

Towards a Compositional Typed Semantics for Universal Dependencies

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Slav Petrov

Syntax helps Semantics

kotini	aratipandu	tinindi
<i>monkey</i>	<i>banana</i>	<i>eat</i>

Syntax helps Semantics

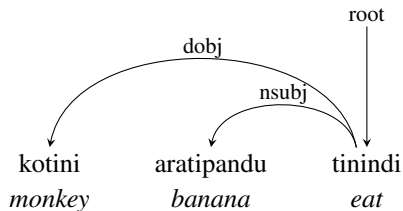
kotini
monkey

aratipandu
banana

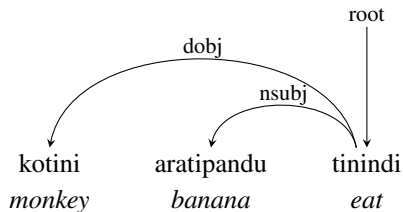
tinindi
eat



Syntax helps Semantics



Syntax helps Semantics



Syntax in humans?

Studies on Peruvian Indian bilinguals indicate Quechua word order influences the local varieties of Spanish [Odlin, 1989]



Syntax in humans?

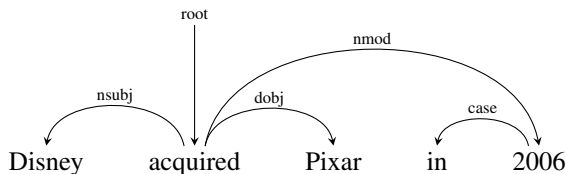
Studies on Peruvian Indian bilinguals indicate Quechua word order influences the local varieties of Spanish [Odlin, 1989]



Arabs show strong preference for SVO in Dutch, whereas Turks for SOV [Jansen et al., 1981; Appel, 1984]

Universal Dependencies

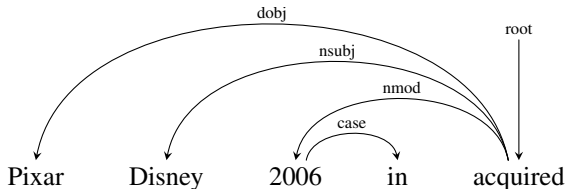
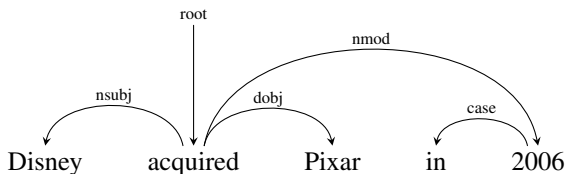
Homogeneous syntactic representation across languages



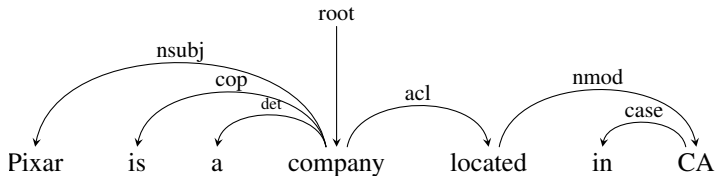
Pixar Disney 2006 in acquired

Universal Dependencies

Homogeneous syntactic representation across languages

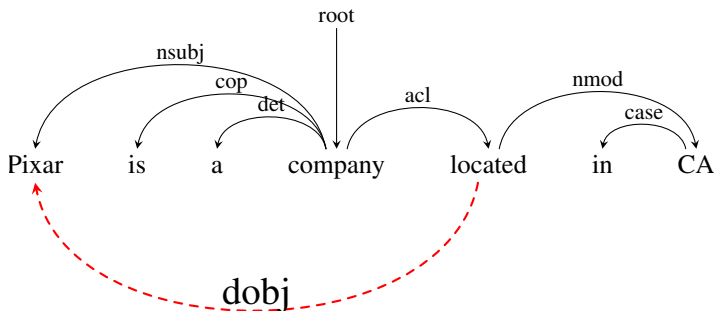


Dependencies to Logical Forms



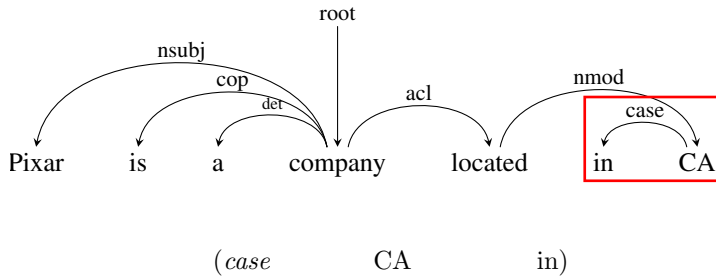
$\exists z. \text{company}(\text{Pixar}) \wedge \text{located}(z_e) \wedge \text{arg}_2(z_e, \text{Pixar}) \wedge \text{arg}_{\text{in}}(z_e, \text{CA})$

Dependencies to Logical Forms

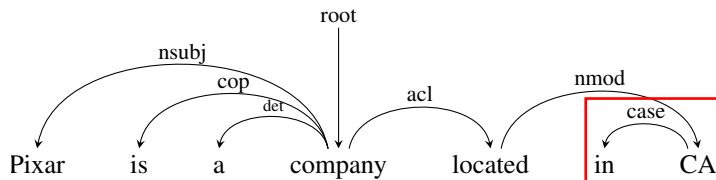


$\exists z. \text{company}(\text{Pixar}) \wedge \text{located}(z_e) \wedge \text{arg}_2(z_e, \text{Pixar}) \wedge \text{arg}_{\text{in}}(z_e, \text{CA})$

Dependencies to Logical Forms

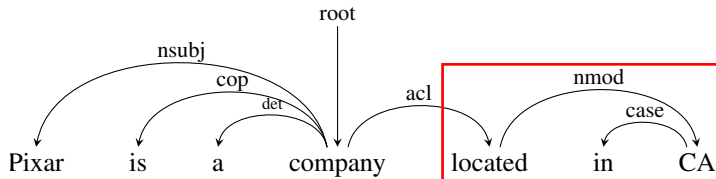


Dependencies to Logical Forms



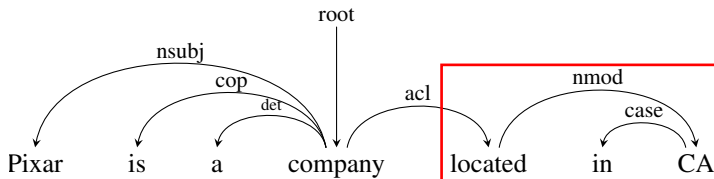
$$\begin{array}{c}
 (case \quad CA \quad in) \\
 \text{?}_e) \quad \lambda f g x. f(x) \quad \lambda x. CA(x_a) \quad \lambda x. empty(x) \\
 \hline
 \lambda x. CA(x_a)
 \end{array}$$

Dependencies to Logical Forms



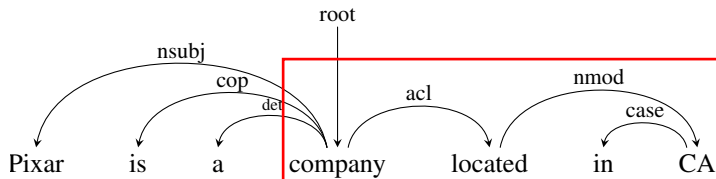
(*nmod* located in CA)).

Dependencies to Logical Forms



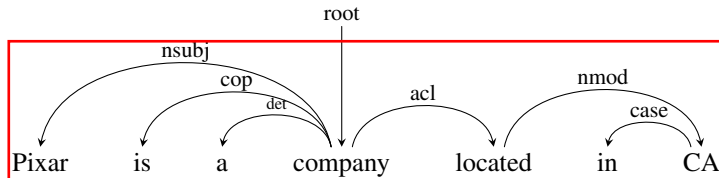
$$\begin{array}{c}
 \begin{array}{c}
 (nmod \quad located \quad in \ CA) \\
 \lambda f g z. \exists x. \quad \lambda x. located(x_e) \quad \frac{}{\lambda x. CA(x_a)} \\
 f(z) \wedge g(x) \wedge \\
 \wedge arg_{in}(z_e, x_a)
 \end{array} \\
 \hline
 \lambda z. located(z_e) \wedge CA(x_a) \wedge arg_{in}(z_e, x_a)
 \end{array}$$

Dependencies to Logical Forms



$$\begin{array}{c}
 \begin{array}{l}
 (acl \\
 \lambda f g x. \exists z. \\
 f(x) \wedge g(x) \wedge \\
 \arg_2(z_e, x_a)
 \end{array}
 \quad
 \begin{array}{l}
 company \\
 \lambda x. compay(x_a)
 \end{array}
 \quad
 \begin{array}{l}
 located\ in\ CA) \\
 \lambda g z. \exists x. located(z_e) \wedge CA(x_a) \wedge \arg_{in}(z_e, x_a)
 \end{array}
 \\
 \hline
 \lambda x. \exists y z. company(x_a) \wedge located(z_e) \wedge CA(y_a) \wedge \arg_2(z_e, x_a) \wedge \arg_{in}(z_e, y_a)
 \end{array}$$

Dependencies to Logical Forms



$$\exists z. \text{company}(\text{Pixar}) \wedge \text{located}(z_e) \wedge \text{arg}_2(z_e, \text{Pixar}) \wedge \text{arg}_{\text{in}}(z_e, \text{CA})$$

Synchronous Syntax-Semantics Interfaces

Derive semantics from syntactic derivation

- ▶ CCG [Bos et al., 2004; Lewis & Steedman, 2013]
- ▶ HPSG [Copestake et al., 2001]
- ▶ LFG [Dalrymple et al., 1995]
- ▶ TAG [Joshi et al., 1995]

CCG has been a popular choice

- ▶ No treebanks for many languages
- ▶ Syntactic categories differ in each language

Why from Universal Dependencies?

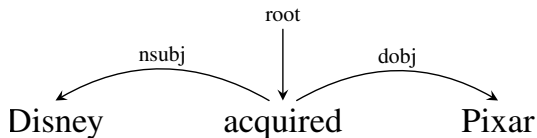
Treebanks in more than 40 languages

Very accurate parsers

[Andor et al., 2016; Chen & Manning, 2014]

Universal Types, both syntactic and semantic

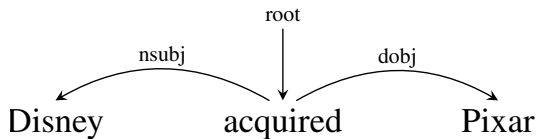
Dependencies to Logical Forms



$$\lambda z. \exists xy. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge \text{Disney}(x_a) \wedge \\ \text{arg}_1(z_e, x_a) \wedge \text{arg}_2(z_e, y_a)$$

Dependencies to Logical Forms

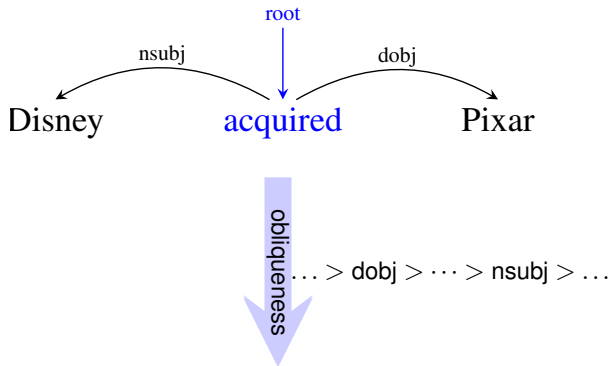
Our Approach



Let dependency labels drive the composition

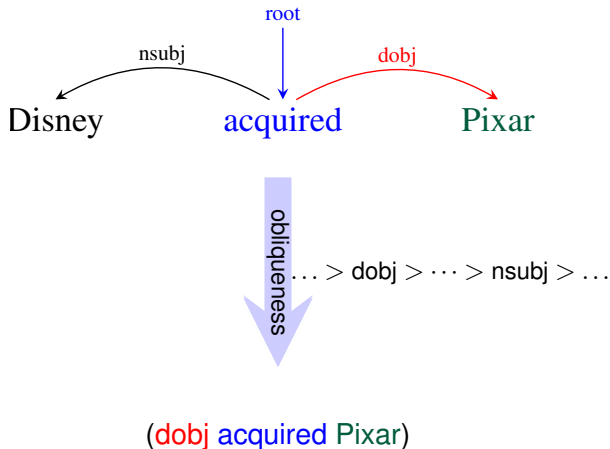
Dependencies to Logical Forms

Our Approach



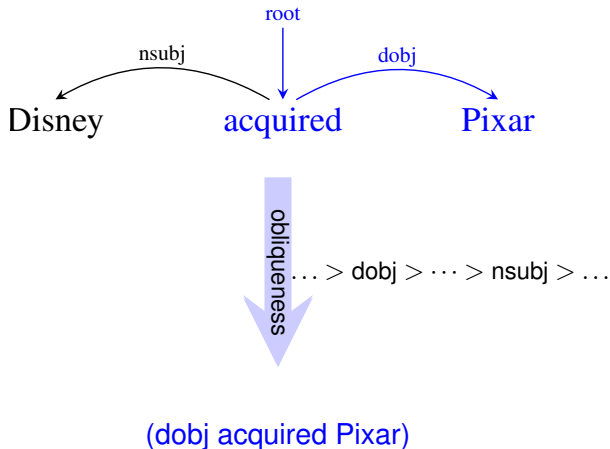
Dependencies to Logical Forms

Our Approach



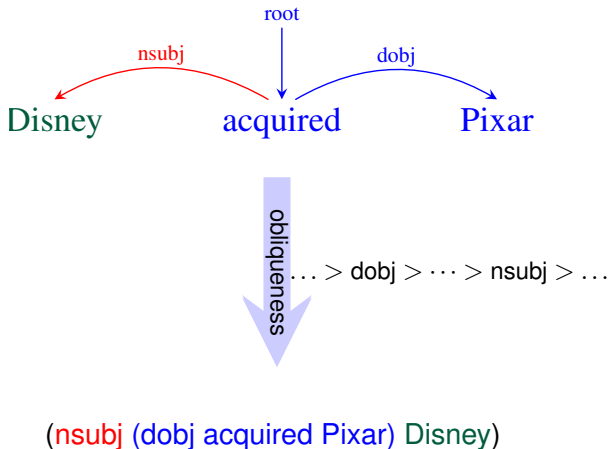
Dependencies to Logical Forms

Our Approach



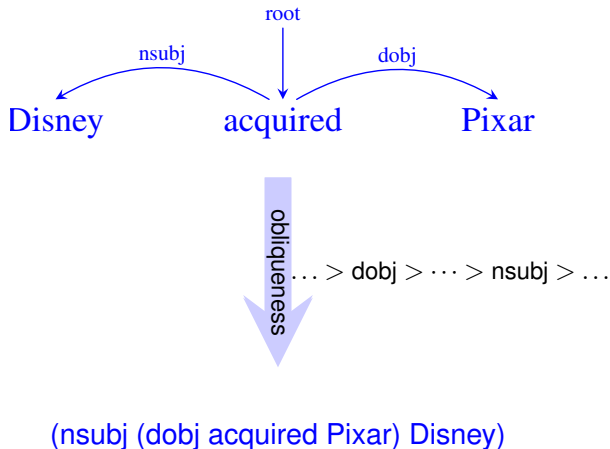
Dependencies to Logical Forms

Our Approach



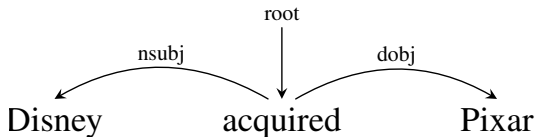
Dependencies to Logical Forms

Our Approach



Dependencies to Logical Forms

Our Approach

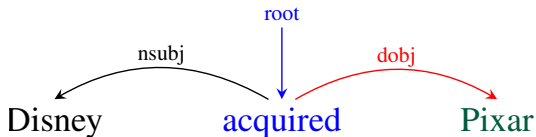


(nsubj (dobj acquired Pixar) Disney)

$$\lambda z. \exists xy. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge \text{Disney}(x_a) \wedge \\ \text{arg}_1(z_e, x_a) \wedge \text{arg}_2(z_e, y_a)$$

Dependencies to Logical Forms

Lambda Calculus

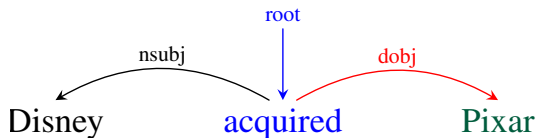


Lambda Calculus Basic Types

- ▶ Individuals: **Ind** (also denoted by $.a$)
- ▶ Events: **Event** (also denoted by $.e$)
- ▶ Truth values: **Bool**

Dependencies to Logical Forms

Lambda Calculus



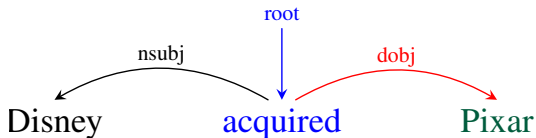
Lambda Expression for words

acquired $\Rightarrow \lambda x. \text{acquired}(x_e)$

Pixar $\Rightarrow \lambda x. \text{Pixar}(x_a)$

Dependencies to Logical Forms

Lambda Calculus

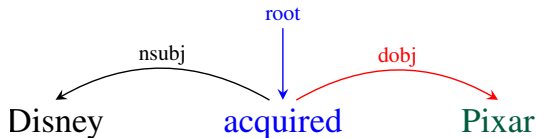


Lambda Expression for dependency labels

$$\text{dobj} \Rightarrow \lambda f \ g \ z . \exists \mathbf{x} . f(\mathbf{z}) \wedge g(\mathbf{x}) \wedge \text{arg}_2(\mathbf{z_e}, \mathbf{x_a})$$

Dependencies to Logical Forms

Lambda Calculus



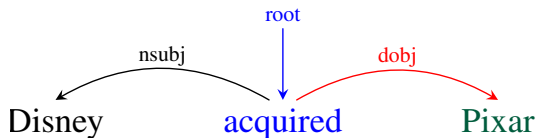
Lambda Expression for dependency labels

$$\text{dobj} \Rightarrow \lambda f \ g \ z . \exists \mathbf{x} . f(\mathbf{z}) \wedge g(\mathbf{x}) \wedge \text{arg}_2(\mathbf{z_e}, \mathbf{x_a})$$

This operation mirrors the tree structure

Dependencies to Logical Forms

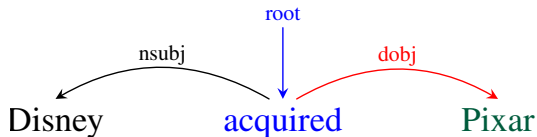
Composition



(dobj	acquired	Pixar)
$\lambda f g z. \exists y.$	$\lambda z. \text{acquired}(z_e)$	$\lambda y. \text{Pixar}(y_a)$
$f(z) \wedge g(y) \wedge$		
$\text{arg}_2(z_e, y_a)$		

Dependencies to Logical Forms

Composition

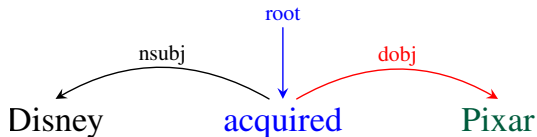


$$\begin{array}{ccc} (\text{dobj} & \text{acquired} & \text{Pixar}) \\ \lambda f g z. \exists y. & \lambda z. \text{acquired}(z_e) & \lambda y. \text{Pixar}(y_a) \\ f(z) \wedge g(y) \wedge & & \\ \text{arg}_2(z_e, y_a) & & \end{array}$$

$$\lambda g z. \exists y. \text{acquired}(z_e) \wedge g(y) \wedge \text{arg}_2(z_e, y_a)$$

Dependencies to Logical Forms

Composition



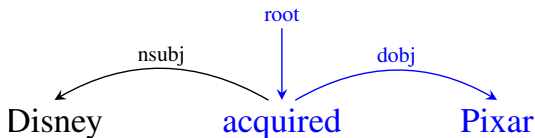
$$\begin{array}{ccc}
 (\text{dobj} & \text{acquired} & \text{Pixar}) \\
 \lambda f g z. \exists y. & \lambda z. \text{acquired}(z_e) & \lambda y. \text{Pixar}(y_a) \\
 f(z) \wedge g(y) \wedge & & \\
 \text{arg}_2(z_e, y_a) & &
 \end{array}$$

$$\begin{array}{c}
 \lambda g z. \exists y. \text{acquired}(z_e) \wedge g(y) \\
 \wedge \text{arg}_2(z_e, y_a)
 \end{array}$$

$$\begin{array}{c}
 \lambda z. \exists y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \\
 \wedge \text{arg}_2(z_e, y_a)
 \end{array}$$

Dependencies to Logical Forms

Composition

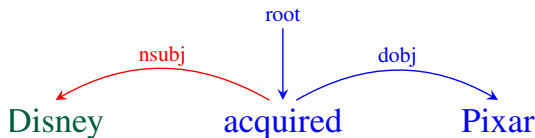


(dobj acquired Pixar)

$$\lambda z. \exists y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \\ \wedge \text{arg}_2(z_e, y_a)$$

Dependencies to Logical Forms

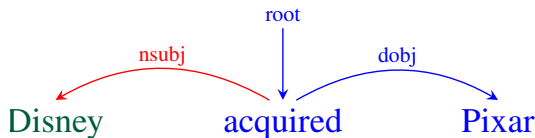
Composition



$$\begin{array}{c}
 (\text{nsubj} \quad (\text{dobj} \quad \text{acquired} \quad \text{Pixar}) \quad \text{Disney}) \\
 \lambda f g z. \exists x. \quad \frac{f(z) \wedge g(x) \wedge \arg_1(z_e, x_a)}{\lambda z. \exists y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge \arg_2(z_e, y_a)} \quad \lambda x. \text{Disney}(x_a)
 \end{array}$$

Dependencies to Logical Forms

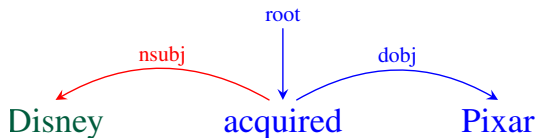
Composition



$$\begin{array}{c}
 \begin{array}{ccc}
 (\text{nsubj} & & (\text{dobj} \quad \text{acquired} \quad \text{Pixar}) \\
 \lambda f g z. \exists x. & & \\
 f(z) \wedge g(x) \wedge & & \\
 \arg_1(z_e, x_a) & & \lambda z. \exists y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \\
 & & \wedge \arg_2(z_e, y_a)
 \end{array} \\
 \hline
 \lambda g z. \exists x y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge g(x) \wedge \\
 \arg_1(z_e, x_a) \wedge \arg_2(z_e, y_a)
 \end{array}
 \quad
 \begin{array}{c}
 \text{Disney} \\
 \lambda x. \text{Disney}(x_a)
 \end{array}$$

Dependencies to Logical Forms

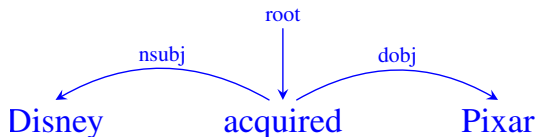
Composition



$$\begin{array}{c}
 \begin{array}{ccc}
 (\text{nsubj} & (\text{dobj} & \text{acquired} & \text{Pixar}) & \text{Disney}) \\
 \lambda f g z. \exists x. & \frac{}{\lambda z. \exists y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a)} & \lambda x. \text{Disney}(x_a) \\
 f(z) \wedge g(x) \wedge & \wedge \arg_2(z_e, y_a) \\
 \arg_1(z_e, x_a) &
 \end{array} \\
 \hline
 \lambda g z. \exists x y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge g(x) \wedge \\
 \arg_1(z_e, x_a) \wedge \arg_2(z_e, y_a) \\
 \hline
 \lambda z. \exists x y. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge \text{Disney}(x_a) \wedge \\
 \arg_1(z_e, x_a) \wedge \arg_2(z_e, y_a)
 \end{array}$$

Dependencies to Logical Forms

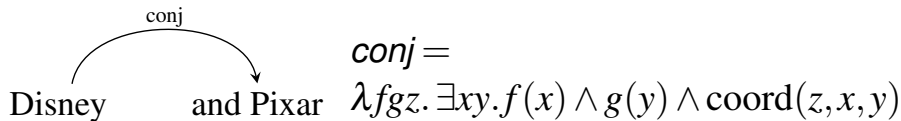
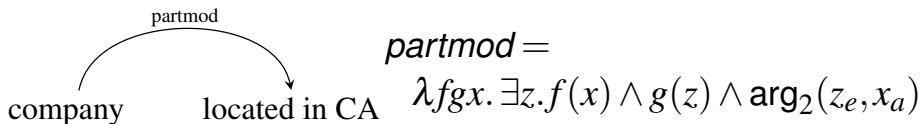
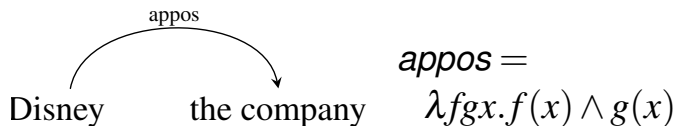
Composition



(nsubj (dobj acquired Pixar) Disney)

$$\lambda z. \exists xy. \text{acquired}(z_e) \wedge \text{Pixar}(y_a) \wedge \text{Disney}(x_a) \wedge$$
$$\text{arg}_1(z_e, x_a) \wedge \text{arg}_2(z_e, y_a)$$

Dependencies to Logical Forms



Comparison with CCG

$$\begin{array}{c}
 \begin{array}{ccc}
 \text{Disney} & \text{acquired} & \text{Pixar} \\
 \hline
 NP & S \backslash NP / NP & NP
 \end{array} \\
 \text{Disney} \quad \lambda y \lambda x \lambda e. \text{acquired}(e) & & \text{Pixar} \\
 \quad \wedge \text{arg}_1(e, x) & & \\
 \quad \wedge \text{arg}_2(e, y) & & \\
 \hline
 & S \backslash NP & > \\
 & \lambda x \lambda e. \text{acquired}(e) & \\
 & \wedge \text{arg}_1(e, x) \wedge \text{arg}_2(e, \text{Pixar}) & \\
 \hline
 & S & < \\
 \lambda e. \text{acquired}(e) \wedge \text{arg}_1(e, \text{Disney}) \wedge \text{arg}_2(e, \text{Pixar}) & &
 \end{array}$$

CCG

Lexicalized semantics

$S \backslash NP / NP : \lambda y \lambda x \lambda e. \text{acquired}(e) \wedge \text{arg}_1(e, x) \wedge \text{arg}_2(e, y)$

Words drive composition

Language specific types

Argument adjunct distinction

$S \backslash NP / PP / NP$ vs. $(S \backslash NP) \backslash (S \backslash NP) / NP$

DepLambda

Simple lexical semantics

$\lambda x. \text{acquired}(x_e)$

Dependencies drive composition

Mostly universal

Every dependent is an adjunct

Lexicalized semantics

$$S \backslash NP / NP : \lambda y \lambda x \lambda e. \text{acquired}(e) \wedge \text{arg}_1(e, x) \wedge \text{arg}_2(e, y)$$

Words drive composition

Language specific types

Argument adjunct distinction

$$S \backslash NP / PP / NP \text{ vs. } (S \backslash NP) \backslash (S \backslash NP) / NP$$

With “complex types” comes power

$$(S[dcl] \backslash NP) / (S[to] \backslash NP_x) / NP_x$$

Simple lexical semantics

$$\lambda x. \text{acquired}(x_e)$$

Dependencies drive composition

Mostly universal

Every dependent is an adjunct

Single-type system is robust, but restricted

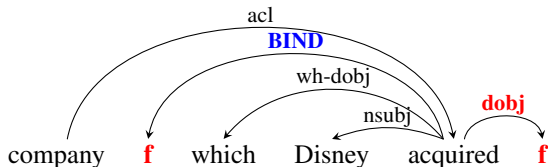
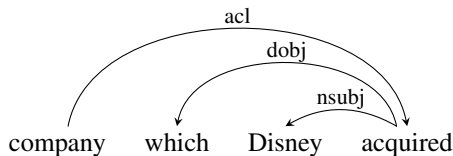
$$\lambda f g x \dots \text{i.e., } \eta \rightarrow \eta \rightarrow \eta \text{ for all labels}$$

Relative clause in CCG

company	which	Disney	acquired
N	$(N \backslash N) / (S_{dcl} / NP)$	N	$(S_{dcl} \backslash NP) / NP$
$\lambda x. company(x)$	$\lambda p \lambda q \lambda x. q(x) \wedge \exists e[p(x, e)]$		$\lambda x \lambda y \lambda e. acquire(e) \wedge A0(y, e) \wedge A1(x, e)$
		NP	
		disney	
		$S_X / (S_X \backslash NP)$	
		$\lambda p \lambda e. p(disney, e)$	
			S_{dcl} / NP
			$\lambda x \lambda e. acquire(e) \wedge A0(disney, e) \wedge A1(x, e)$
			NN
			$\lambda p \lambda x. p(x) \wedge \exists e[acquire(e) \wedge A0(disney, e) \wedge A1(x, e)]$
			N
			$\lambda x. company(x) \wedge \exists e[acquire(e) \wedge A0(disney, e) \wedge A1(x, e)]$

Relative Clause in DepLambda

following Carpenter (1998)



DepLambda in a nutshell

Dependency tree is a series of compositions

- ▶ Dependency label defines the composition function
- ▶ Each function takes two typed-semantic sub-expressions
- ▶ Returns typed-semantics of the larger expression

Function could be any computation

- ▶ e.g., a neural network

Richer context-sensitive types will allow richer composition functions

- ▶ e.g., neural networks with tensors/neural networks
- ▶ directly to the target-application semantics

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DepLambda in a nutshell

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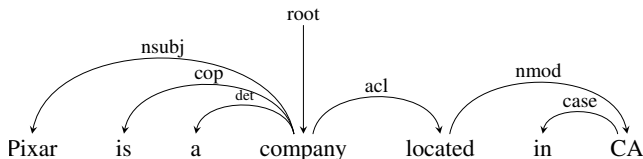
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Recap



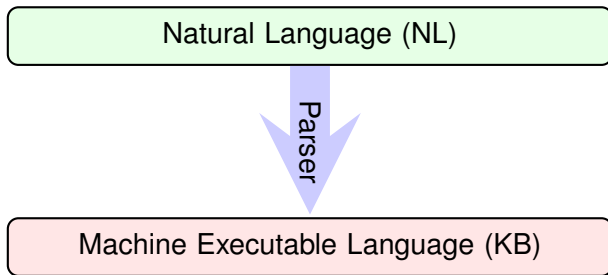
... > dobj > ... > nsubj > ...

$$\begin{array}{c}
 \dots \quad (acl) \quad company \quad (nmod) \quad located \quad (case) \quad CA \quad in \quad)) \dots \\
 \lambda f g x. \exists z. \quad \lambda x. company(x_a) \quad \lambda f g z. \exists x. \quad \lambda x. located(x_e) \quad \lambda f g x. f(x) \quad \lambda x. CA(x_a) \quad \lambda x. empty(x) \\
 f(x) \wedge g(z) \wedge \quad \quad \quad f(z) \wedge g(x) \quad \quad \quad \hline
 arg_2(z_e, x_a) \quad \quad \quad arg_{in}(z_e, x_a) \quad \quad \quad \lambda x. CA(x_a)
 \end{array}$$

lambda expression composition

$$\exists z. company(Pixar) \wedge located(z_e) \wedge arg_2(z_e, Pixar) \wedge arg_{in}(z_e, CA)$$

Grounded Semantic Parsing



Freebase Semantic Parsing



Titanic

1997 · Drama film/Romance · 3h 30m

7.7/10 · [IMDb](#)

88% · [Rotten Tomatoes](#)

James Cameron's "Titanic" is an epic, action-packed romance set against the ill-fated maiden voyage of the R.M.S. Titanic; the pride and joy of the White Star Line and, at the time, the larg... [More](#)

Initial release: November 18, 1997 ([London](#))

Director: [James Cameron](#)

Featured song: [My Heart Will Go On](#)

Cast



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Jack Dawson



[Kate Winslet](#)
Rose DeWitt Bukater



[Billy Zane](#)
Caledon Hockley



[Gloria Stuart](#)
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[Kathy Bates](#)
Molly Brown

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Leonardo DiCaprio starred as Jack in Titanic which was directed by James Cameron.



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Freebase Semantic Parsing

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Parser

KB

$\text{cast}(\text{TITANIC}, \text{DICAPRIO}, \text{JACK}) \wedge$
 $\text{director}(\text{TITANIC}, \text{CAMERON})$



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$\text{cast}(\text{TITANIC}, \text{DICAPRIO}, \text{JACK}) \wedge$
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Execute



TRUE



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Freebase Semantic Parsing

[Berant et al., 2013, Kwiatkowski et al., 2013]

Question

Who is the director of Titanic?

Answer

{James Cameron}

Freebase Semantic Parsing

[Berant et al., 2013, Kwiatkowski et al., 2013]

Question

Who is the director of Titanic?

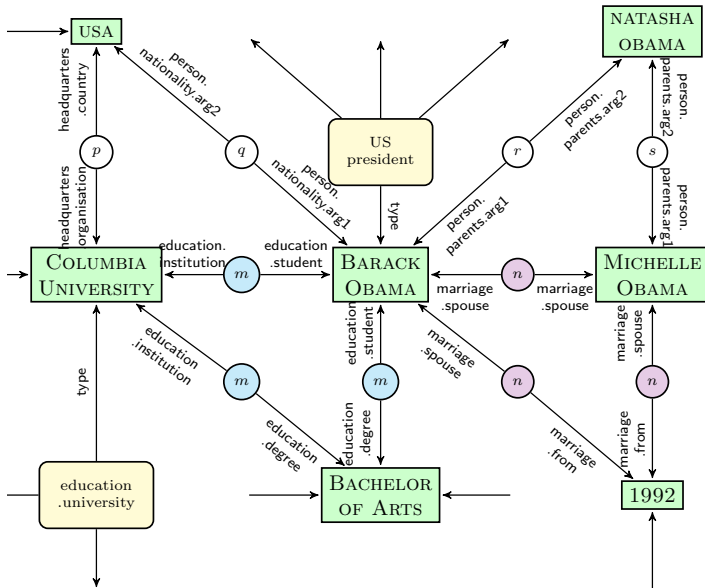
Logical Form

$\lambda x. \text{film.directed_by}(\textit{Titanic}, x)$

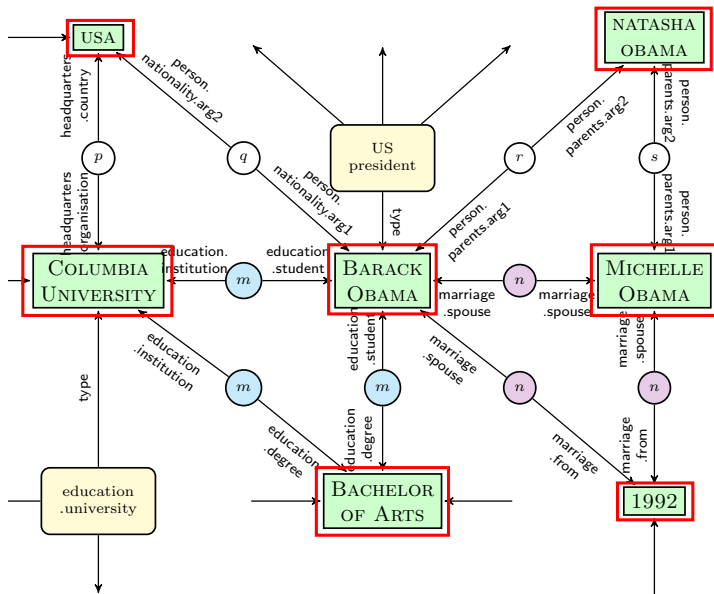
Answer

{James Cameron}

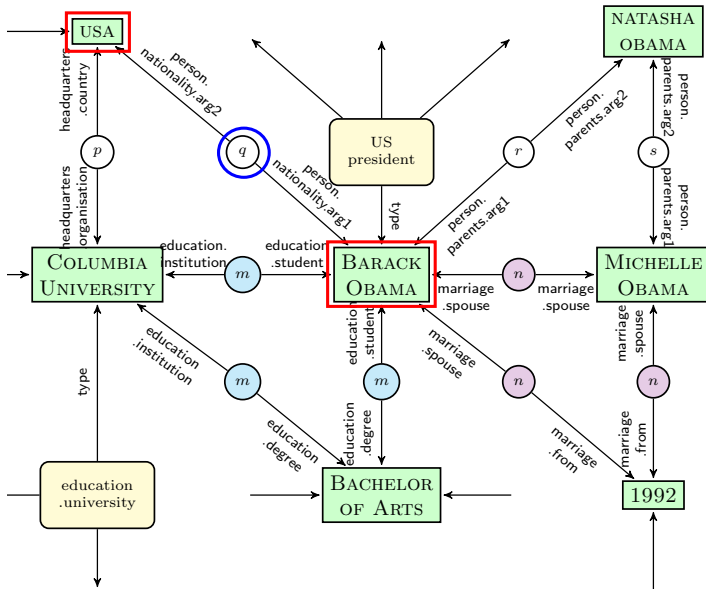
Freebase is a Graph



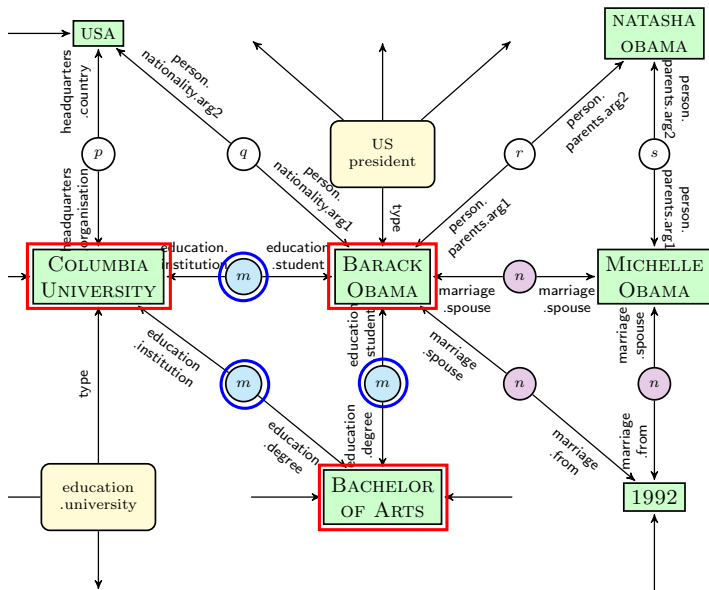
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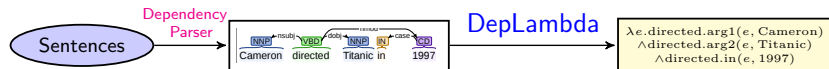
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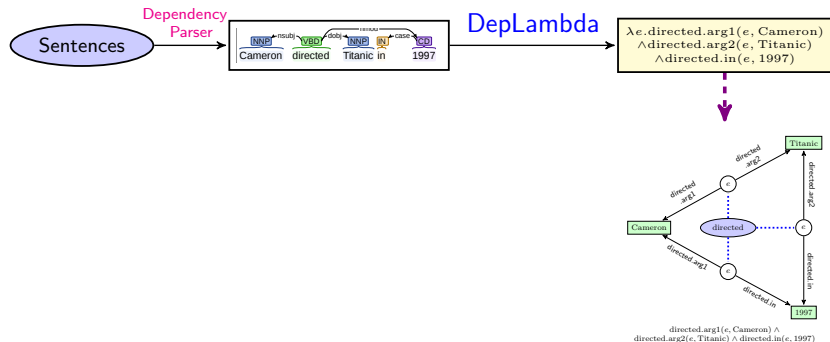
Freebase is a Graph



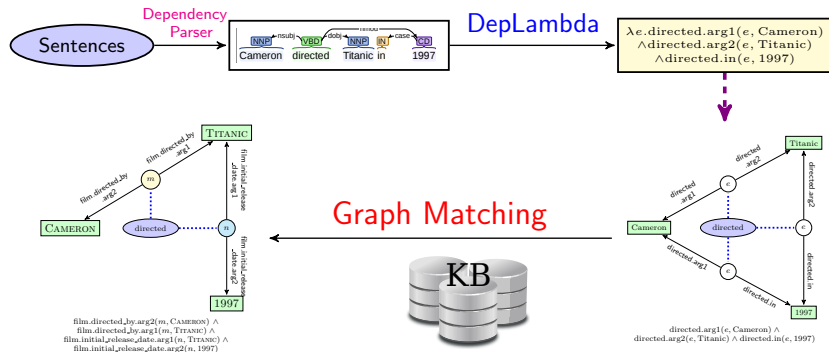
Semantic Parsing as Graph Matching



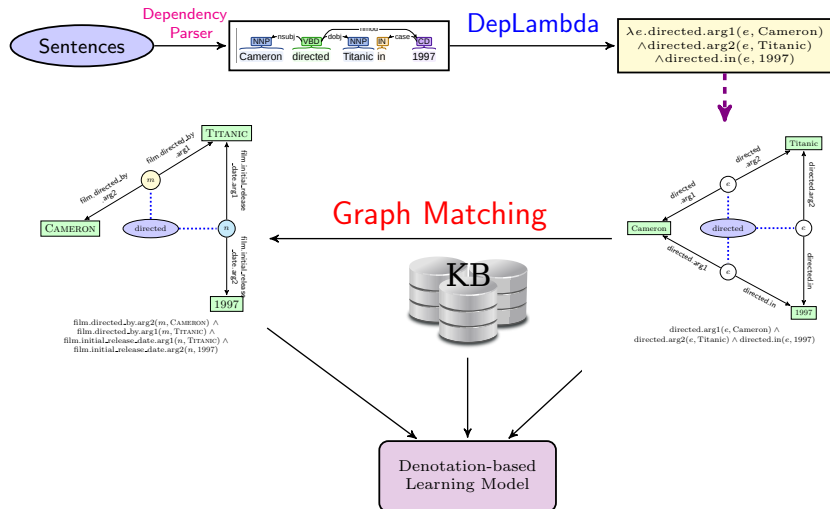
Semantic Parsing as Graph Matching



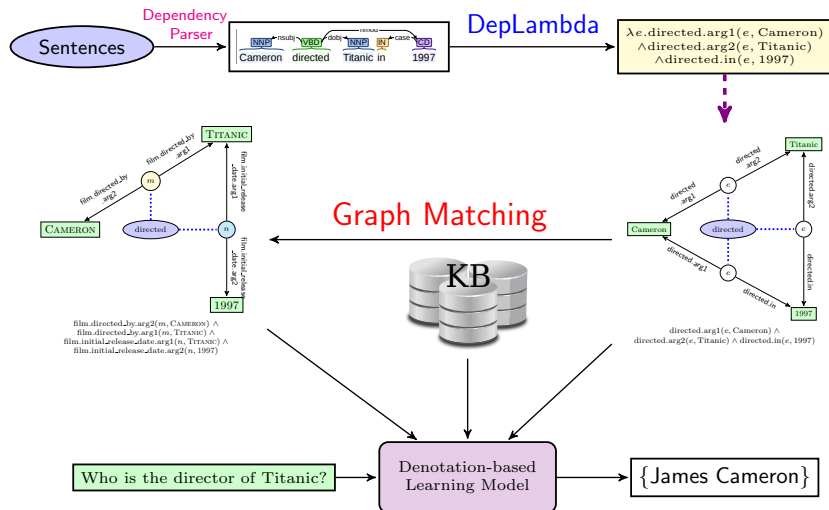
Semantic Parsing as Graph Matching



Semantic Parsing as Graph Matching



Semantic Parsing as Graph Matching



Logical Form to Ungrounded Graph

Cameron directed Titanic in 1997

$\lambda e.\text{directed}.\text{arg1}(e, \text{Cameron}) \wedge \text{directed}.\text{arg2}(e, \text{Titanic}) \wedge$
 $\text{directed}.\text{in}(e, 1997)$

Titanic

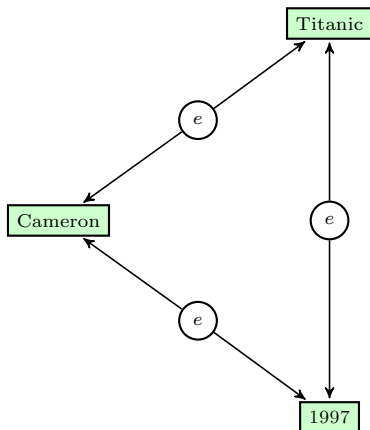
Cameron

1997

Logical Form to Ungrounded Graph

Cameron directed Titanic in 1997

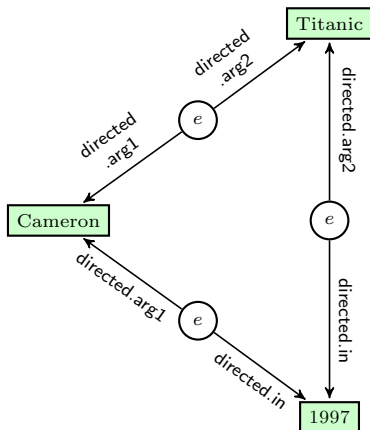
$\lambda e.\text{directed}.\text{arg1}(\mathbf{e}, \text{Cameron}) \wedge \text{directed}.\text{arg2}(\mathbf{e}, \text{Titanic}) \wedge$
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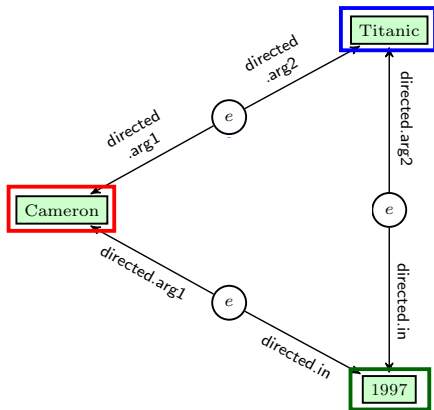
Logical Form to Ungrounded Graph

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$\lambda e.\text{directed.arg1}(e, \text{Cameron}) \wedge \text{directed.arg2}(e, \text{Titanic}) \wedge$
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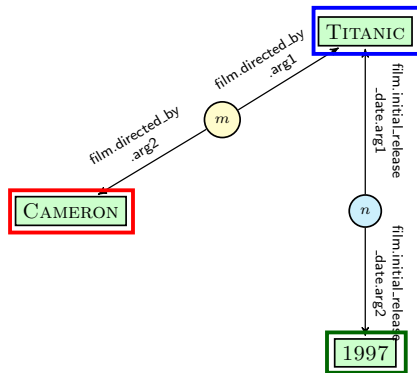


Graph Matching



$\text{directed.arg1}(e, \text{Cameron}) \wedge$
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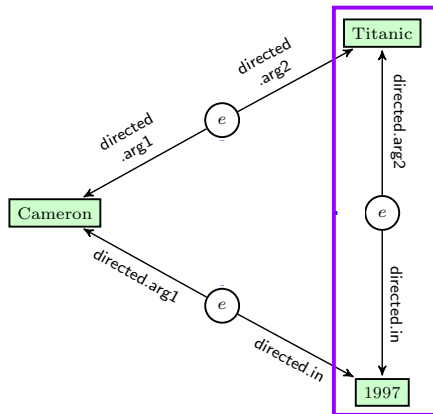
Ungrounded Graph



$\text{film.directed_by.arg2}(m, \text{CAMERON}) \wedge$
 $\text{film.directed_by.arg1}(m, \text{TITANIC}) \wedge$
 $\text{film.initial_release_date.arg1}(n, \text{TITANIC}) \wedge$
 $\text{film.initial_release_date.arg2}(n, 1997)$

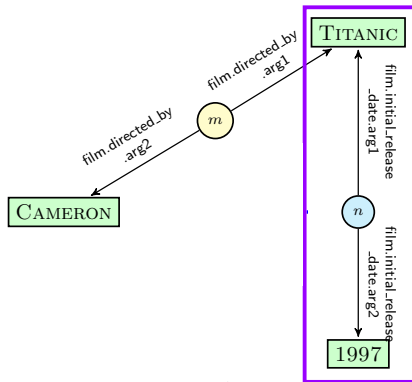
Grounded Graph

Graph Matching



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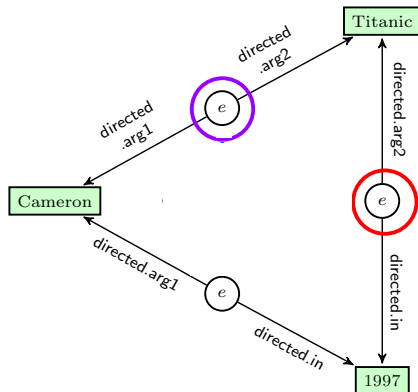
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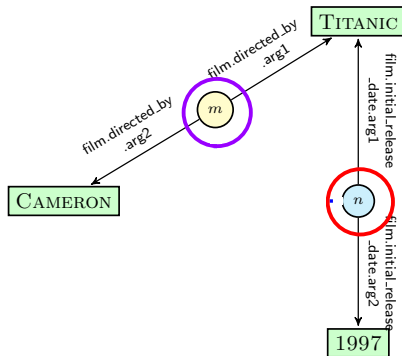
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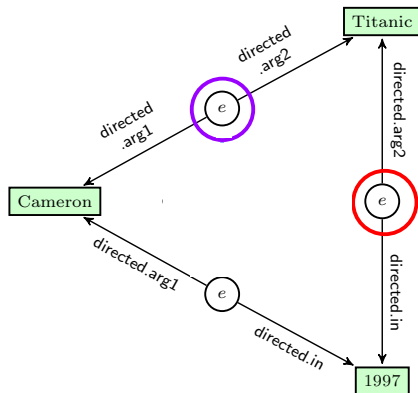
Ungrounded Graph



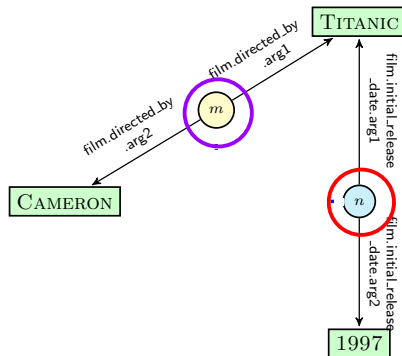
$\text{film.directed_by.arg2}(m, \text{CAMERON}) \wedge$
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Grounded Graph

Graph Matching

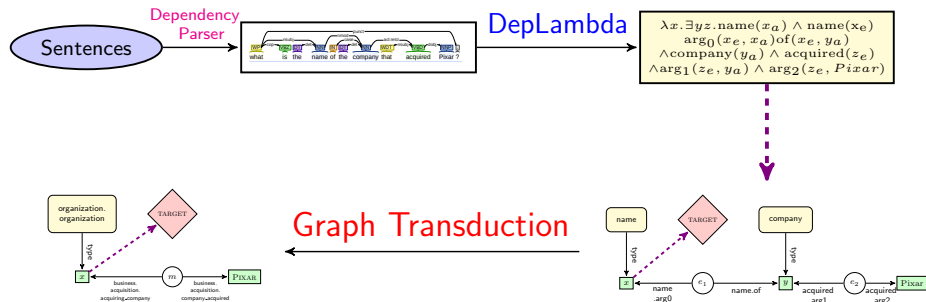


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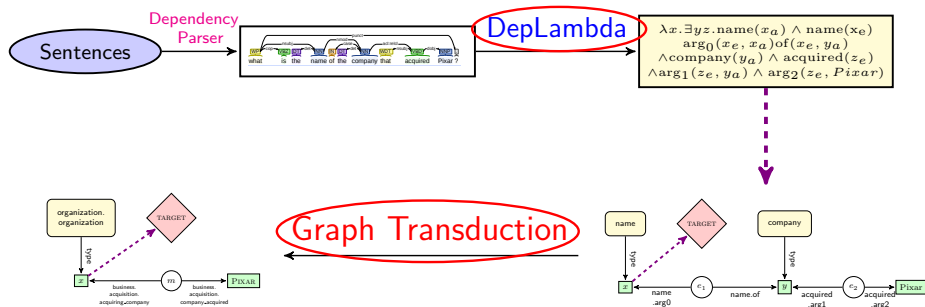


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Freebase Semantic Parsing



Freebase Semantic Parsing



Baselines

SIMPLEGRAPH: All entities connected to a single event

- ▶ Does not handle compositional questions

CCGGGRAPH: CCG logical forms

DEPTREE: Directly transduce a dependency tree to target graph

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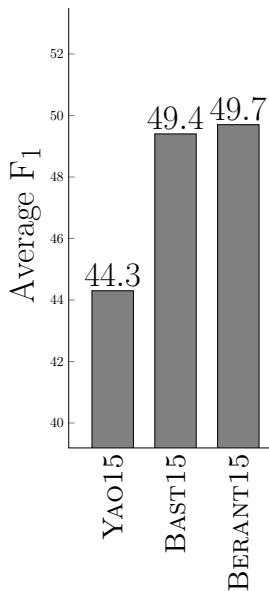
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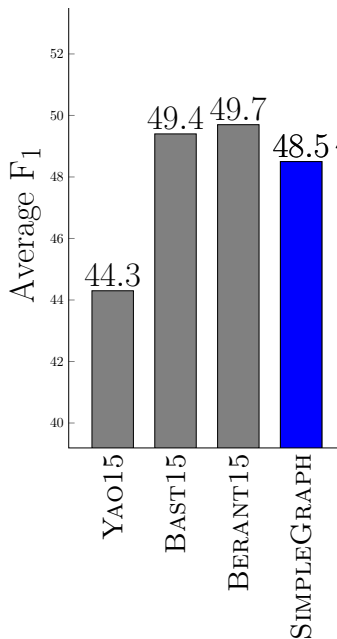
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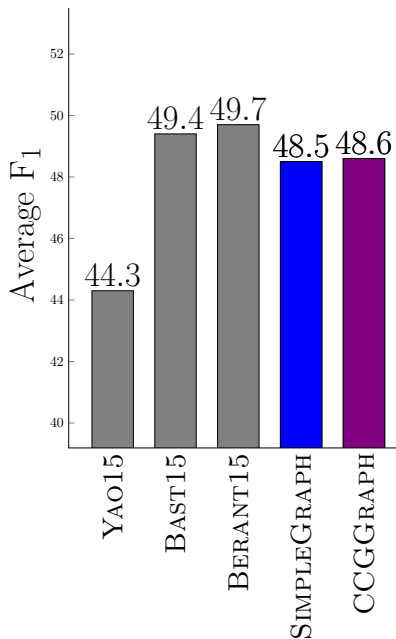
Results on WebQuestions



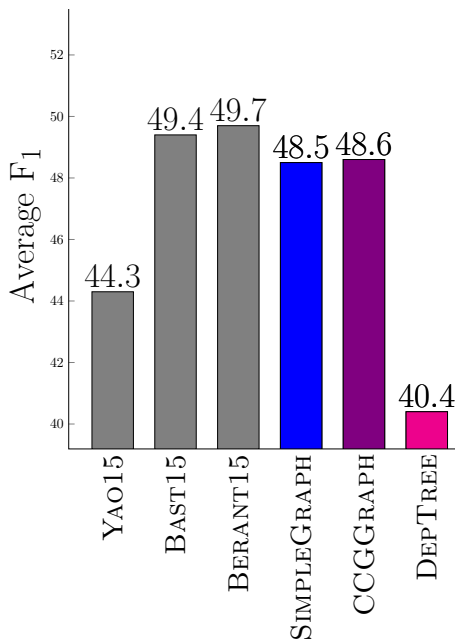
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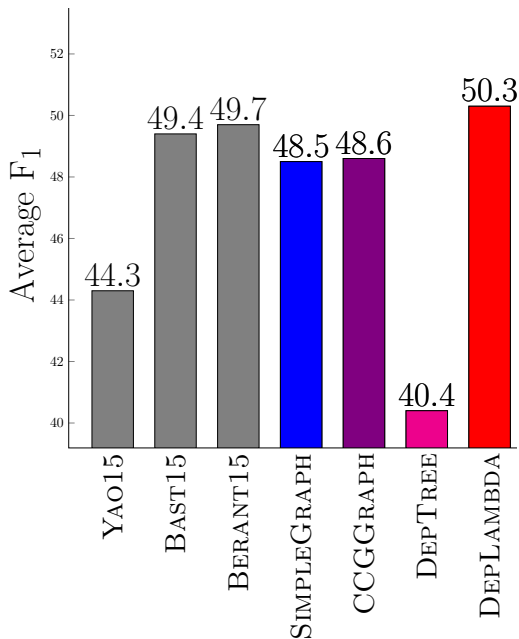
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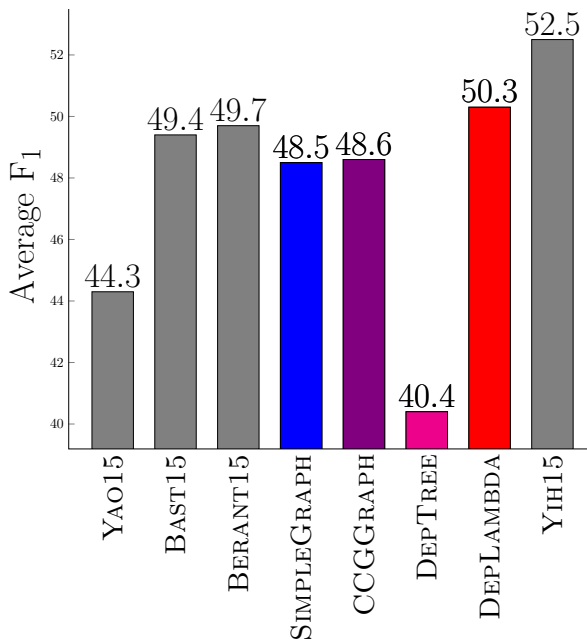
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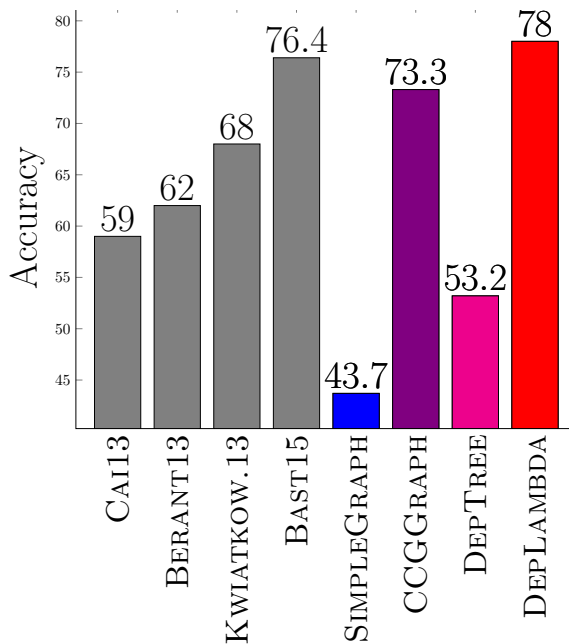
Results on WebQuestions



Results on WebQuestions



Results on Free917



Summary

- ▶ Lambda Calculus for converting Dependencies to Logical Forms
- ▶ Semantic parsing as Graph Transduction
- ▶ State-of-the-art results on Free917 and competitive results on WebQuestions
- ▶ Work in progress: DepLambda for multiple languages

Rules available from

<https://github.com/sivareddy/deplambda>

Thank You!

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Thank You!

Learning Model

Structured Perceptron: Ranks a pair of grounded and ungrounded graph

$$(\hat{g}, \hat{u}) = \arg \max_{g, u} \Phi(g, u, q, KB) \cdot \theta$$

Features: Φ is defined over sentence, grounded and ungrounded graph

Training: Use a surrogate graph (dynamic oracle) to update weights

$$\theta \leftarrow \theta + \Phi(g^+, u^+, q, KB) - \Phi(\hat{g}, \hat{u}, q, KB)$$

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Learning Model

Oracle Graphs: Search for all the grounded graphs reachable via the ungrounded graph.

Pick all the graphs with minimal F_1 -loss against the gold answer.

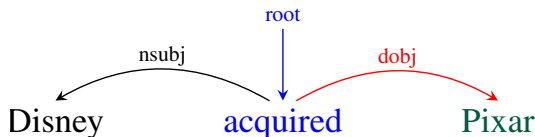
Surrogate Gold Graph:

$$(u^+, g^+) = \arg \max_{(u, g) \in O_{KB, A}(q)} \theta^t \cdot \Phi(u, g, q, KB),$$

Beam Search: Limit the predictions to 100 graph pairs, and choose the best prediction for update.

Dependencies to Logical Forms

Single Type System

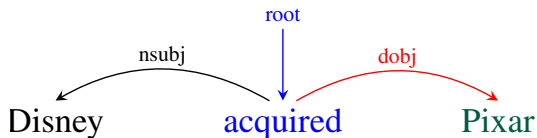


All constituents are of the same lambda expression type

$\text{TYPE}[\text{acquired}] = \text{TYPE}[\text{Pixar}] = \text{TYPE}[(\text{dobj } \text{acquired } \text{Pixar})]$

Dependencies to Logical Forms

Single Type System

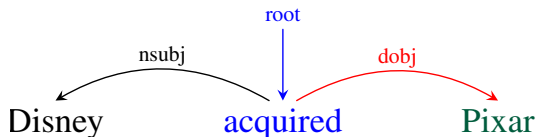


All **words** have a *lambda expression* of type η

- ▶ $\text{TYPE}[\text{acquired}] = \eta$
- ▶ $\text{TYPE}[\text{Pixar}] = \eta$
- ▶ $\text{TYPE}[(\text{dobj acquired Pixar})] = \eta$

Dependencies to Logical Forms

Single Type System

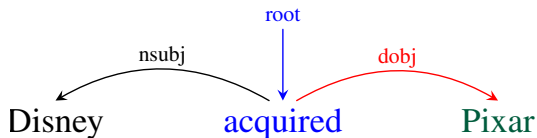


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Dependencies to Logical Forms

Single Type System



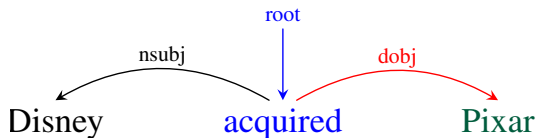
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$\implies \text{TYPE}[\text{dobj}] = \eta \rightarrow \eta \rightarrow \eta$

Dependencies to Logical Forms

Lambda Calculus for Single Type System



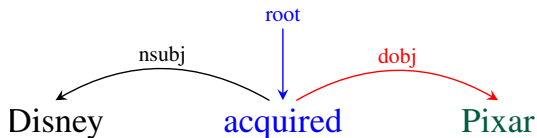
Lambda Expression for words

acquired $\Rightarrow \lambda x_e. \text{acquired}(x_e)$

Pixar $\Rightarrow \lambda x_a. \text{Pixar}(x_a)$

Dependencies to Logical Forms

Lambda Calculus for Single Type System



Lambda Expression for words

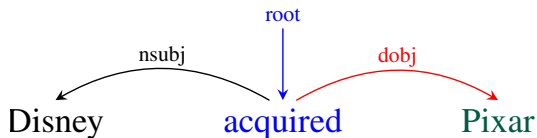
acquired $\Rightarrow \lambda x_e. \text{acquired}(x_e)$ $\Rightarrow \text{TYPE} = \text{Event} \rightarrow \text{Bool}$

Pixar $\Rightarrow \lambda x_a. \text{Pixar}(x_a)$ $\Rightarrow \text{TYPE} = \text{Ind} \rightarrow \text{Bool}$

Here $\text{TYPE}[\text{acquired}] \neq \text{TYPE}[\text{Pixar}]$ ❌

Dependencies to Logical Forms

Lambda Calculus for Single Type System



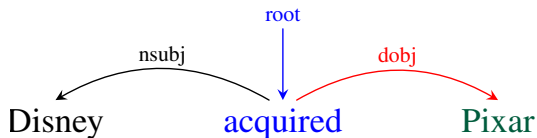
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Dependencies to Logical Forms

Lambda Calculus for Single Type System



Lambda Expression for words

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Pixar $\Rightarrow \lambda x_a \mathbf{x_e}. \text{Pixar}(x_a) \quad \Rightarrow \text{TYPE} = \mathbf{Ind} \times \mathbf{Event} \rightarrow \mathbf{Bool}$

Here $\eta = \text{TYPE}[\text{acquired}] = \text{TYPE}[\text{Pixar}] \checkmark$

Lambda calculus for “*Single Type*” System

More dependency labels

- ▶ (appos Disney the_company)

$$appos = \lambda fgx. f(x) \wedge g(x)$$

This function unifies two nodes

- ▶ (partmod a_company acquired_by_Disney)

$$partmod = \lambda fgx. \exists z. f(x) \wedge g(z) \wedge \arg_1(z_e, x_a)$$

This function reverses the dependency arc direction, but still returns the head

Lambda calculus for “*Single Type*” System

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Lambda calculus for “*Single Type*” System

More dependency labels

- ▶ (conj Disney_and Pixar)

$$conj = \lambda fgz. \exists xy. f(x) \wedge g(y) \wedge \mathbf{coord}(z, x, y)$$

This function creates a struct with two variables

- ▶ (rcmod Disney which_acquired_Pixar)

```
(rcmod Disney  
  (wh-dobj (BIND f (nsubj (dobj acquired f) Pixar))  
            which))
```


Lambda calculus for “*Single Type*” System

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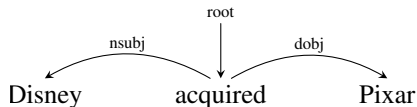
CCG to Logical Forms

Steedman, 2000, 2012; Bos et al., 2004; Lewis & Steedman, 2013; Reddy et al., 2014

Disney	acquired	Pixar
\overline{NP}	$\overline{S \backslash NP / NP}$	\overline{NP}
Disney	$\lambda y \lambda x \lambda e. \text{acquired}(e)$ $\wedge \text{arg}_1(e, x)$ $\wedge \text{arg}_2(e, y)$	Pixar
	$\xrightarrow{\hspace{10em}}$	
	$S \backslash NP$	
	$\lambda x \lambda e. \text{acquired}(e)$ $\wedge \text{arg}_1(e, x) \wedge \text{arg}_2(e, \text{Pixar})$	
	$\xleftarrow{\hspace{10em}}$	
	S	
	$\lambda e. \text{acquired}(e) \wedge \text{arg}_1(e, \text{Disney}) \wedge \text{arg}_2(e, \text{Pixar})$	

Dependencies to Logical Forms

Challenges



The obvious idea

if proper noun then

assign $\lambda x.\text{word}(x)$

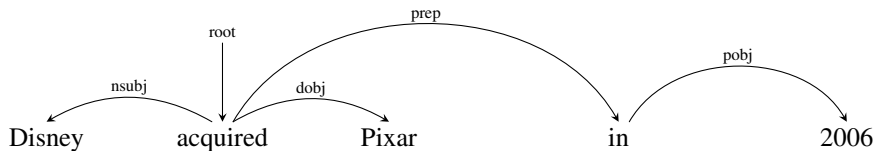
else if verb with subject and object then

assign $\lambda fge.\exists xy.f(x) \wedge g(y) \wedge \text{word}(e) \wedge \text{arg}_1(e,x) \wedge \text{arg}_2(e,y)$

end if

Dependencies to Logical Forms

Challenges



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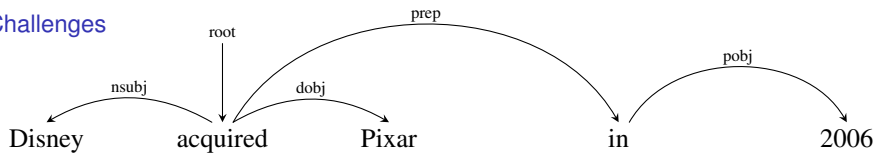
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end if

But, what about?

Dependencies to Logical Forms

Challenges



Problems

1. Rules \propto dependency label permutations
2. Complex lexical semantics
3. Highly sensitive to parse errors
4. Prone to type collisions

Comparison with CCG

Handling of control verbs is painful.

Sentence:

John persuaded Jim to acquire Pixar.

Binarized Tree:

(nsubj (xcomp (dobj persuaded Jim) to_acquire_Pixar) John)

Elegant handling in CCG

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Conjunctions

Sentence:

Eminem signed to Interscope and discovered 50 Cent.

Binarized tree:

(nsubj (conj-vp (cc s_to_I and) d_50) Eminem)

Substitution:

conj-vp $\Rightarrow \lambda fgx. \exists yz. f(y) \wedge g(z) \wedge \text{coord}(x, y, z)$

Logical Expression:

$\lambda w. \exists xyz. \text{Eminem}(x_a) \wedge \text{coord}(w, y, z)$
 $\wedge \text{arg}_1(w_e, x_a) \wedge \text{s_to_I}(y) \wedge \text{d_50}(z)$

Post processing:

$\lambda e. \exists xyz. \text{Eminem}(x_a) \wedge \text{arg}_1(y_e, x_a)$
 $\wedge \text{arg}_1(z_e, x_a) \wedge \text{s_to_I}(y) \wedge \text{d_50}(z)$

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Relative Clause

following Moortgat (1988); Pereira (1990); Carpenter (1998)

Sentence:

Apple which Jobs founded

Binarized tree:

(rcmod Apple
 (wh-dobj (BIND f (nsubj (dobj founded f) Jobs))
 which))

Substitution:

wh-dobj $\Rightarrow \lambda fgz.f(z)$

rcmod $\Rightarrow \lambda fgz.f(z) \wedge g(z)$

Logical Expression:

$\lambda u. \exists xy. \text{founded}(x_e) \wedge \text{Jobs}(y_a)$
 $\wedge \text{arg}_1(x_e, y_a) \wedge \text{arg}_2(x_e, u_a) \wedge \text{Apple}(u_a)$

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Expressivity

How isomorphic are the representations compared to Knowledge Graph?

Average Oracle F_1 Table.

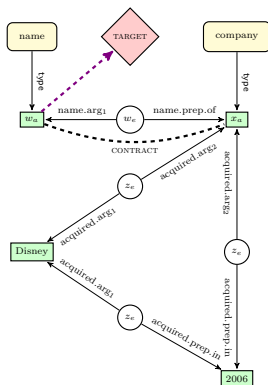
Search Space

How many ways to reach an answer?

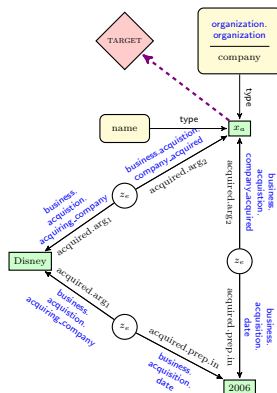
Average Oracle F_1 Table.

Graph Transformation: CONTRACT operation

What is the **name of the company** which Disney acquired in 2006?



Ungrounded graph



Grounded graph

Graph Mismatch: EXPAND operation

What to do Washington DC December?

Before EXPAND

- ▶ $\lambda z. \exists xyw. \text{TARGET}(x_a) \wedge \text{do}(z_e) \wedge \text{arg}_1(z_e, x_a) \wedge \text{Washington_DC}(y_a) \wedge \text{December}(w_a)$

After EXPAND

- ▶ $\lambda z. \exists xyw. \text{TARGET}(x_a) \wedge \text{do}(z_e) \wedge \text{arg}_1(z_e, x_a) \wedge \text{Washington_DC}(y_a) \wedge \text{dep}(z_e, y_a) \wedge \text{December}(w_a) \wedge \text{dep}(z_e, w_a)$