query
- Input
 - Previous control state
 - Query
 - Contextual Words (from *BiLSTM*)
- First it calculates a time specific query representation
- Then it combines it with the preceding control
- Finally we project it down to the actual words of the query with an attention mechanism

Read Unit

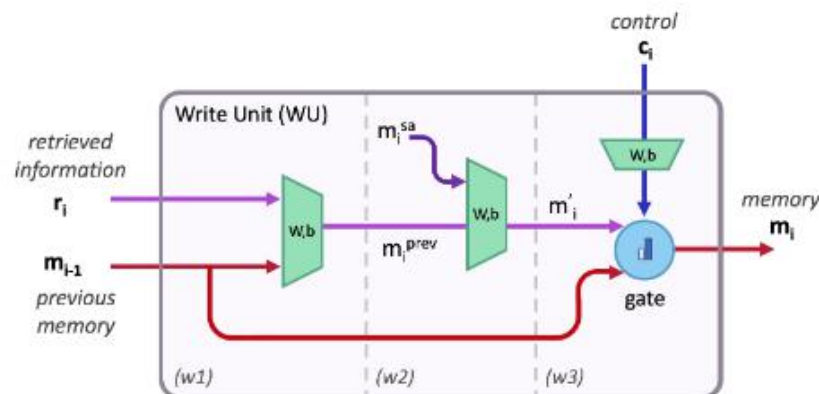
The Structure



- Retrieves information from the Knowledge base given the current control state and previous memory state
- Input
 - Previous memory
 - Knowledge Base
 - Current Control
- Will consider each item from the KB (the features of the image)
- We will then consider the interaction between the previous memory state (through a Hadamard product)
- The last step allows us do transitive reasoning (previous memory will guide you to that)
- After we consider the information in light of the current control
- We finalize by doing attention

Write Unit


The structure



- Update the new memory state, merging old and new information
- Input
 - Retrieved Information (from the Read Unit)
 - Previous Memory
 - Current Control
- We combine the previous memory with the new memory
- A self-attention mechanism (which gives access to all the previous memories) to help it guide the construction of the new memory
- We will have a gate that will allow us to decide if we want to update the memory or not

Our Attempt

What we did

- We first finished the Multi-GPU section of the original code (missing the gradient calculation between GPUs)
- We then tried to implement the MAC network on Keras (original works in tensorflow and torch), due to lack of time we didn't finish it
- Our goal was to train on Babi to confirm the hypothesis that MAC isn't only for VQA
- You can find both the improvement and the attempt here:
 [TiagoCortinhal](#)

CLEVR

Implementation used

- Implementation of Hudson et al, 2018
- As presented in the paper

How training was done

- We used all CLEVR dataset
- trained over 25 epochs and 4 reasoning steps
- We achieved results of 97%

CLEVR

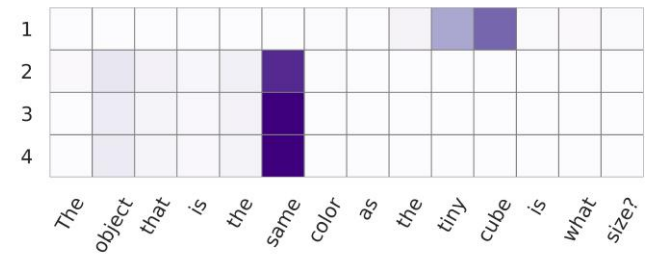
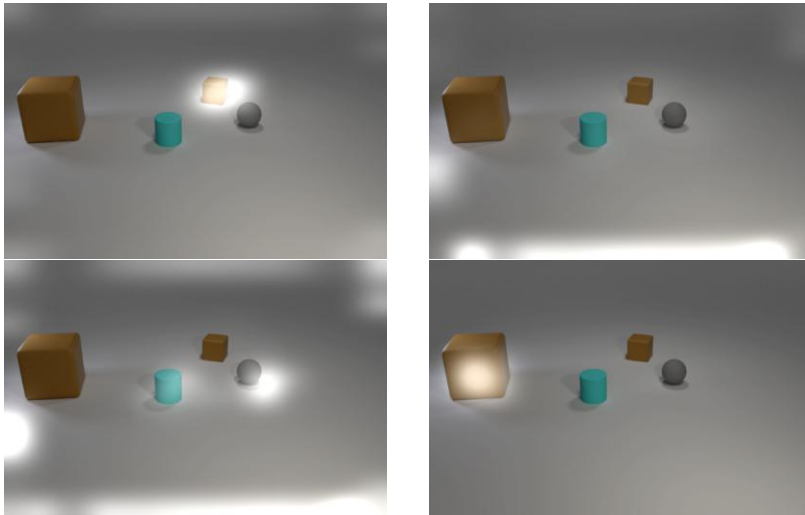


Figure: G:Large P:Large

CLEVR

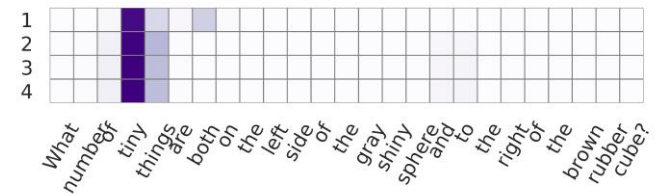
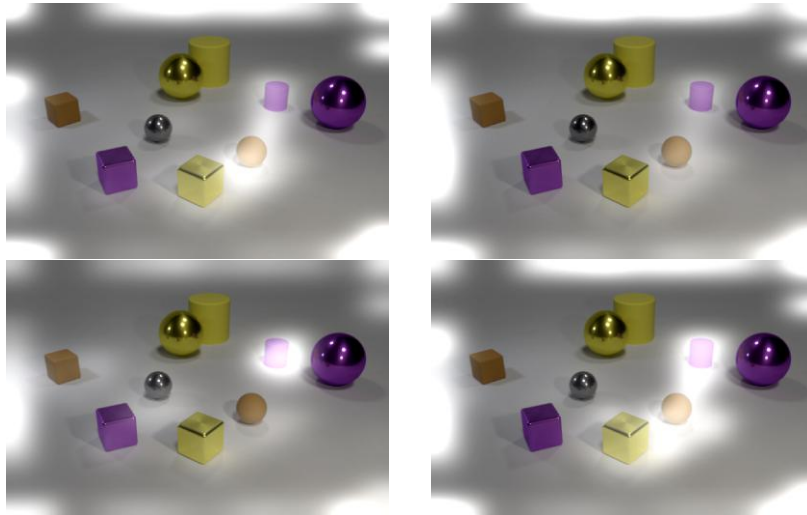


Figure: G:0 P:1

CLEVR

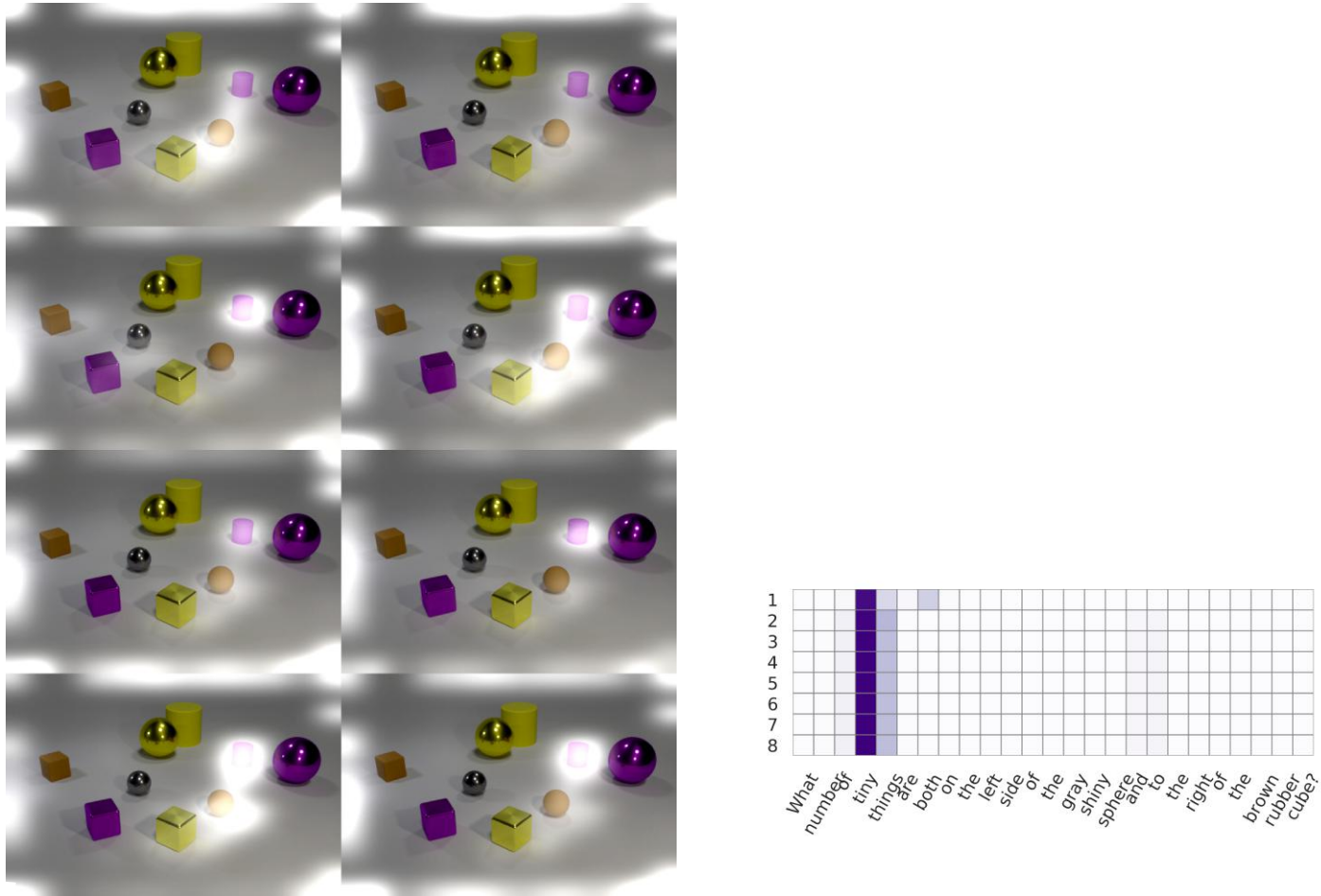


Figure: G:0 P:0

CLEVR

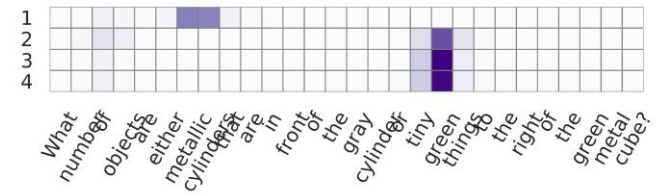
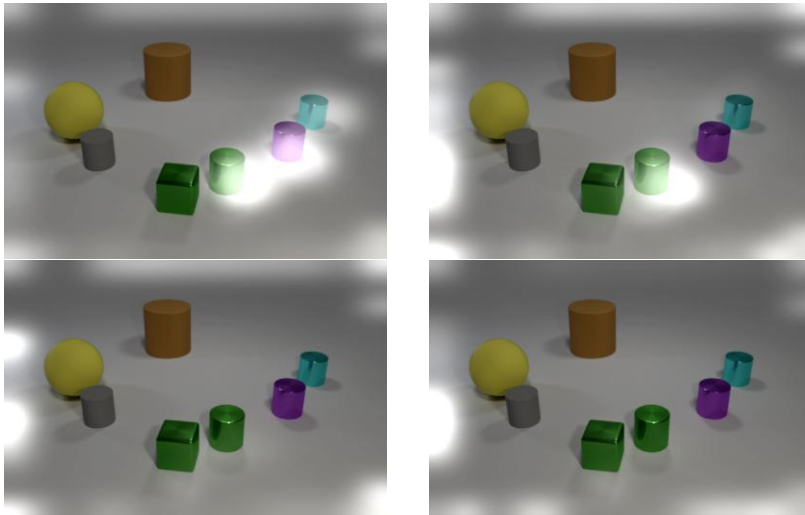
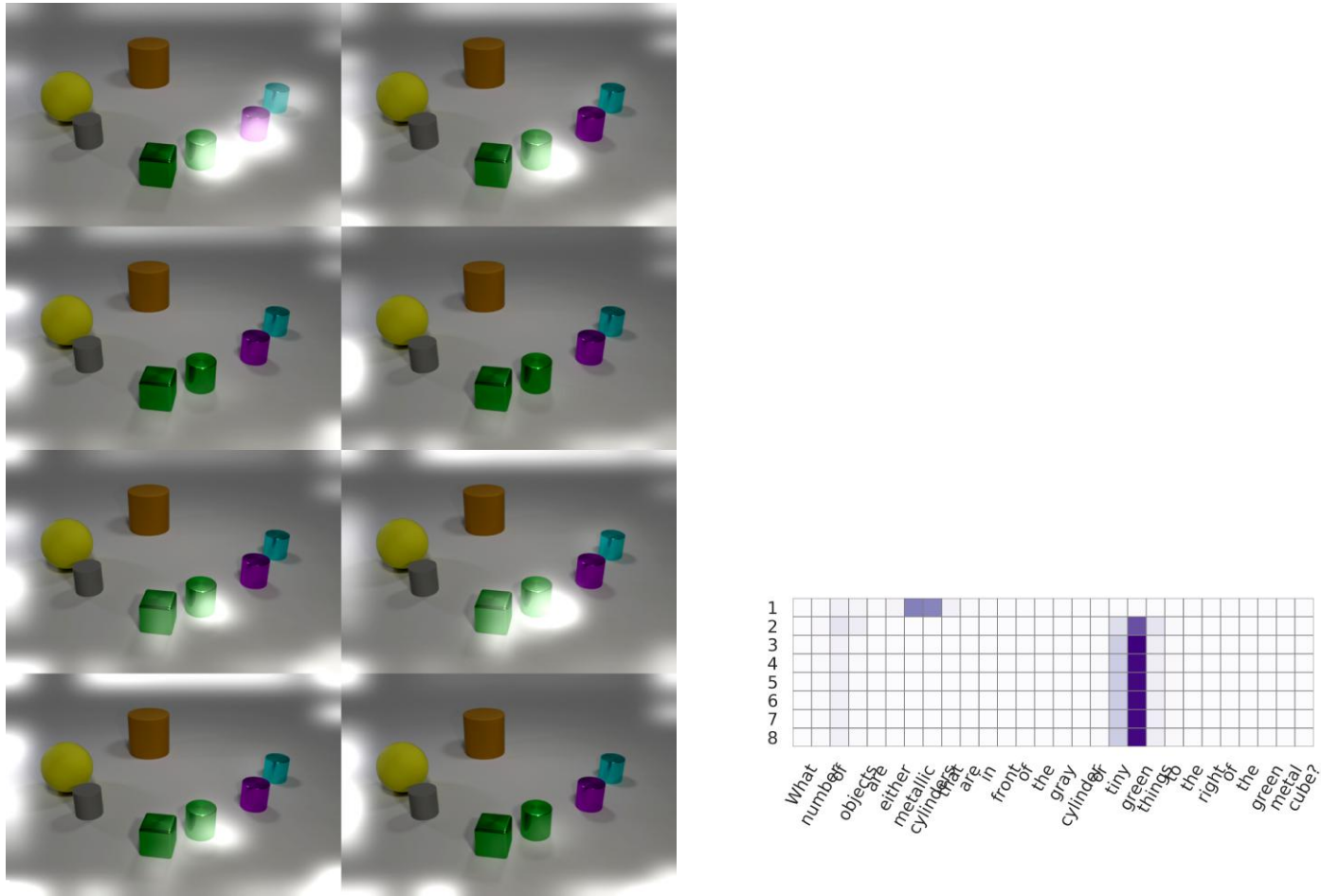


Figure: G:1 P:2

CLEVER



Structure of Babi

Babi

```

1 John travelled to the hallway.
2 Mary journeyed to the bathroom.
3 Where is John?      hallway 1
4 Daniel went back to the bathroom.
5 John moved to the bedroom.
6 Where is Mary?      bathroom 2
7 John went to the hallway.
8 Sandra journeyed to the kitchen.
9 Where is Sandra?    kitchen 8
10 Sandra travelled to the hallway.
11 John went to the garden.
12 Where is Sandra?    hallway 10
13 Sandra went back to the bathroom.
14 Sandra moved to the kitchen.
15 Where is Sandra?    kitchen 14

1 Jason is thirsty.
2 Where will jason go? kitchen 1
3 Antoine is bored.
4 Where will antoine go? garden 3
5 Jason went to the kitchen.
6 Why did jason go to the kitchen? thirsty 1
7 Antoine journeyed to the garden.
8 Why did antoine go to the garden? bored 3
9 Sumit is hungry.
10 Where will sumit go? kitchen 9
11 Antoine grabbed the football there.
12 Why did antoine get the football? bored 3
13 Sumit went to the kitchen.
14 Why did sumit go to the kitchen? hungry 9
15 Jason took the milk there.
16 Why did jason get the milk? thirsty 1
17 Sumit took the apple there.
18 Why did sumit get the apple? hungry 9
19 Yann is thirsty.
20 Where will yann go? kitchen 19
21 Yann travelled to the kitchen.
22 Why did yann go to the kitchen? thirsty 19

```

- Was we have seen before, we will have several tasks:
 - Single/Two/Three supporting fact
 - Counting
 - Listing
 - Size reasoning
 - etc
- Story size varies, and we can have stories with several questions

Difference between CLEVR and Babi

- We have 20 different tasks ranging from simple questions to more intricate ones like path-finding or time reasoning

Babi

Implementation used

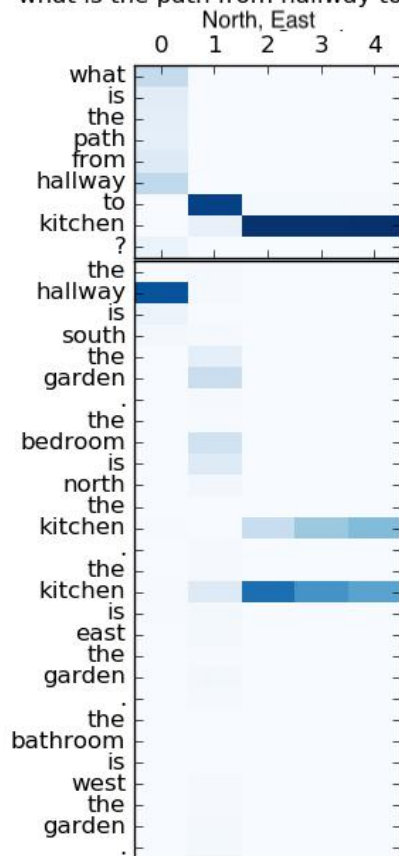
- Implementation of Selvakumar et al, 2018.
- Follows the implementation of the original paper

How training was done

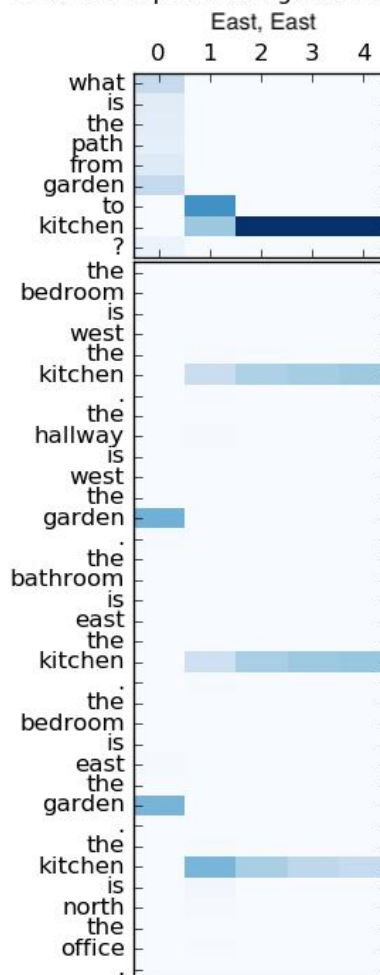
- We had actually to regenerate the pathfinder task due to low accuracy (from 10k examples to 50k)
- trained over 10 epochs and 5 reasoning steps
- we obtained similar scores to the article (95%)

Babi Results

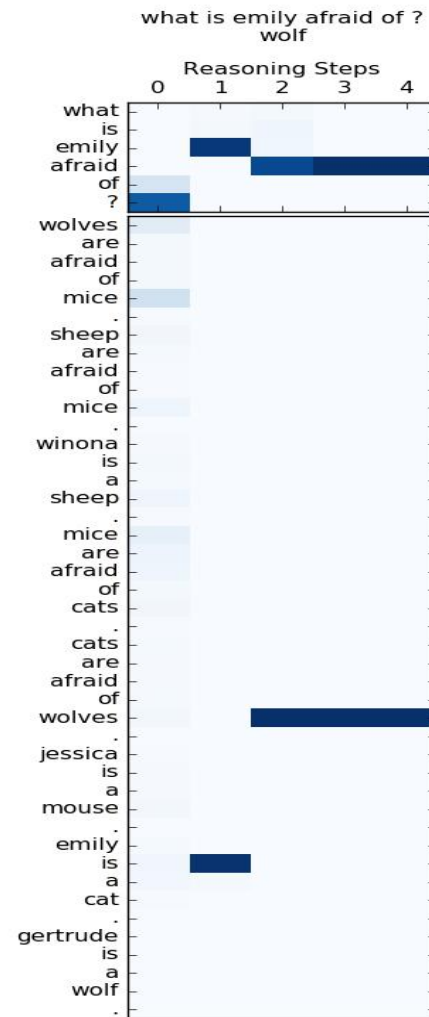
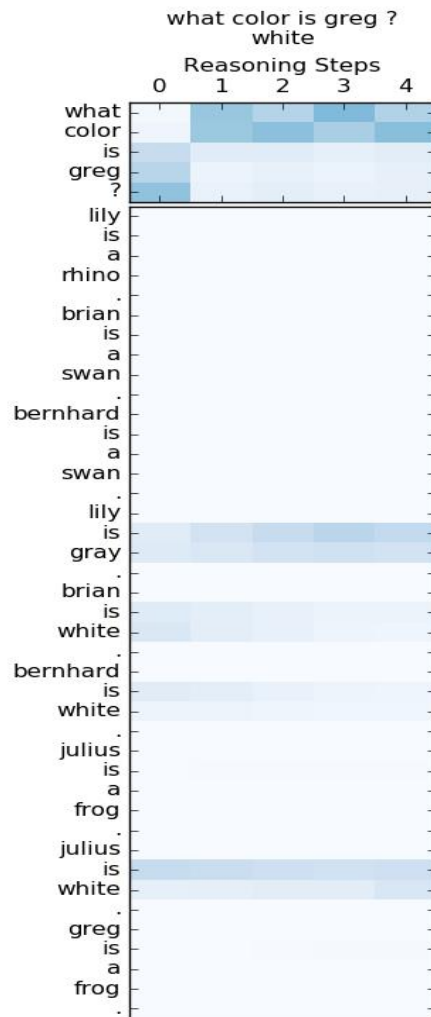
what is the path from hallway to kitchen ?



what is the path from garden to kitchen ?



Babi Results





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The End!!