



Automated Theorem Proving

-- A Deep Learning Approach

Industrial Sponsor: Real AI

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Outline

- Introduction
- Metamath: The Foundation
- Holophrasm: An Overview
- Future Work



Outline

- **Introduction**
- Metamath: The Foundation
- Holophrasm: An Overview
- Future Work



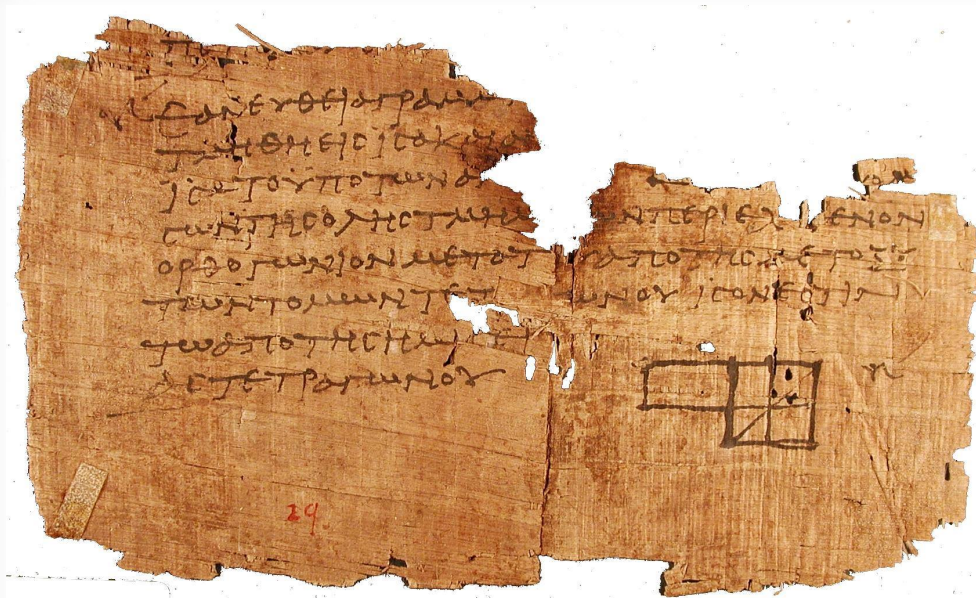
Company Background



- Research & development of safe and beneficial artificial general intelligence
- Altruistic intelligence for problem solving



Proving Math Theorems ...



PROPOSITION XIX. THEOREM

106. *If two parallel lines are cut by a transversal, the corresponding angles are equal.*
 [Converse of Prop. XIV.]

Given parallel lines AB and CD and the cor. $\angle 1$ and 2 .
 To prove $\angle 1 = \angle 2$.

STATEMENTS	Reasons
$\angle 1 = \angle 3$.	Vertical \angle are equal.
$\angle 2 = \angle 3$.	Alt. int. \angle of \parallel lines are equal.
$\therefore \angle 1 = \angle 2$.	Things equal to the same thing are equal to each other.

Q. E. D.



Deep Learning

- Language translation
- Image recognition
- Alphago

⋮

- **Mathematics??**



<https://www.theguardian.com/technology/2017/may/23/alphago-google-ai-beats-ke-jie-china-go>



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- Holophrasm
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Formal Systems

- Enables computers to work on proofs
- Metamath, Mizar, HOL Light



Metamath

- Over 19,000 proofs
- Built from 22 axioms
 - Logic + ZFC axioms



Metamath Proof Explorer Home Page

[First >](#)
[Last >](#)

[Mirrors](#) > [Home](#) > [MPE Home](#) > [Th. List](#) > [Recent](#)

The aleph null above is the symbol for the first infinite cardinal number, discovered by Georg Cantor in 1873 (see theorem [aleph0](#)).

This is the starting page for the Metamath Proof Explorer subproject (set.mm database). See the main [Metamath Home Page](#) for an overview of Metamath and download links.

Contents of this page

- [Metamath Proof Explorer Overview](#)
- [How Metamath Proofs Work](#)
- [The Axioms](#) (Propositional Calculus, Predicate Calculus, Set Theory, The Tarski-Grothendieck Axiom)
- [The Theory of Classes](#) New 13-Dec-2015
- [A Theorem Sampler](#)
- [2 + 2 = 4 Trivia](#)
- [Appendix 1: A Note on the Axioms](#)
- [Appendix 2: Traditional Textbook Axioms of Predicate Calculus](#)
- [Appendix 3: Distinct Variables](#) (History, Notes) Revised 21-Dec-2016
- [Appendix 4: A Note on Definitions](#)
- [Appendix 5: How to Find Out What Axioms a Proof Depends On](#)
- [Appendix 6: Notation for Function and Operation Values](#)
- [Appendix 7: Some Predicate Calculus Subsystems](#)
- [Reading Suggestions](#)
- [Bibliography](#)
- [Browsers and Fonts](#)

Related pages

- [Theorem List \(Table of Contents\)](#)
- [Most Recent Proofs \(this mirror\)](#) (latest)
- [Conventions and Style](#) New 15-Jan-2017
- [Bibliographic Cross-Reference](#)
- [Definition List \(3MB\)](#)
- [Deduction Form and Natural Deduction](#) New 7-Feb-2017 (Natural Deduction Rules New 9-Feb-2017)
- [Weak Deduction Theorem](#) (an older method)
- [Real and Complex Numbers](#)
- [ZFC Axioms With No Distinct Variables](#)
- [ASCII Symbol Equivalents for Text-Only Browsers](#)
- [Ghilbert Proof Language](#) [retrieved 21-Dec-2016]

To search this site you can use [Google](#) [retrieved 21-Dec-2016] restricted to a mirror site. For example, to find references to infinity enter "infinity site:us.metamath.org". More efficient searching is possible with direct use of the [Metamath program](#), once you get used to its [ASCII tokens](#). See the wildcard features in "help search" and "help show statement".



Dissecting a proof: $t=t$

- Two axioms:

$$(A1) \quad (t = r \rightarrow (t = s \rightarrow r = s))$$

$$(A2) \quad (t + 0) = t$$

- Modus Ponens:

$$(MP) \quad ((P \wedge (P \rightarrow Q)) \rightarrow Q)$$



Dissecting a proof: $t=t$

- Required 'tools'
- Assertion
- Proof

`th1 $p |- t = t`

`tt tze tpl tt weq tt tt weq tt a2 tt tze tpl
tt weq tt tze tpl tt weq tt tt weq wim tt a2
tt tze tpl tt tt a1 mp mp`

```
$( Declare the constant symbols we will use $)
  $c 0 + = -> ( ) term wff |- $.
$( Declare the metavariables we will use $)
  $v t r s P Q $.
$( Specify properties of the metavariables $)
  tt $f term t $.
  tr $f term r $.
  ts $f term s $.
  wp $f wff P $.
  wq $f wff Q $.
$( Define "term" (part 1) $)
  tze $a term 0 $.
$( Define "term" (part 2) $)
  tpl $a term ( t + r ) $.
$( Define "wff" (part 1) $)
  weq $a wff t = r $.
$( Define "wff" (part 2) $)
  wim $a wff ( P -> Q ) $.
$( State axiom a1 $)
  a1 $a |- ( t = r -> ( t = s -> r = s ) ) $.
$( State axiom a2 $)
  a2 $a |- ( t + 0 ) = t $.
  ${
    min $e |- P $.
    maj $e |- ( P -> Q ) $.
  }
$( Define the modus ponens inference rule $)
  mp $a |- Q $.
  $}
$( Prove a theorem $)
  th1 $p |- t = t $=
  $( Here is its proof: $)
    tt tze tpl tt weq tt tt weq tt a2 tt tze tpl
    tt weq tt tze tpl tt weq tt tt weq wim tt a2
    tt tze tpl tt tt a1 mp mp
  $.
```



Dissecting a proof: $t=t$

- Reverse Polish Notation (RPN)

Usual notation	RPN
$2+3$	$23+$
$4+6-8$	$46+8-$

- Hypotheses: operands
- Theorems: operators



Dissecting a proof: $t=t$

Proof	tt tze tpl tt weq ... a1 mp mp
-------	--------------------------------



Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

tt \$f term t \$.

A hypothesis

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

term t

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt **tze** tpl tt weq ... a1 mp mp

tze \$a term 0 \$.

An axiom with a **CONSTANT**

term t

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

term 0

term t

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze **tpl** tt weq ... a1 mp mp

tpl \$a term (t + r) \$.

An axiom with two VARIABLES

term 0

term t

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze **tpl** tt weq ... a1 mp mp

tpl \$a term (**t** + **r**) \$.

term 0

term t

Stack

Compare

tt \$f term **r** \$.

tt \$f term **t** \$.

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze **tpl** tt weq ... a1 mp mp

tpl \$a term (t + r) \$.

Perform
Substitution

Replace t by t
Replace r by 0

term 0

term t

Stack

term (t + 0)

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

Remove the objects which were
used to perform the substitution

Stack

term (t + 0)

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

term (t + 0)

Stack

Processing Area



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

term t

term (t + 0)

Stack



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

wff ($t + 0$) = t

Stack



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

$| - ((t + 0) = t \rightarrow ((t + 0) = t \rightarrow t = t))$

$| - (t + 0) = t$

$wff ((t + 0) = t \rightarrow t = t)$

$wff (t + 0) = t$

$| - (t + 0) = t$

$wff t = t$

$wff (t + 0) = t$

Stack



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

$\vdash ((t + 0) = t \rightarrow t = t)$

$\vdash (t + 0) = t$

$\text{wff } t = t$

$\text{wff } (t + 0) = t$

Stack



Dissecting a proof: $t=t$

Proof

tt tze tpl tt weq ... a1 mp mp

| - $t = t$

Stack



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Holophrasm

Daniel P.Z. Whalen (2016), Holophrasm: a neural Automated Theorem Prover for higher-order logic. : [arxiv:1608.02644](https://arxiv.org/abs/1608.02644)

Github : <https://github.com/dwhalen/holophrasm>

Automated theorem prover based on metamath, python, and RNN.

Found proofs for 388 of 2720, or 14.3% of the test propositions



Proof Tree

```
$( Declare the constant symbols we will use $)
  $c 0 + = -> ( ) term wff |- $.
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$( Define "wff" (part 1) $)
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$( Define "wff" (part 2) $)
  wim $a wff ( P -> Q ) $.
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$( State axiom a2 $)
  a2 $a |- ( t + 0 ) = t $.
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    maj $e |- ( P -> Q ) $.
  }
$( Define the modus ponens inference rule $)
  mp $a |- Q $.
  $}
$( Prove a theorem $)
  th1 $p |- t = t $=
  $( Here is its proof: $)
    tt tze tpl tt weq tt tt weq tt a2 tt tze tpl
    tt weq tt tze tpl tt weq tt tt weq wim tt a2
    tt tze tpl tt tt a1 mp mp
  $.
```



Proof Tree

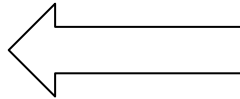
```
$( State axiom a1 $)
  a1 $a |- ( t = r -> ( t = s -> r = s ) ) $.
$( State axiom a2 $)
  a2 $a |- ( t + 0 ) = t $.
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    tt weq tt tze tpl tt weq tt tt weq wim tt a2
    tt tze tpl tt tt a1 mp mp
  $.
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$



```
$( State axiom a1 $)  
a1 $a |- ( t = r -> ( t = s -> r = s ) ) $.
```



Proof Tree

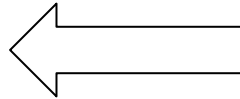
Axiom a1

Assertion : $\vdash t=r \Rightarrow$

$(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$



```
$( State axiom a2 $)
a2 $a |- ( t + 0 ) = t $.
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

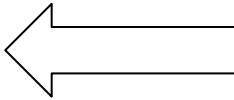
Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$



```
{
  min $e |- P $.
  maj $e |- ( P -> Q ) $.
$( Define the modus ponens inference rule $)
  mp $a |- Q $.
$}
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash t = t$



$\$(\text{ Prove a theorem } \$)$
 $\text{th1 } \$p \mid - t = t \$=$



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash t = t$



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$

$\alpha = \alpha$



Proof Tree

$$\alpha = \alpha$$

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

Hypotheses : $\vdash P$

$\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$

```
$( Prove a theorem $)
  th1 $p |- t = t $=
    $( Here is its proof: $)
      tt tze tpl tt weq tt tt weq tt a2 tt tze tpl
      tt weq tt tze tpl tt weq tt tt weq wim tt a2
      tt tze tpl tt tt a1 mp mp
    $.
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

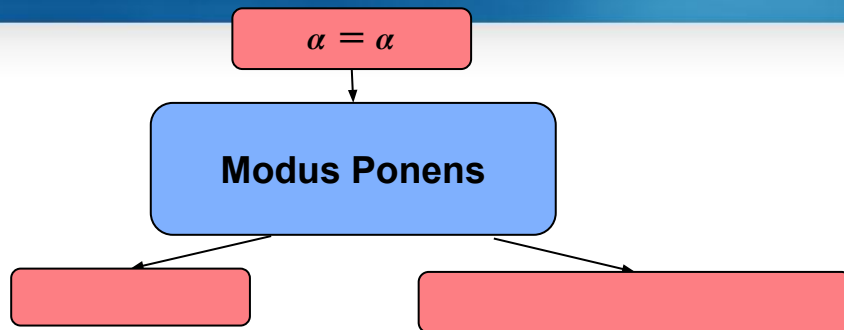
Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$





Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

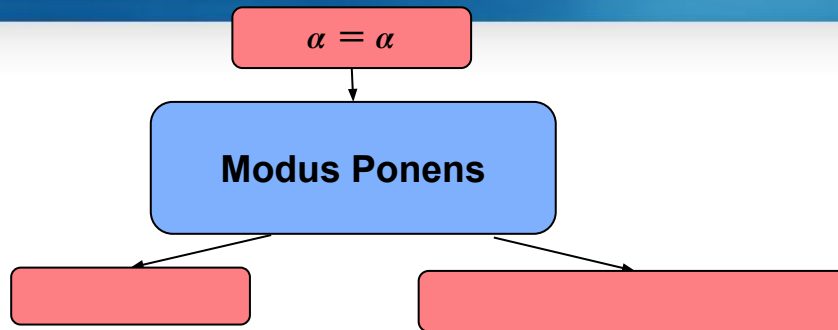
Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$



```
$( Here is its proof: $)  
tt tze tpl tt weq tt tt weq tt a2 tt tze tpl  
tt weq tt tze tpl tt weq tt tt weq wim tt a2  
tt tze tpl tt tt a1 mp mp
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
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Axiom a2

Assertion : $\vdash (t+0) = t$

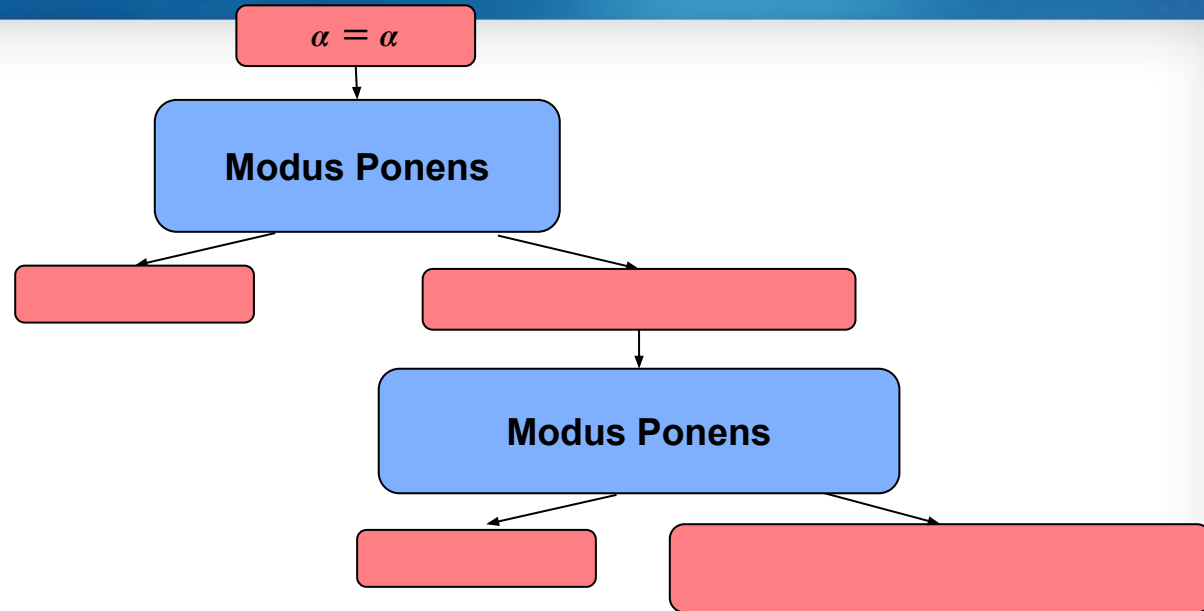
Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$





Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$

$(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

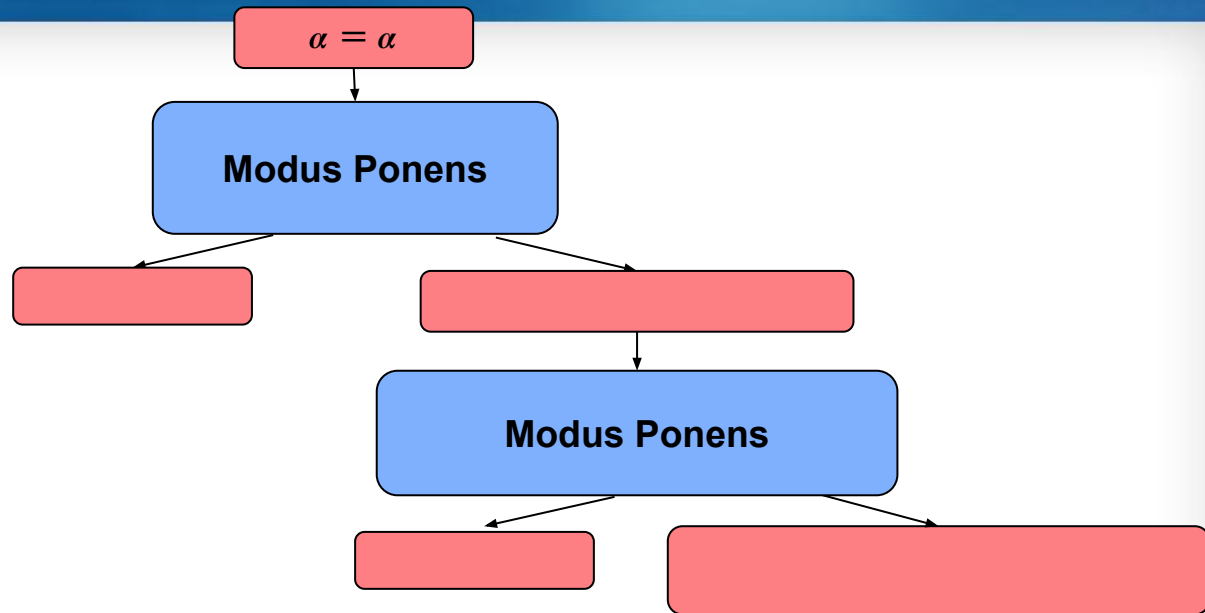
Hypotheses : $\vdash P$

$\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$



\$(Here is its proof: \$)

```

tt tze tpl tt weq tt tt weq tt a2 tt tze tpl
tt weq tt tze tpl tt weq tt tt weq wim tt a2
tt tze tpl tt tt a1 mp mp
  
```



Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$

$(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

Modus Ponens(mp)

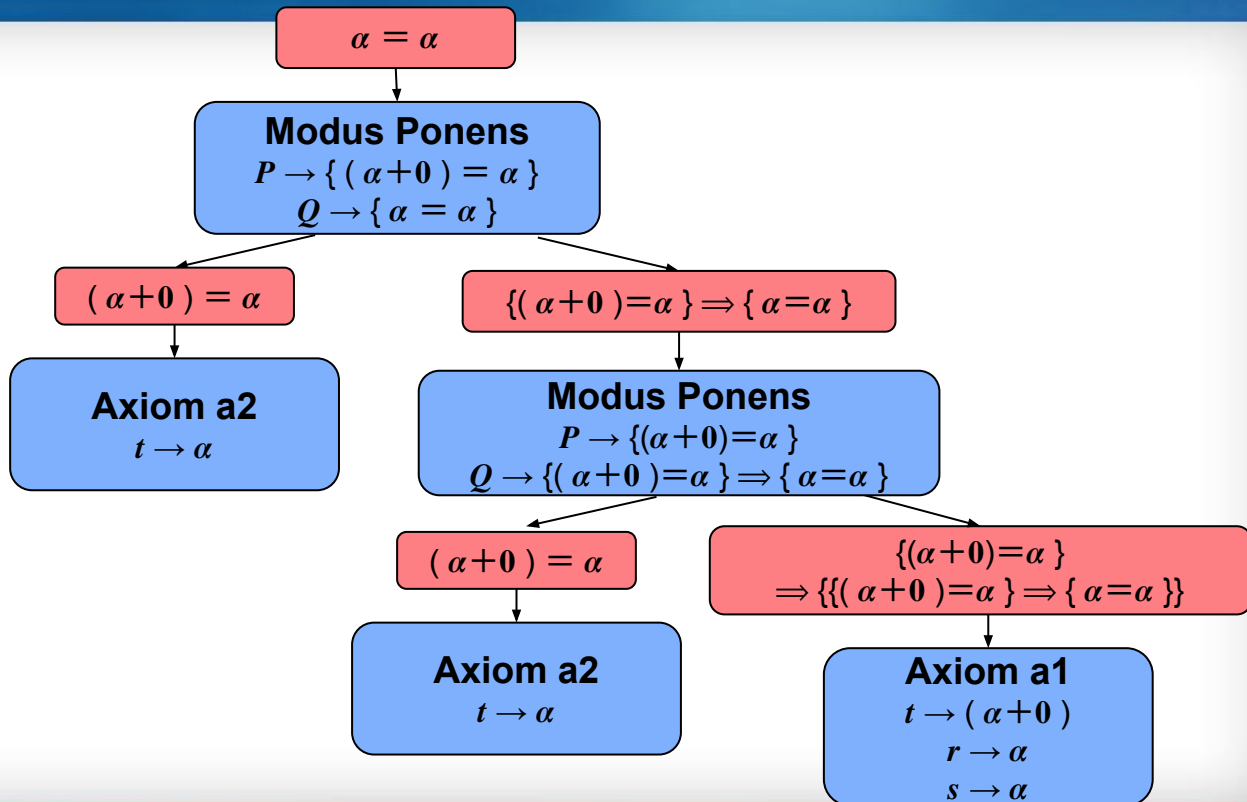
Hypotheses : $\vdash P$

$\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$





Proof Tree

Axiom a1

Assertion : $\vdash t=r \Rightarrow$
 $(t=s \Rightarrow r=s)$

Axiom a2

Assertion : $\vdash (t+0) = t$

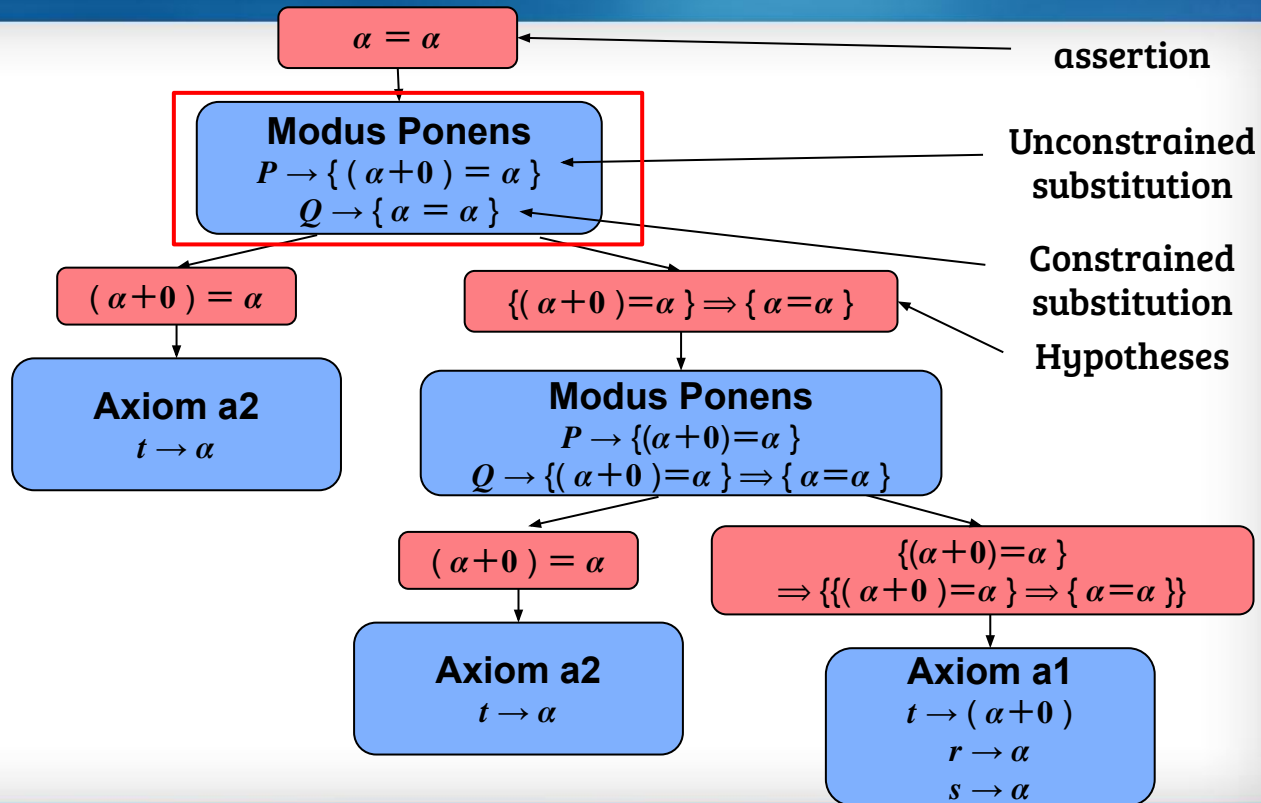
Modus Ponens(mp)

Hypotheses : $\vdash P$
 $\vdash P \Rightarrow Q$

Assertion : $\vdash Q$

Theorem th1

Assertion : $\vdash \alpha = \alpha$





Proof Tree

So this theorem consists of

1. Assertion : $\alpha = \alpha$
2. Hypotheses(& axioms) : $a1, a2, mp$
3. Free variables : α, t, P, \dots
4. Pairs of disjoint variables : $(t, s), (P, Q), \dots$
(restricting substitutions)



Proof Tree

Q. How to find this proof tree?

A. Of course, Deep Learning!



Holophrasm

- Given a theorem, find a proof tree using deep learning.
- 3 neural networks
 - Relevance
 - Generative
 - Payoff



Holophrasm

Theorem mp2b

Hypotheses

- $\vdash \varphi$
- $\vdash \varphi \rightarrow \psi$
- $\vdash \psi \rightarrow \chi$

Assertion

$\vdash \chi$

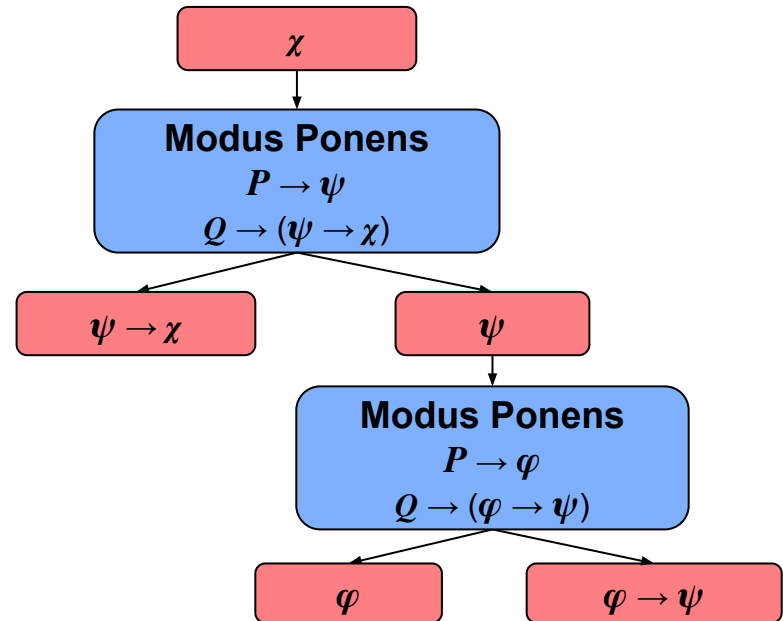
Axiom Modus Ponens

Hypotheses

- $\vdash P$
- $\vdash P \rightarrow Q$

Assertion

$\vdash Q$





Holophrasm

Theorem mp2b

Hypotheses

- $\vdash \varphi$
- $\vdash \varphi \rightarrow \psi$
- $\vdash \psi \rightarrow \chi$

Assertion

$\vdash \chi$

Assertion

χ

Hypotheses

- $\vdash \varphi$
- $\vdash \varphi \rightarrow \psi$
- $\vdash \psi \rightarrow \chi$

Axiom Modus Ponens

Hypotheses

- $\vdash P$
- $\vdash P \rightarrow Q$

Assertion

$\vdash Q$



Holophrasm

Theorem mp2b

Hypotheses

$\vdash \varphi$
 $\vdash \varphi \rightarrow \psi$
 $\vdash \psi \rightarrow \chi$

Assertion

$\vdash \chi$

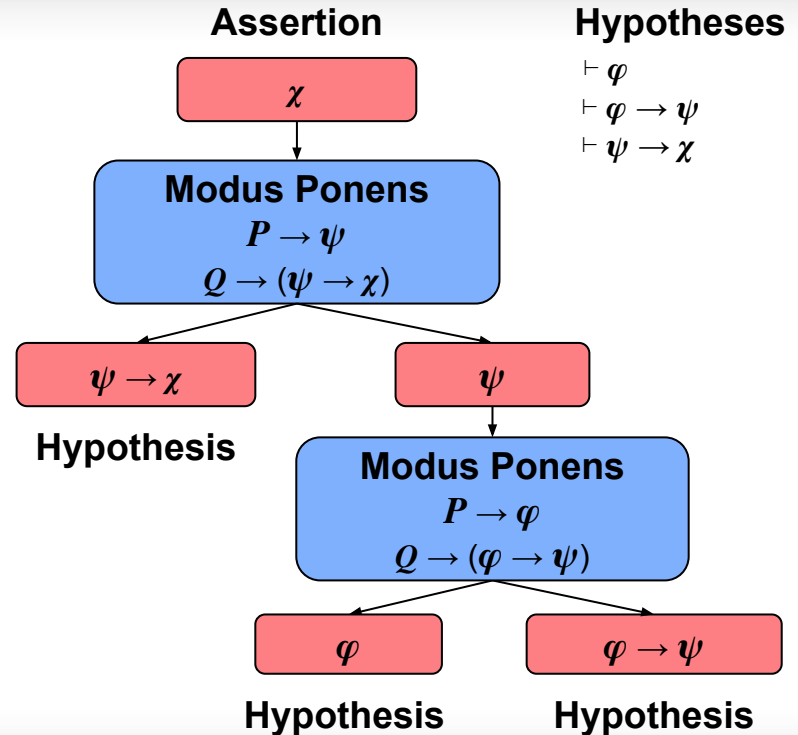
Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

Assertion

$\vdash Q$





Relevance network

- Given an assertion, find the next proposition that will be used to prove the assertion.

Assertion

χ

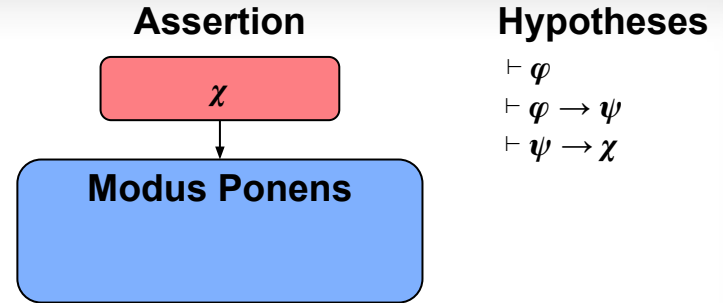
Hypotheses

$\vdash \varphi$
 $\vdash \varphi \rightarrow \psi$
 $\vdash \psi \rightarrow \chi$



Relevance network

- Given an assertion, find the next proposition that will be used to prove the assertion.





Relevance network

- Inputs
 - an assertion and a set of hypotheses of theorem we want to prove

ex)

Theorem mp2b

Hypotheses

Assertion

$\vdash \varphi$

$\vdash \chi$

$\vdash \varphi \rightarrow \psi$

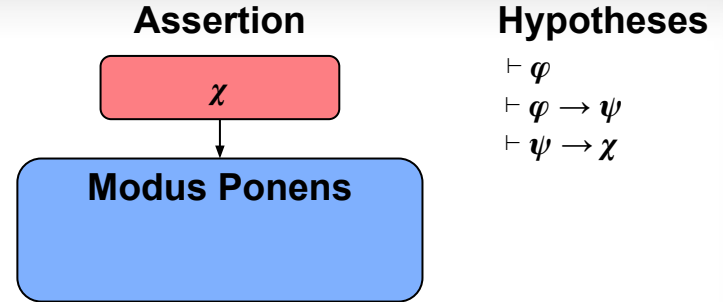
$\vdash \psi \rightarrow \chi$

- Uses RNN
 - GRU



Generative network

- Given an assertion, and the proposition, find the substitutions for the unconstrained variables of the proposition.



Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

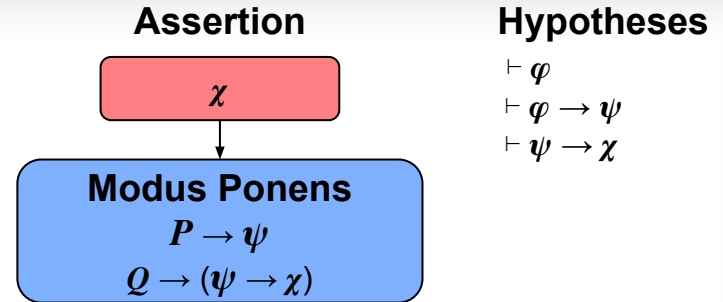
Assertion

$\vdash Q$



Generative network

- Given an assertion, and the proposition, find the substitutions for the unconstrained variables of the proposition.



Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

Assertion

$\vdash Q$



Generative network

- Inputs
 - a set of hypotheses of the theorem we want to prove
 - a set of hypotheses of the proposition

ex)

Hypotheses of mp2b

$\vdash \varphi$
 $\vdash \varphi \rightarrow \psi$
 $\vdash \psi \rightarrow \chi$

Hypotheses of modus Ponens

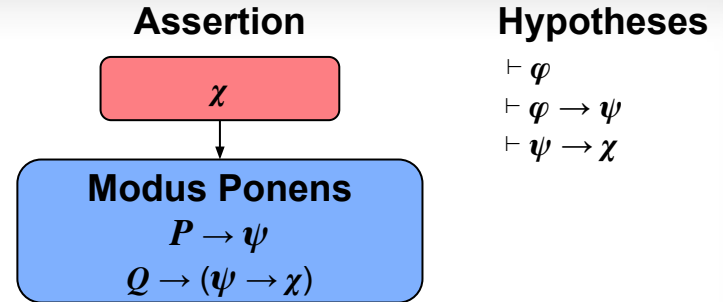
$\vdash P$
 $\vdash P \rightarrow Q$

- Uses sequence to sequence model
 - GRU



Payoff network

- repeat the previous procedures on the remaining assertions to find a proof tree.



Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

Assertion

$\vdash Q$



Payoff network

- repeat the previous procedures on the remaining assertions to find a proof tree.

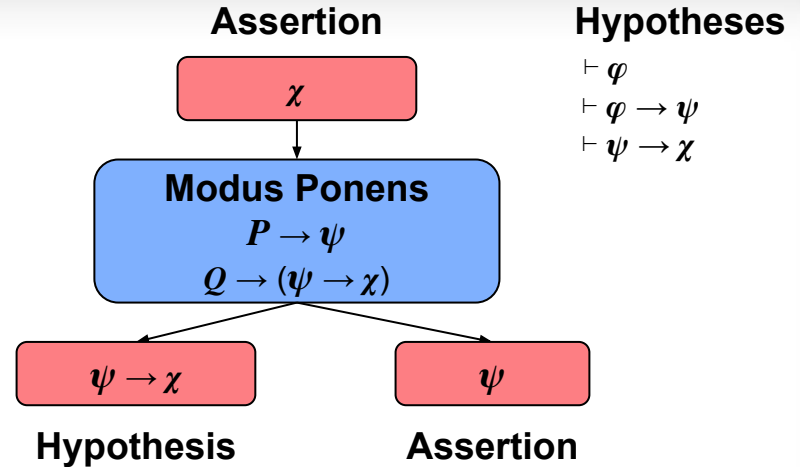
Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

Assertion

$\vdash Q$





Payoff network

- repeat the previous procedures on the remaining assertions to find a proof tree.

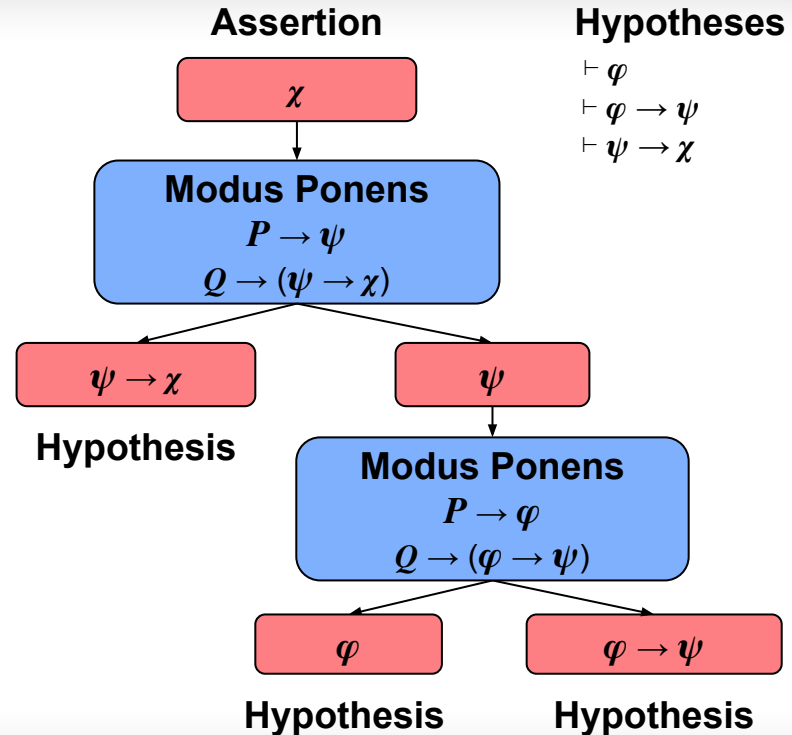
Axiom Modus Ponens

Hypotheses

$\vdash P$
 $\vdash P \rightarrow Q$

Assertion

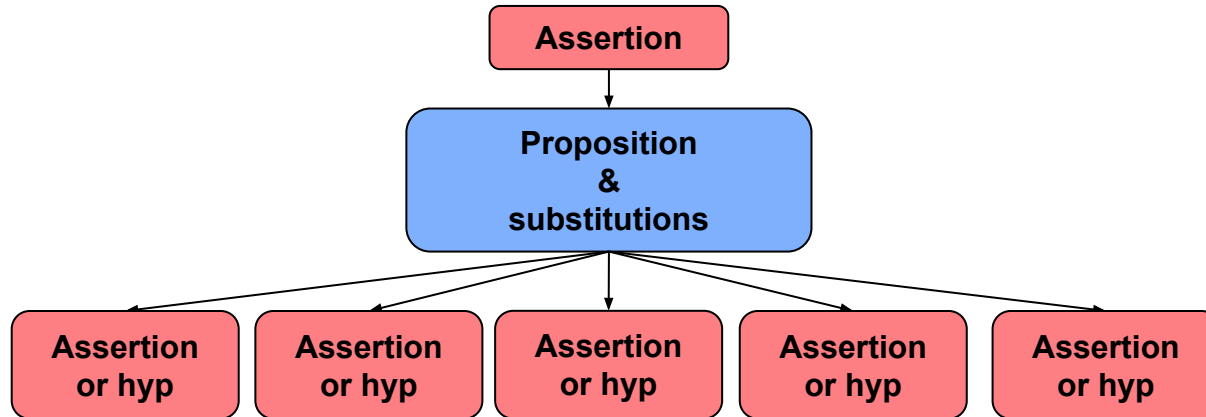
$\vdash Q$





Payoff network

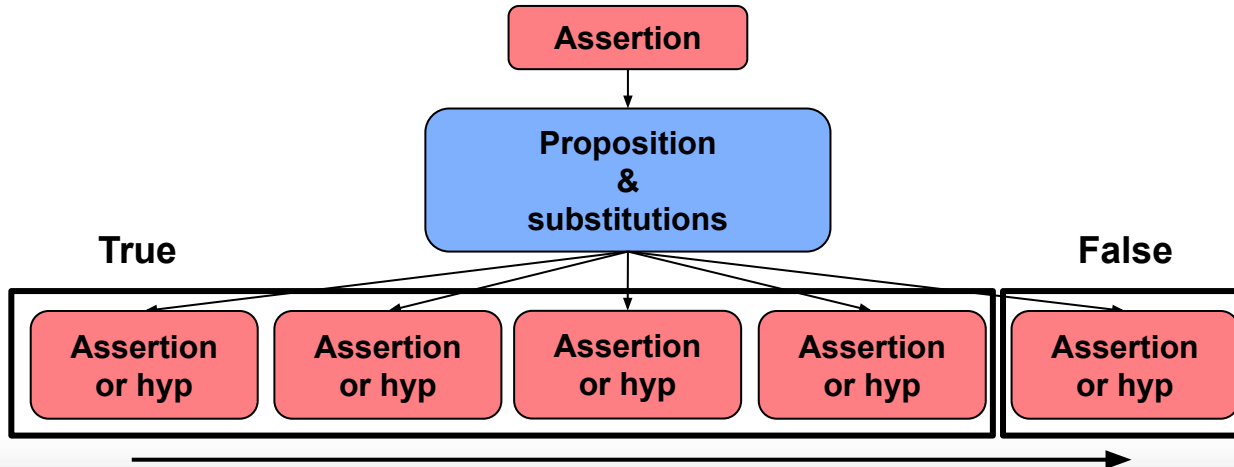
- What if there are too many assertions?





Payoff network

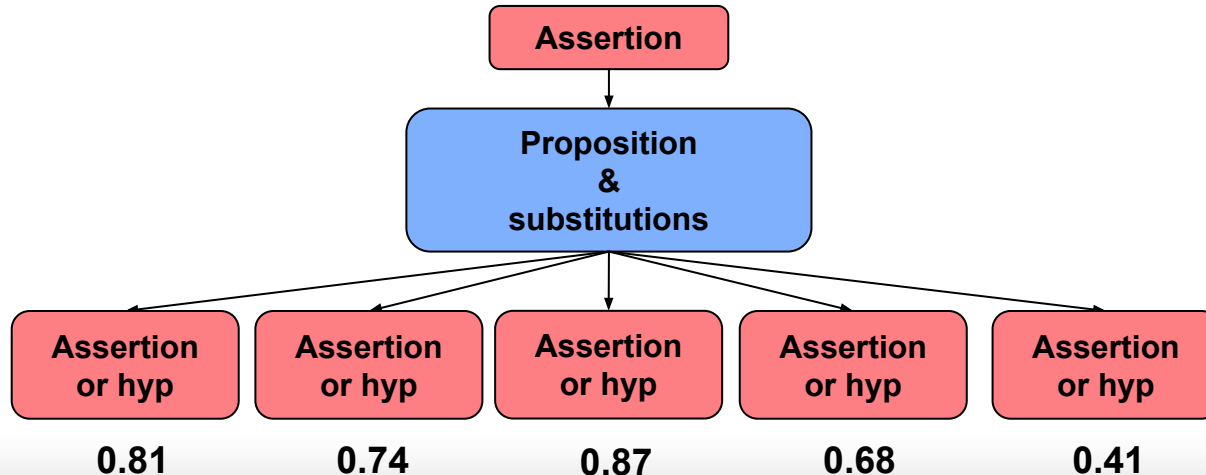
- What if there are too many assertions?





Payoff network

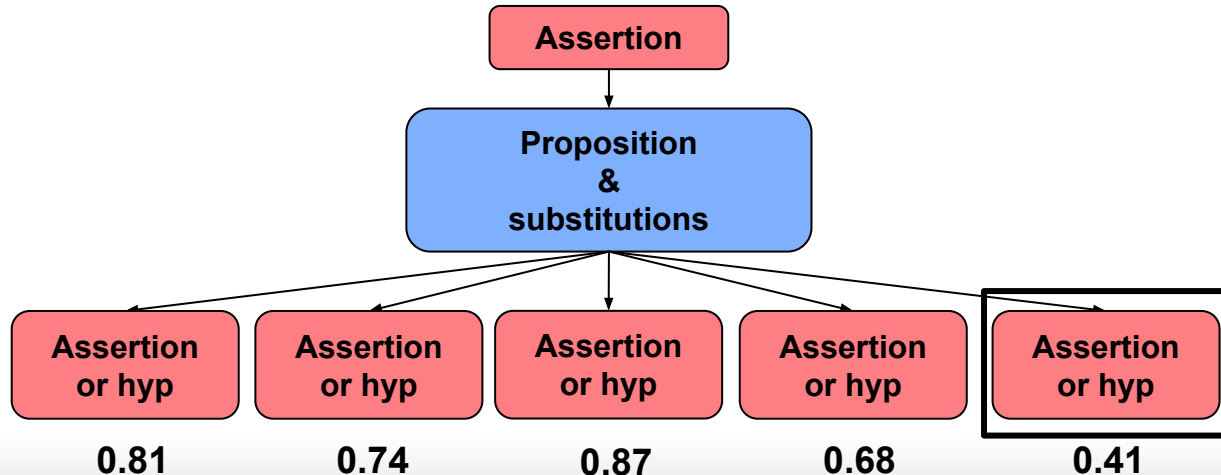
- Given a set of assertions, find the least promising assertion (least probable to be proved)





Payoff network

- Given a set of assertions, find the least promising assertion (least probable to be proved)





Payoff network

- Input: an assertion and a set of hypotheses of theorem
- Uses RNN
 - GRU
 - Bidirectional network



Outline

- Introduction
- Metamath: The Foundation
- Holophrasm: An Overview
- **Future Work**

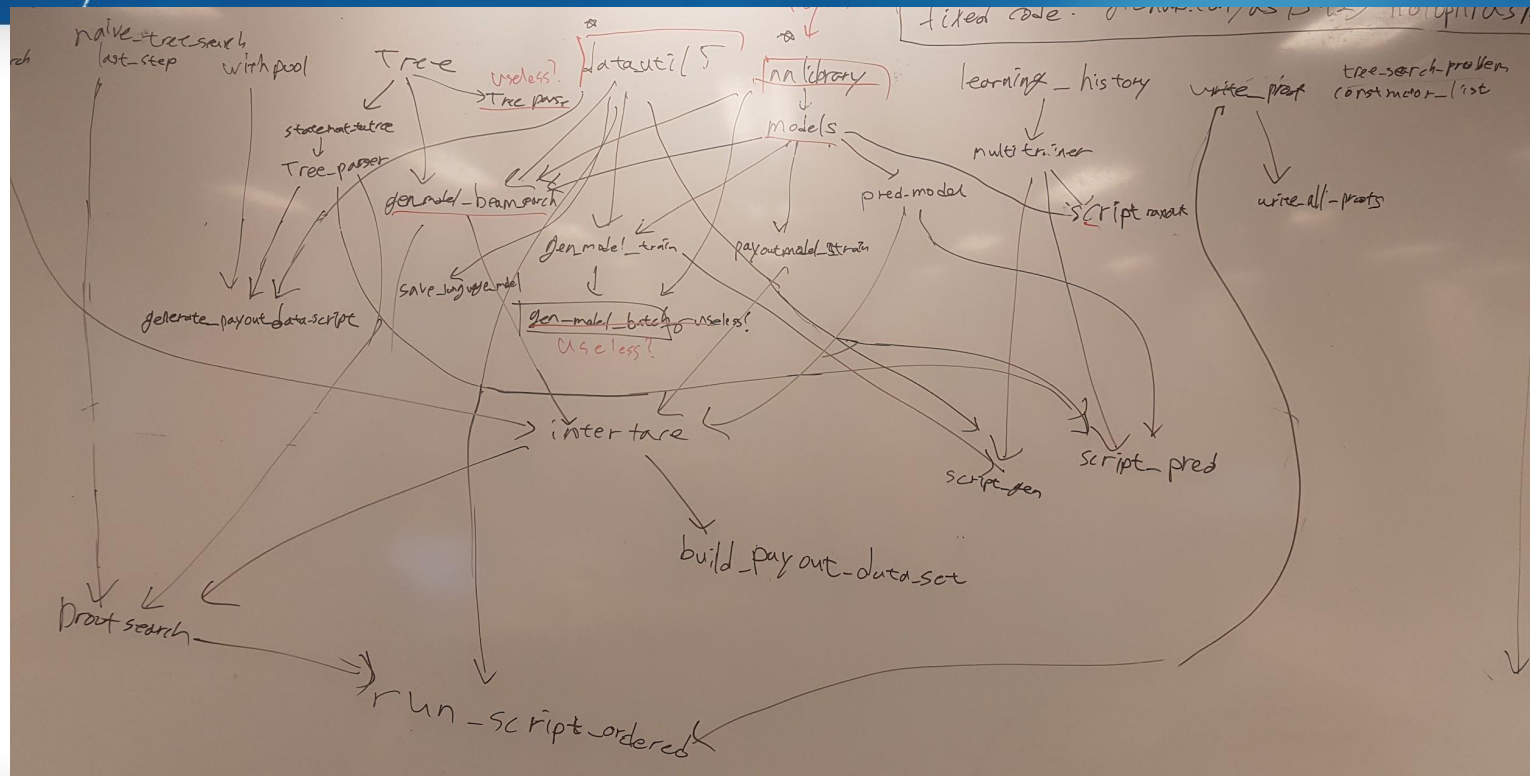


Future Work

- Reconstruct Holophrasm and make descriptions



Future Work





Future Work

Accelerating Deep Learning Research with the Tensor2Tensor Library

Monday, June 19, 2017

Posted by Lukasz Kaiser, Senior Research Scientist, Google Brain Team

Deep Learning (DL) has enabled the rapid advancement of many useful technologies, such as [machine translation](#), [speech recognition](#) and [object detection](#). In the research community, one can find code open-sourced by the authors to help in replicating their results and further advancing deep learning. However, most of these DL systems use unique setups that require significant engineering effort and may only work for a specific problem or architecture, making it hard to run new experiments and compare the results.

Today, we are happy to release [Tensor2Tensor](#) (T2T), an open-source system for training deep learning models in TensorFlow. T2T facilitates the creation of state-of-the-art models for a wide variety of ML applications, such as translation, parsing, image captioning and more, enabling the exploration of various ideas much faster than previously possible. This release also includes a library of datasets and models, including the best models from a few recent papers ([Attention Is All You Need](#), [Depthwise Separable Convolutions for Neural Machine Translation](#) and [One Model to Learn Them All](#)) to help kick-start your own DL research.

Translation Model	Training time	BLEU (difference from baseline)
Transformer (T2T)	3 days on 8 GPU	28.4 (+7.8)
SliceNet (T2T)	6 days on 32 GPUs	26.1 (+5.5)
GNMT + Mixture of Experts	1 day on 64 GPUs	26.0 (+5.4)
ConvS2S	18 days on 1 GPU	25.1 (+4.5)
GNMT	1 day on 96 GPUs	24.6 (+4.0)
ByteNet	8 days on 32 GPUs	23.8 (+3.2)
MOSES (phrase-based baseline)	N/A	20.6 (+0.0)

BLEU scores (higher is better) on the standard WMT English-German translation task.

TensorFlow™


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TensorFlow 1.2 has arrived!

We're excited to announce the release of TensorFlow 1.2! Check out the release notes for all the latest.


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Reference

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Thank you for listening!



Thank you for listening!
Any questions?