

Does Logic-based Reasoning Work for Dutch?

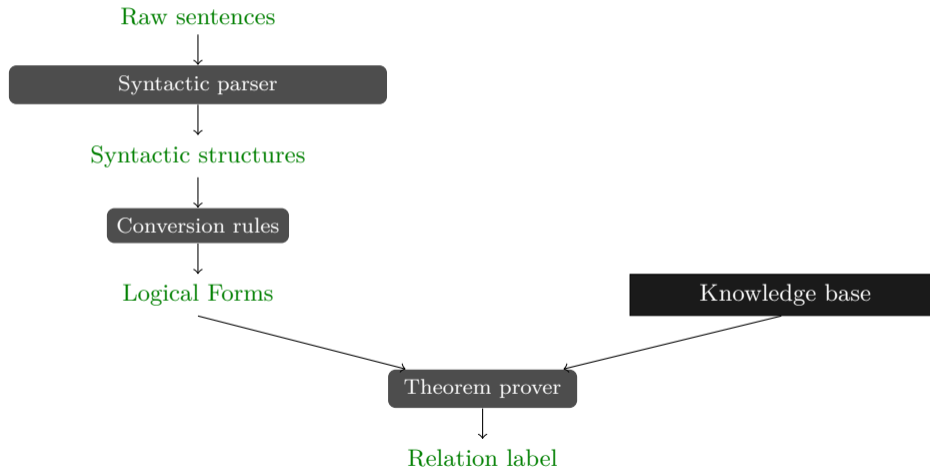
Lasha Abzianidze Konstantinos Kogkalidis

Utrecht Institute of Linguistics OTS, Utrecht University

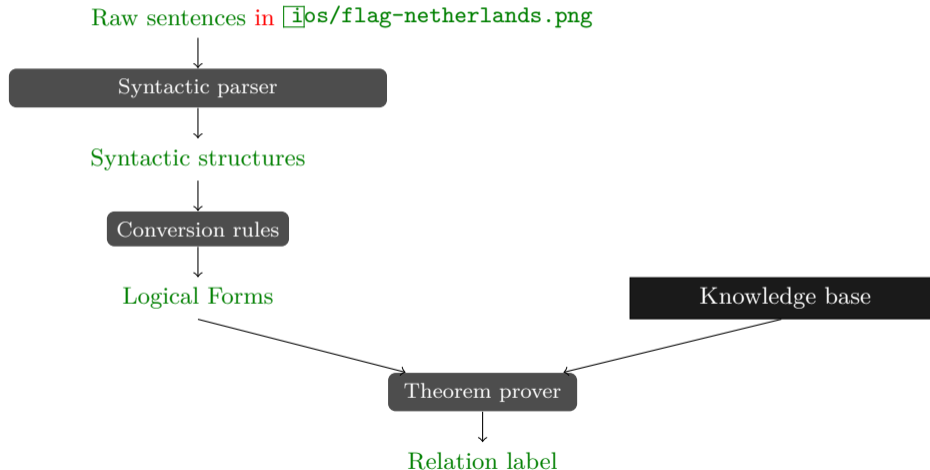
~~06-07-2021 © CLIN~~

21-06-2021 © NLP RG

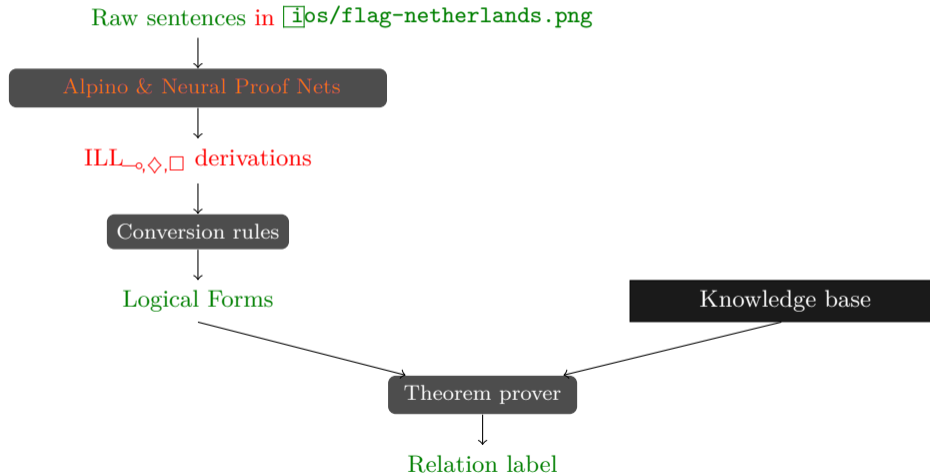
Logic-based approach to NLI



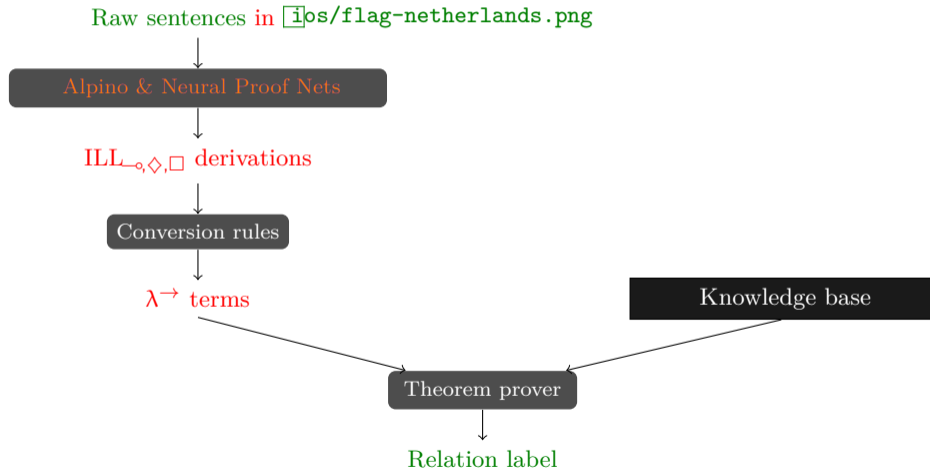
Logic-based approach to NLI



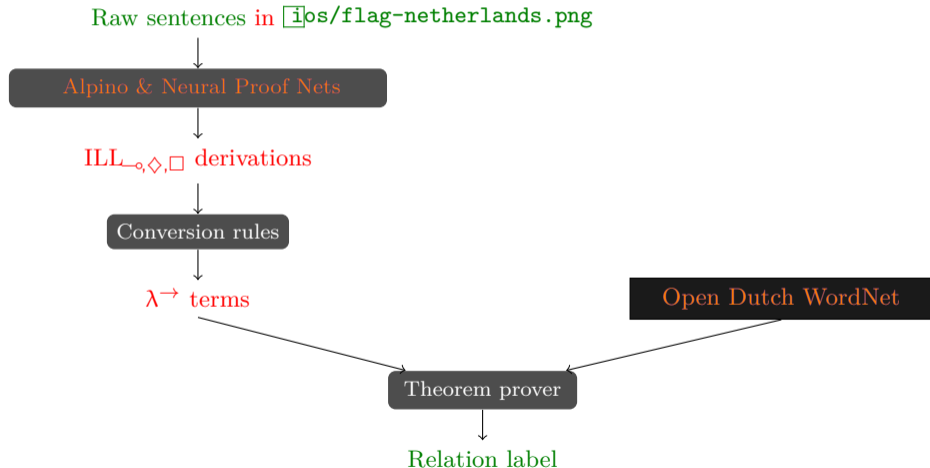
Logic-based approach to NLI



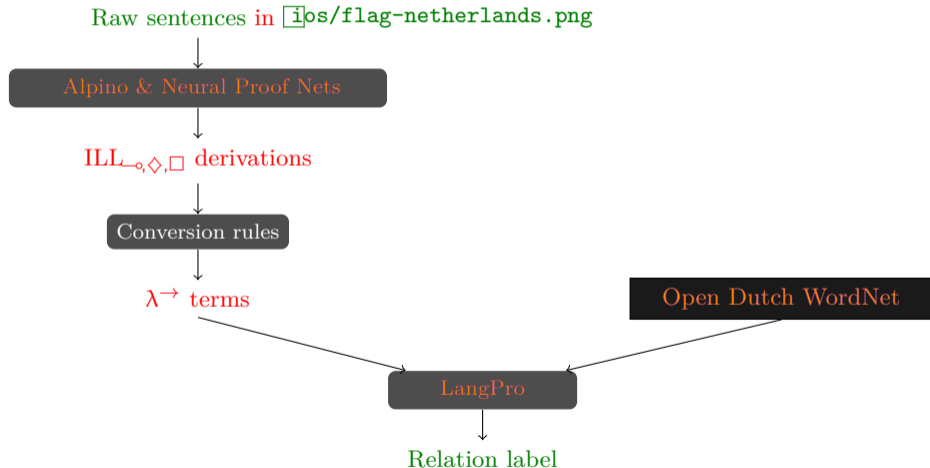
Logic-based approach to NLI



Logic-based approach to NLI



Logic-based approach to NLI



The linear λ -calculus

Words are assigned ILL types, inductively defined as: $\mathcal{T} := a \mid t_1 \multimap t_2 \mid \diamond^\delta t \mid \square^\alpha t$
where

- ▶ a an atom, from a finite set \mathcal{A} :
 $\text{np}, \text{s}_{\text{main}}, \text{s}_{\text{sub}}, \text{pron}, \dots$
- ▶ $t_1 \multimap t_2$ a linear function that consumes t_1 to produce t_2
 $\text{np} \multimap \text{s}_{\text{main}}, \text{np} \multimap (\text{np} \multimap \text{s}_{\text{main}}), (\text{np} \multimap \text{s}_{\text{sub}}) \multimap (\text{np} \multimap \text{np}), \dots$

Syntactic derivations \equiv proofs $\stackrel{\text{chc}}{\equiv}$ functional programs:

$$\tau := c^t \mid (\tau_1^{t_1 \multimap t_2} \tau_2^{t_1})^{t_2} \mid (\lambda x^{t_1}. \tau^{t_2})^{t_1 \multimap t_2} \mid \delta_{\square} \tau \mid \delta^{\square} \tau \mid \delta_{\diamond} \tau \mid \delta^{\diamond} \tau$$

A boy plays:

$$\text{play}^{\diamond \text{su}_{\text{np} \multimap \text{s}_{\text{main}}}} \left(\text{su}^{\diamond} \left(\left(\text{det}_{\square} a^{\square \text{det}(\text{n} \multimap \text{np})} \right) \text{boy}^{\text{n}} \right) \right)$$

The linear λ -calculus

Words are assigned ILL types, inductively defined as: $\mathcal{T} := a \mid t_1 \multimap t_2 \mid \diamond^\delta t \mid \square^\alpha t$
where

- ▶ a an atom, from a finite set \mathcal{A} :

$np, s_{main}, s_{sub}, pron, \dots$

- ▶ $t_1 \multimap t_2$ a linear function that consumes t_1 to produce t_2

$np \multimap s_{main}, np \multimap (np \multimap s_{main}), (np \multimap s_{sub}) \multimap (np \multimap np), \dots$

Syntactic derivations \equiv proofs $\stackrel{chc}{\equiv}$ functional programs:

$\tau := c^t \mid (\tau_1^{t_1 \multimap t_2} \tau_2^{t_1})^{t_2} \mid (\lambda x^{t_1}. \tau^{t_2})^{t_1 \multimap t_2} \mid \delta_\square \tau \mid \delta^\square \tau \mid \delta_\diamond \tau \mid \delta^\diamond \tau$

A boy plays:

$play^{\diamond su_{np \multimap s_{main}}} \left(su^\diamond \left(\left(\det_\square a^{\square \det(n \multimap np)} \right) boy^n \right) \right)$

The linear λ -calculus

Words are assigned ILL types, inductively defined as: $\mathcal{T} := a \mid t_1 \multimap t_2 \mid \diamond^\delta t \mid \square^\alpha t$
where

- ▶ a an atom, from a finite set \mathcal{A} :

$np, s_{main}, s_{sub}, pron, \dots$

- ▶ $t_1 \multimap t_2$ a linear function that consumes t_1 to produce t_2
 $np \multimap s_{main}, np \multimap (np \multimap s_{main}), (np \multimap s_{sub}) \multimap (np \multimap np), \dots$

Syntactic derivations \equiv proofs $\stackrel{chc}{\equiv}$ functional programs:

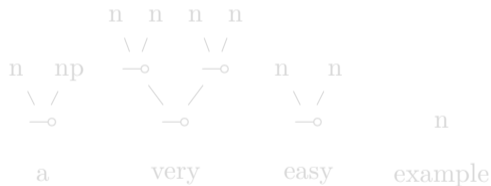
$$\tau := c^t \mid (\tau_1^{t_1 \multimap t_2} \tau_2^{t_1})^{t_2} \mid (\lambda x^{t_1}. \tau^{t_2})^{t_1 \multimap t_2} \mid \delta_{\square} \tau \mid \delta^{\square} \tau \mid \delta_{\diamond} \tau \mid \delta^{\diamond} \tau$$

A boy plays:

$$\text{play}^{\diamond su np \multimap s_{main}} \left(\text{su}^{\diamond} \left(\left(\text{det}_{\square} a^{\square \text{det}(n \multimap np)} \right) \text{boy}^n \right) \right)$$

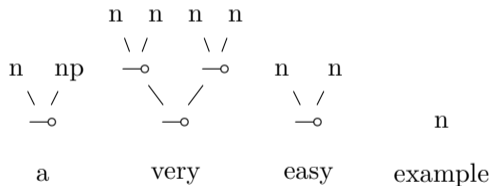
Proof Nets

Types are trees, with nodes inductively polarized: + we have - we seek



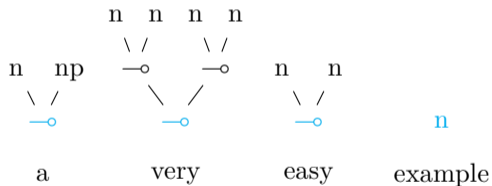
Proof Nets

Types are trees, with nodes inductively polarized: + we have - we seek



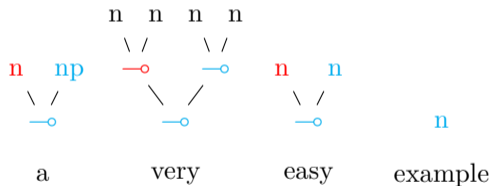
Proof Nets

Types are trees, with nodes inductively polarized: + we have - we seek



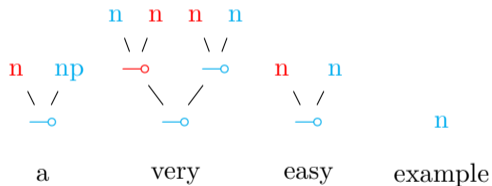
Proof Nets

Types are trees, with nodes inductively polarized: + we have - we seek



Proof Nets

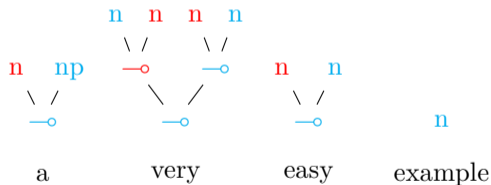
Types are trees, with nodes inductively polarized: + we have - we seek



Proof Nets

Proof net

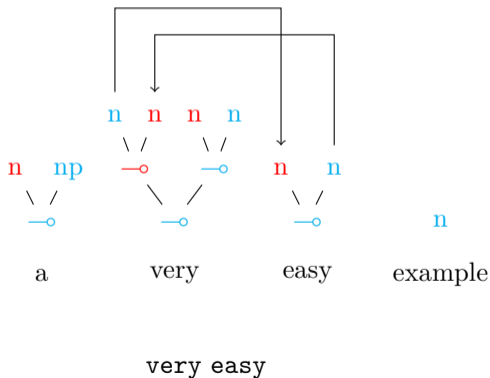
a polarized forest (proof frame) and a bijection between $+$ and $-$ (axiom links)



Proof Nets

Proof net

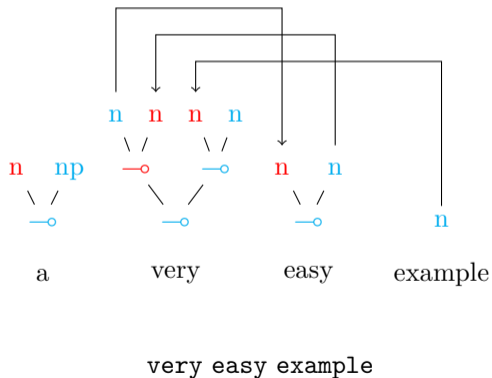
a polarized forest (proof frame) and a bijection between $+$ and $-$ (axiom links)



Proof Nets

Proof net

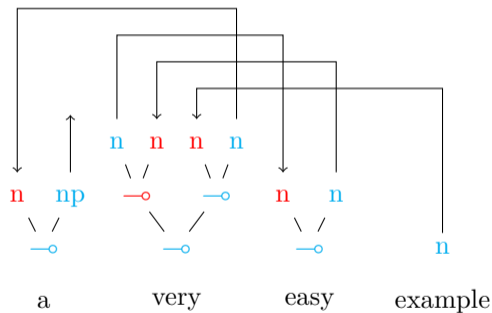
a polarized forest (proof frame) and a bijection between $+$ and $-$ (axiom links)



Proof Nets

Proof net

a polarized forest (proof frame) and a bijection between $+$ and $-$ (axiom links)



a (very easy example)

Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

a very easy example

Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction



a very easy example

Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

n_1



a

very

easy

example

Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

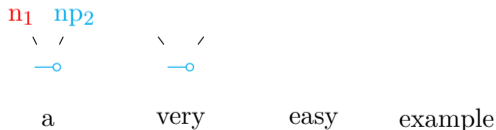
n_1 np_2
 \
 /
 —○

a very easy example

Neural Proof Nets: from sentences to λ -terms

Supertagging

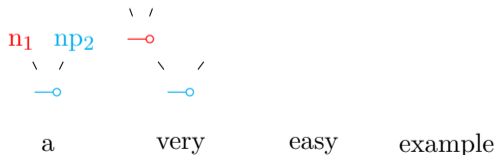
From sentences to proof frames with seq2seq transduction



Neural Proof Nets: from sentences to λ -terms

Supertagging

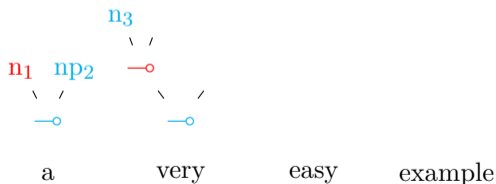
From sentences to proof frames with seq2seq transduction



Neural Proof Nets: from sentences to λ -terms

Supertagging

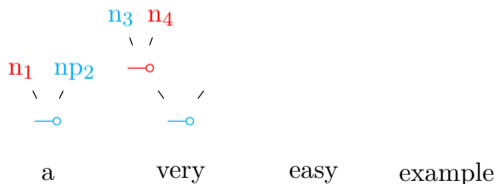
From sentences to proof frames with seq2seq transduction



Neural Proof Nets: from sentences to λ -terms

Supertagging

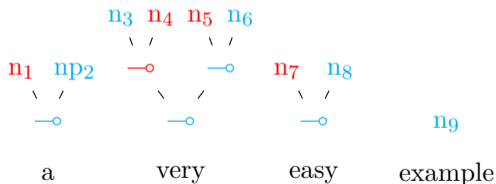
From sentences to proof frames with seq2seq transduction



Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction



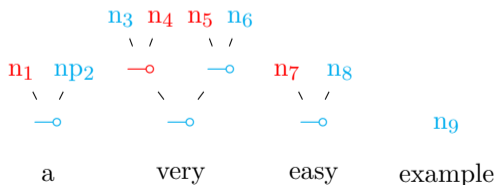
Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

Proving

From proof frames to axiom links with Sinkhorn-Knopp



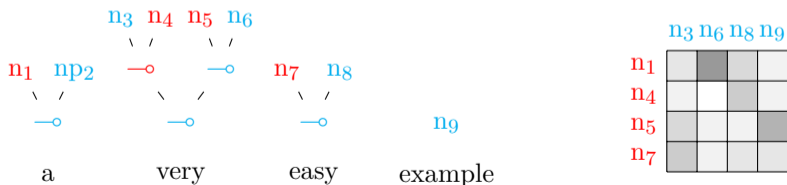
Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

Proving

From proof frames to axiom links with Sinkhorn-Knopp



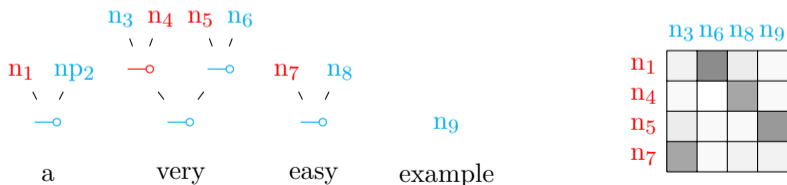
Neural Proof Nets: from sentences to λ -terms

Supertagging

From sentences to proof frames with seq2seq transduction

Proving

From proof frames to axiom links with Sinkhorn-Knopp



Other ingredients: Alpino & ODWN

Alpino:

- ▶ Stochastic Attribute Value Grammar (HPSG) for Dutch
- ▶ Builds dependency structures
- ▶ Used for pre-processing Dutch treebanks

Open Dutch WordNet:

- ▶ 52K synsets (vs 117K Princeton WN)
- ▶ Derived from the Dutch lexical semantic database Cornetto
- ▶ We converted ODWN into the prolog format
- ▶ Used relations: synonymy, hypernymy, antonymy, similarity, and derivation.

Other ingredients: Alpino & ODWN

Alpino:

- ▶ Stochastic Attribute Value Grammar (HPSG) for Dutch
- ▶ Builds dependency structures
- ▶ Used for pre-processing Dutch treebanks

Open Dutch WordNet:

- ▶ 52K synsets (vs 117K Princeton WN)
- ▶ Derived from the Dutch lexical semantic database Cornetto
- ▶ We converted ODWN into the prolog format
- ▶ Used relations: synonymy, hypernymy, antonymy, similarity, and derivation.

Other ingredients: Alpino & ODWN

Alpino:

- ▶ Stochastic Attribute Value Grammar (HPSG) for Dutch
- ▶ Builds dependency structures
- ▶ Used for pre-processing Dutch treebanks

Open Dutch WordNet:

- ▶ 52K synsets (vs 117K Princeton WN)
- ▶ Derived from the Dutch lexical semantic database Cornetto
- ▶ We converted ODWN into the prolog format
- ▶ Used relations: synonymy, hypernymy, antonymy, similarity, and derivation.

Natural Tableau: proving ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

Natural Tableau: proving ios/flag-united-kingdom.png

1 a hedgehog (be (λx . a boy (λy . by y cradle x))) : \mathbb{T}

2 a person (λx . an animal (λy . hold y x)) : \mathbb{F}

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

Natural Tableau: proving ios/flag-united-kingdom.png

1 a hedgehog (be (λx . a boy (λy . by y cradle x))) : \mathbb{T}

2 a person (λx . an animal (λy . hold y x)) : \mathbb{F}

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy (λy . by y cradle h) : \mathbb{T}

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

Natural Tableau: proving ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be $(\lambda x. \text{a boy } (\lambda y. \text{by } y \text{ cradle } x))) : \mathbb{T}$

2 a person $(\lambda x. \text{an animal } (\lambda y. \text{hold } y \ x)) : \mathbb{F}$

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy $(\lambda y. \text{by } y \text{ cradle } h) : \mathbb{T}$

[4] |

5 boy : [b] : \mathbb{T}

6 by b cradle : [h] : \mathbb{T}

Natural Tableau: proving ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be $(\lambda x. \text{a boy } (\lambda y. \text{by } y \text{ cradle } x))) : \mathbb{T}$

2 a person $(\lambda x. \text{an animal } (\lambda y. \text{hold } y \ x)) : \mathbb{F}$

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy $(\lambda y. \text{by } y \text{ cradle } h) : \mathbb{T}$

[4] |

5 boy : [b] : \mathbb{T}

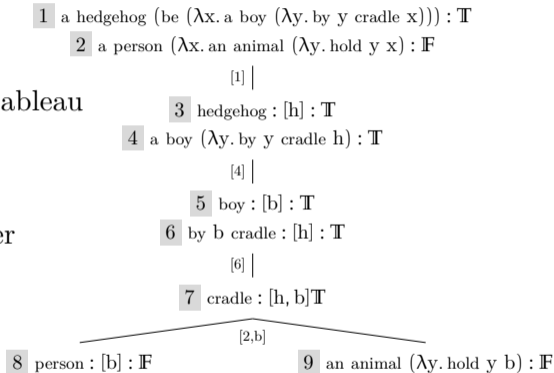
6 by b cradle : [h] : \mathbb{T}

[6] |

7 cradle : [h, b] \mathbb{T}

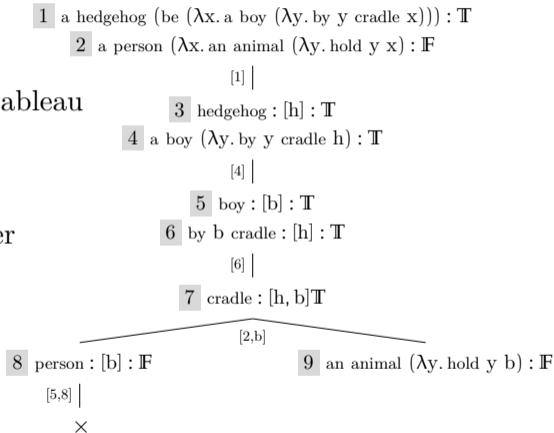
Natural Tableau: proving ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal



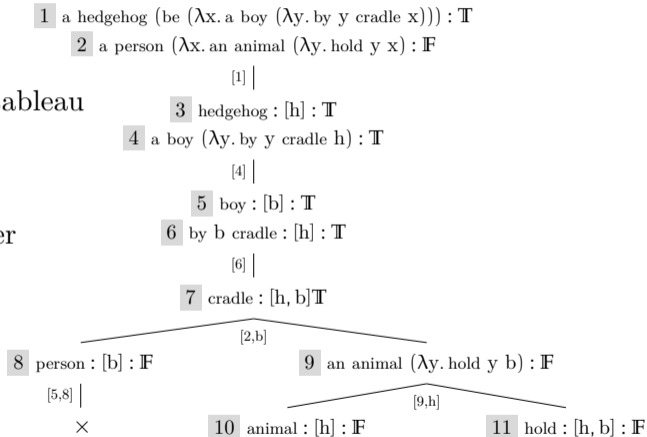
Natural Tableau: proving ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal



Natural Tableau: proving

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal



Natural Tableau: proving

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be (λx . a boy (λy . by y cradle x))) : \mathbb{T}

2 a person (λx . an animal (λy . hold y x)) : \mathbb{F}

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy (λy . by y cradle h) : \mathbb{T}

[4] |

5 boy : [b] : \mathbb{T}

6 by b cradle : [h] : \mathbb{T}

[6] |

7 cradle : [h, b] : \mathbb{T}

[2,b]

8 person : [b] : \mathbb{F}

9 an animal (λy . hold y b) : \mathbb{F}

[5,8] |

×

[9,h]

10 animal : [h] : \mathbb{F}

11 hold : [h, b] : \mathbb{F}

[3,10] |

×

Natural Tableau: proving

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be (λx . a boy (λy . by y cradle x))) : \mathbb{T}

2 a person (λx . an animal (λy . hold y x)) : \mathbb{F}

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy (λy . by y cradle h) : \mathbb{T}

[4] |

5 boy : [b] : \mathbb{T}

6 by b cradle : [h] : \mathbb{T}

[6] |

7 cradle : [h, b] : \mathbb{T}

[2,b]

8 person : [b] : \mathbb{F}

9 an animal (λy . hold y b) : \mathbb{F}

[5,8] |

×

[9,h]

10 animal : [h] : \mathbb{F}

11 hold : [h, b] : \mathbb{F}

[3,10] |

×

[7,11] |

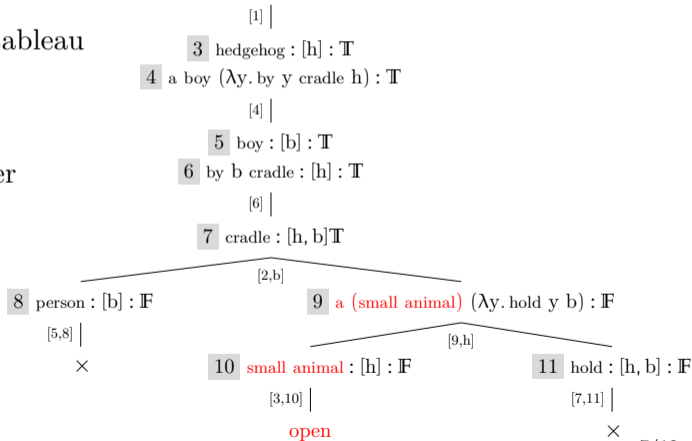
×

Natural Tableau: proving and learning .png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be $(\lambda x. a \text{ boy } (\lambda y. \text{by } y \text{ cradle } x))) : \mathbb{T}$

2 a person $(\lambda x. a \text{ (small animal)} (\lambda y. \text{hold } y \text{ } x)) : \mathbb{F}$



Natural Tableau: proving and learning ios/flag-united-kingdom.png

- ▶ Natural logic + semantic tableau
- ▶ Uniform to Ent./Cont.
- ▶ Prove with refutation
- ▶ Uniform to premise number
- ▶ \approx Syntactic trees
- ▶ Native higher-order logic
- ▶ Decomposing meaning
- ▶ Learning as abduction:
hedgehog \sqsubseteq small animal

1 a hedgehog (be $(\lambda x. a \text{ boy } (\lambda y. \text{by } y \text{ cradle } x))) : \mathbb{T}$

2 a person $(\lambda x. a \text{ (small animal)} (\lambda y. \text{hold } y \text{ } x)) : \mathbb{F}$

[1] |

3 hedgehog : [h] : \mathbb{T}

4 a boy $(\lambda y. \text{by } y \text{ cradle } h) : \mathbb{T}$

[4] |

5 boy : [b] : \mathbb{T}

6 by b cradle : [h] : \mathbb{T}

[6] |

7 cradle : [h, b] \mathbb{T}

[2,b]

8 person : [b] : \mathbb{F}

9 a (small animal) $(\lambda y. \text{hold } y \text{ } b) : \mathbb{F}$

[5,8] |

×

[9,h]

10 small animal : [h] : \mathbb{F}

11 hold : [h, b] : \mathbb{F}

[3,10] |

open

[7,11] |

×

Syntactic λ -terms to λ -logical forms

$\text{play}^{\diamond \text{su np} \rightarrow \text{s}_{\text{main}}} \left(\text{su}^{\diamond} \left(\left(\text{det}_{\square} \text{a}^{\square \text{det}(\text{n} \rightarrow \text{onp})} \right) \text{boy}^{\text{n}} \right) \right) \rightsquigarrow \text{play}^{\text{np}, \text{s}} \left(\text{a}^{\text{n}, \text{np}} \text{boy}^{\text{n}} \right)$

$\text{large}^{\text{np}, \text{np}} \left(\text{brown}^{\text{np}, \text{np}} \left(\text{a}^{\text{n}, \text{np}} \text{dog}^{\text{n}} \right) \right) \rightsquigarrow \text{a}^{\text{n}, \text{np}} \left(\text{large}^{\text{n}, \text{n}} \left(\text{brown}^{\text{n}, \text{n}} \text{dog}^{\text{n}} \right) \right)$

$\text{and} \left(\lambda x. \text{brown}(x \text{ dog}) \right) \left(\lambda y. \text{black}(y \text{ dog}) \right) \text{no}$
 $\rightsquigarrow \text{and}^{\text{np}, \text{np}, \text{np}} \left(\text{no} \left(\text{brown} \text{dog} \right) \right) \left(\text{no} \left(\text{black} \text{dog} \right) \right)$

$\text{cut}^{\text{pp}, \text{n}, \text{np}, \text{s}} \left(\text{in}^{\text{n}, \text{pp}} \text{slice}_{\text{n}} \right) \text{meat}_{\text{n}}$
 $\rightsquigarrow \text{cut}^{\text{pp}, \text{np}, \text{np}, \text{s}} \left(\text{in}^{\text{np}, \text{pp}} \left\{ \text{slice}^{\text{n}} \right\}^{\text{np}} \right) \left\{ \text{meat}^{\text{n}} \right\}^{\text{np}}$

- ▶ Map POS tags and shift to slightly Generalized POS tags: UPOS & Penn
- ▶ Use only these syntactic categories: n , np_x , s_x , pp , pr
- ▶ Function words \mapsto canonical terms (excl. prepositions)

Syntactic λ -terms to λ -logical forms

play^{◇_{su} np → o_{s_{main}}} (su[◇] ((det_□ a^{□_{det} (n → onp)}) boyⁿ))) \rightsquigarrow play^{np, s} (a^{n, np} boyⁿ)

large^{np, np} (brown^{np, np} (a^{n, np} dogⁿ)) \rightsquigarrow a^{n, np} (large^{n, n} (brown^{n, n} dogⁿ))

and (λx . brown(x dog)) (λy . black(y dog)) no
 \rightsquigarrow and^{np, np, np} (no (brown dog)) (no (black dog))

cut^{pp, n, np, s} (in^{n, pp} slice_n) meat_n \rightsquigarrow cut^{pp, np, np, s} (in^{np, pp} {sliceⁿ}^{np}) {meatⁿ}^{np}

- ▶ Map POS tags and shift to slightly Generalized POS tags: UPOS & Penn
- ▶ Use only these syntactic categories: n, np_x, s_x, pp, pr
- ▶ Function words \mapsto canonical terms (excl. prepositions)

Syntactic λ -terms to λ -logical forms

play^{◇_{su} np → o_{s_{main}}} (su[◇] ((det_□ a^{□_{det} (n → onp)}) boyⁿ))) \rightsquigarrow play^{np, s} (a^{n, np} boyⁿ)

large^{np, np} (brown^{np, np} (a^{n, np} dogⁿ)) \rightsquigarrow a^{n, np} (large^{n, n} (brown^{n, n} dogⁿ))

and (λx . brown(x dog)) (λy . black(y dog)) no
 \rightsquigarrow and^{np, np, np} (no (brown dog)) (no (black dog))

cut^{pp, n, np, s} (in^{n, pp} slice_n) meat_n \rightsquigarrow cut^{pp, np, np, s} (in^{np, pp} {sliceⁿ}^{np}) {meatⁿ}^{np}

- ▶ Map POS tags and shift to slightly Generalized POS tags: UPOS & Penn
- ▶ Use only these syntactic categories: n, np_x, s_x, pp, pr
- ▶ Function words \mapsto canonical terms (excl. prepositions)

Syntactic λ -terms to λ -logical forms

$\text{play}^{\diamond \text{su np} \rightarrow \text{s}_{\text{main}}} \left(\text{su}^{\diamond} \left(\left(\text{det}_{\square} \text{a}^{\square \text{det}(\text{n} \rightarrow \text{onp})} \right) \text{boy}^{\text{n}} \right) \right) \rightsquigarrow \text{play}^{\text{np,s}} (\text{a}^{\text{n,np}} \text{boy}^{\text{n}})$

$\text{large}^{\text{np,np}} (\text{brown}^{\text{np,np}} (\text{a}^{\text{n,np}} \text{dog}^{\text{n}})) \rightsquigarrow \text{a}^{\text{n,np}} (\text{large}^{\text{n,n}} (\text{brown}^{\text{n,n}} \text{dog}^{\text{n}}))$

$\text{and} (\lambda x. \text{brown}(x \text{ dog})) (\lambda y. \text{black}(y \text{ dog})) \text{no} \rightsquigarrow \text{and}^{\text{np,np,np}} (\text{no} (\text{brown} \text{ dog})) (\text{no} (\text{black} \text{ dog}))$

$\text{cut}^{\text{pp,n,np,s}} (\text{in}^{\text{n,pp}} \text{slice}_{\text{n}}) \text{meat}_{\text{n}} \rightsquigarrow \text{cut}^{\text{pp,np,np,s}} (\text{in}^{\text{np,pp}} \{\text{slice}^{\text{n}}\}_{\text{np}}) \{\text{meat}^{\text{n}}\}_{\text{np}}$

- ▶ Map POS tags and shift to slightly Generalized POS tags: UPOS & Penn
- ▶ Use only these syntactic categories: $n, \text{np}_x, \text{s}_x, \text{pp}, \text{pr}$
- ▶ Function words \mapsto canonical terms (excl. prepositions)

Syntactic λ -terms to λ -logical forms

$\text{play}^{\diamond \text{su np} \rightarrow \text{os}_{\text{main}}} \left(\text{su}^{\diamond} \left(\left(\text{det}_{\square} \text{a}^{\square \text{det}(\text{n} \rightarrow \text{onp})} \right) \text{boy}^{\text{n}} \right) \right) \rightsquigarrow \text{play}^{\text{np}, \text{s}} \left(\text{a}^{\text{n}, \text{np}} \text{boy}^{\text{n}} \right)$

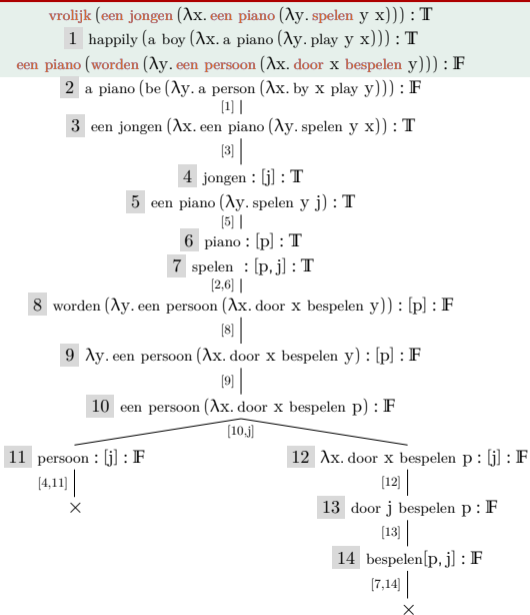
$\text{large}^{\text{np}, \text{np}} \left(\text{brown}^{\text{np}, \text{np}} \left(\text{a}^{\text{n}, \text{np}} \text{dog}^{\text{n}} \right) \right) \rightsquigarrow \text{a}^{\text{n}, \text{np}} \left(\text{large}^{\text{n}, \text{n}} \left(\text{brown}^{\text{n}, \text{n}} \text{dog}^{\text{n}} \right) \right)$

$\text{and} \left(\lambda x. \text{brown}(x \text{ dog}) \right) \left(\lambda y. \text{black}(y \text{ dog}) \right) \text{no}$
 $\rightsquigarrow \text{and}^{\text{np}, \text{np}, \text{np}} \left(\text{no} \left(\text{brown} \text{dog} \right) \right) \left(\text{no} \left(\text{black} \text{dog} \right) \right)$

$\text{cut}^{\text{pp}, \text{n}, \text{np}, \text{s}} \left(\text{in}^{\text{n}, \text{pp}} \text{slice}_{\text{n}} \right) \text{meat}_{\text{n}} \rightsquigarrow \text{cut}^{\text{pp}, \text{np}, \text{np}, \text{s}} \left(\text{in}^{\text{np}, \text{pp}} \left\{ \text{slice}^{\text{n}} \right\}^{\text{np}} \right) \left\{ \text{meat}^{\text{n}} \right\}^{\text{np}}$

- ▶ Map POS tags and shift to slightly Generalized POS tags: UPOS & Penn
- ▶ Use only these syntactic categories: n , np_x , s_x , pp , pr
- ▶ Function words \mapsto canonical terms (excl. prepositions)

Natural Tableau for Dutch ios/



1771 Entailment

Een jongen speelt vrolijk piano
en piano wordt bespeeld door een persoon

1771 Entailment

A boy is happily playing the piano
A piano is being played by a person

Experiments & Results

Parser / pos

Alpino / Alpino

Alpino / spaCy

NPN / Alpino

NPN / spaCy

LangPro 2×2

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ
Alpino / Alpino	72.7
Alpino / spaCy	74.8
NPN / Alpino	72.0
NPN / spaCy	74.3
LangPro 2×2	76.0

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	$T\epsilon$	$T\alpha:T\epsilon$
Alpino / Alpino	72.7	(+9.3)
Alpino / spaCy	74.8	(+9.5)
NPN / Alpino	72.0	(+8.6)
NPN / spaCy	74.3	(+9.1)
LangPro 2×2	76.0	(+9.8)

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1
Alpino / spaCy	74.8	(+9.5)	75.9
NPN / Alpino	72.0	(+8.6)	72.8
NPN / spaCy	74.3	(+9.1)	75.0
LangPro 2\times2	76.0	(+9.8)	77.1

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ	T α :E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1	(+1.8) 75.9
Alpino / spaCy	74.8	(+9.5)	75.9	(+1.7) 77.6
NPN / Alpino	72.0	(+8.6)	72.8	(+1.5) 74.3
NPN / spaCy	74.3	(+9.1)	75.0	(+1.4) 76.4
LangPro 2\times2	76.0	(+9.8)	77.1	(+1.6) 78.7

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ	T α :E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1	(+1.8) 75.9
Alpino / spaCy	74.8	(+9.5)	75.9	(+1.7) 77.6
NPN / Alpino	72.0	(+8.6)	72.8	(+1.5) 74.3
NPN / spaCy	74.3	(+9.1)	75.0	(+1.4) 76.4
LangPro 2\times2	76.0	(+9.8)	77.1	(+1.6) 78.7

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ	T α :E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1	(+1.8) 75.9
Alpino / spaCy	74.8	(+9.5)	75.9	(+1.7) 77.6
NPN / Alpino	72.0	(+8.6)	72.8	(+1.5) 74.3
NPN / spaCy	74.3	(+9.1)	75.0	(+1.4) 76.4
LangPro 2×2	76.0	(+9.8)	77.1	(+1.6) 78.7

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ	T α :E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1	(+1.8) 75.9
Alpino / spaCy	74.8	(+9.5)	75.9	(+1.7) 77.6
NPN / Alpino	72.0	(+8.6)	72.8	(+1.5) 74.3
NPN / spaCy	74.3	(+9.1)	75.0	(+1.4) 76.4
LangPro 2\times2	76.0	(+9.8)	77.1	(+1.6) 78.7

- ▶ POS tagging: spaCy is $1.7 \ll$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \ll$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Experiments & Results

Parser / pos	T ϵ	T α :T ϵ	E ϵ	T α :E ϵ
Alpino / Alpino	72.7	(+9.3)	74.1	(+1.8) 75.9
Alpino / spaCy	74.8	(+9.5)	75.9	(+1.7) 77.6
NPN / Alpino	72.0	(+8.6)	72.8	(+1.5) 74.3
NPN / spaCy	74.3	(+9.1)	75.0	(+1.4) 76.4
LangPro 2×2	76.0	(+9.8)	77.1	(+1.6) 78.7

- ▶ POS tagging: spaCy is $1.7 \leq$ better than Alpino;
- ▶ Parsers: Alpino > NPN? Sentences parsed 98.6% vs 94.9%;
- ▶ The ensemble is $1.2 \leq$ better than the best monoLP;
- ▶ The number of false proofs increases by 4% after α bductive learning.

Comparison to the transformer-based NLI models

Models	All	Ent	Cont
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0	86.2	86.7
mBERT	79.9	79.0	81.9
RobBert	81.7	76.9	85.3

Comparison to the transformer-based NLI models

Models	All $\pm\Delta$	Ent	Cont
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0 -0.3	86.2	86.7
mBERT	79.9 +0.7	79.0	81.9
RobBert	81.7 +0.9	76.9	85.3

Comparison to the transformer-based NLI models

Models	All $\pm\Delta$	Ent $\pm\Delta$	Cont
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0 -0.3	86.2 +2.0	86.7
mBERT	79.9 +0.7	79.0 +4.7	81.9
RobBert	81.7 +0.9	76.9 +6.4	85.3

Comparison to the transformer-based NLI models

Models	All $\pm\Delta$	Ent $\pm\Delta$	Cont $\pm\Delta$
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0 -0.3	86.2 +2.0	86.7 +0.8
mBERT	79.9 +0.7	79.0 +4.7	81.9 +3.1
RobBert	81.7 +0.9	76.9 +6.4	85.3 +1.1

Comparison to the transformer-based NLI models

Models	All $\pm\Delta$	Ent $\pm\Delta$	Cont $\pm\Delta$
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0 -0.3	86.2 +2.0	86.7 +0.8
mBERT	79.9 +0.7	79.0 +4.7	81.9 +3.1
RobBert	81.7 +0.9	76.9 +6.4	85.3 +1.1

Problems failed by all three DL models but solved by LangPro:

1556 Entailment

A man is carrying a **tree**
A man is carrying a **plant**

175 Entailment

A family is watching a little boy who is hitting a baseball
A boy is hitting a baseball

Comparison to the transformer-based NLI models

Models	All $\pm\Delta$	Ent $\pm\Delta$	Cont $\pm\Delta$
LangPro 2×2	78.7	50.6	66.3
BERTje	82.0 -0.3	86.2 +2.0	86.7 +0.8
mBERT	79.9 +0.7	79.0 +4.7	81.9 +3.1
RobBert	81.7 +0.9	76.9 +6.4	85.3 +1.1

Problems failed by all three DL models but solved by LangPro:

1556 Entailment

A man is carrying a **tree**
A man is carrying a **plant**

175 Entailment

A family is watching a little boy who is hitting a baseball
A boy is hitting a baseball

4092 Contradiction

The person is **not** drawing
A man is drawing a picture

6356 Contradiction

A woman in a red dress is **putting away** an instrument
A woman in a red dress is **playing** an instrument

Findings: SICK NL & Open Dutch WN

SICK NL is more challenging than the original dataset due to MT:

- ▶ Transferred gold labels are not gold:

3181 Neutral?		3181 Neutral
A man is trekking in the woods	↔	Een man is aan het wandelen in het bos
<hr/> The man is not hiking in the woods		<hr/> De man is niet aan het wandelen in het bos

- ▶ Extra reasoning due to translation shifts:
drawing a picture ↦ een tekening maakt | tekent een foto
dirt bike race ↦ crossmotorwedstrijd | crossmotorrace.

Lexical relations learned from the training set:

lopen ≡ rennen

pizza ⊆ voedsel/food

halter/dumbbell ⊆ gewicht/weight

leeg/empty | vol/full

Findings: SICK NL & Open Dutch WN

SICK NL is more challenging than the original dataset due to MT:

- ▶ Transferred gold labels are not gold:

3181 Neutral?		3181 Neutral
A man is trekking in the woods	↔	Een man is aan het wandelen in het bos
<hr/> The man is not hiking in the woods		<hr/> De man is niet aan het wandelen in het bos

- ▶ Extra reasoning due to translation shifts:

drawing a picture ↦ een tekening maakt | tekent een foto

dirt bike race ↦ crossmotorwedstrijd | crossmotorrace.

Lexical relations learned from the training set:

lopen ≡ rennen

pizza ⊆ voedsel/food

halter/dumbbell ⊆ gewicht/weight

leeg/empty | vol/full

Findings: SICK NL & Open Dutch WN

SICK NL is more challenging than the original dataset due to MT:

- ▶ Transferred gold labels are not gold:

3181 Neutral?		3181 Neutral
A man is trekking in the woods	↔	Een man is aan het wandelen in het bos
<hr/> The man is not hiking in the woods		<hr/> De man is niet aan het wandelen in het bos

- ▶ Extra reasoning due to translation shifts:

drawing a picture ↦ een tekening maakt | tekent een foto

dirt bike race ↦ crossmotorwedstrijd | crossmotorrace.

Lexical relations learned from the training set:

lopen ≡ rennen

pizza ⊆ voedsel/food

halter/dumbbell ⊆ gewicht/weight

leeg/empty | vol/full

Conclusion & Future work

- ▶ First Dutch NLI system based on logic
- ▶ YES! Logic-based Reasoning Works for Dutch: promising results
- ▶ Automatic translation makes the NLI data more challenging

Conclusion & Future work

- ▶ First Dutch NLI system based on logic
- ▶ YES! Logic-based Reasoning Works for Dutch: promising results
- ▶ Automatic translation makes the NLI data more challenging

- ▶ Employ the Dutch CCG parser
- ▶ Qualitative comparison of the Dutch syntactic parsers
- ▶ Multilingual LangPro: SICK <https://www.aclweb.org/anthology/W18-1001> + CCG parsers + logical Forms