Weakly Supervised Learning of Semantic Parsers for Mapping Instructions to Actions

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TACLVOL. I (2013):49-62 / NAACL 2013



Semantic Parsing

Show me all papers about semantic parsing

$\lambda x.paper(x) \wedge topic(x, SEMPAR)$



Answers Demonstrations Situated examples

More Domains

Databases Large knowledge-bases Instructions Referring expressions Regular expressions

Later this session

Semantic Parsing

Show me all papers about semantic parsing



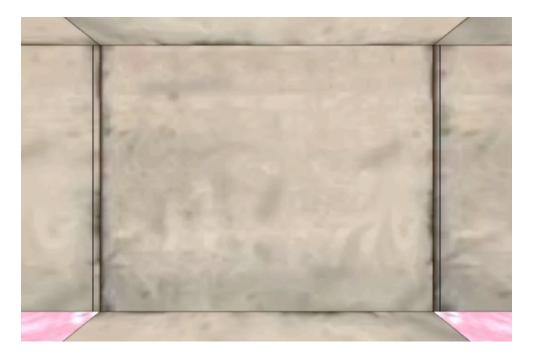


Executing Navigation Instructions

place your back against the wall of the t intersection

turn left

go forward along the pink flowered carpet hall two segments to the intersection with the brick hall



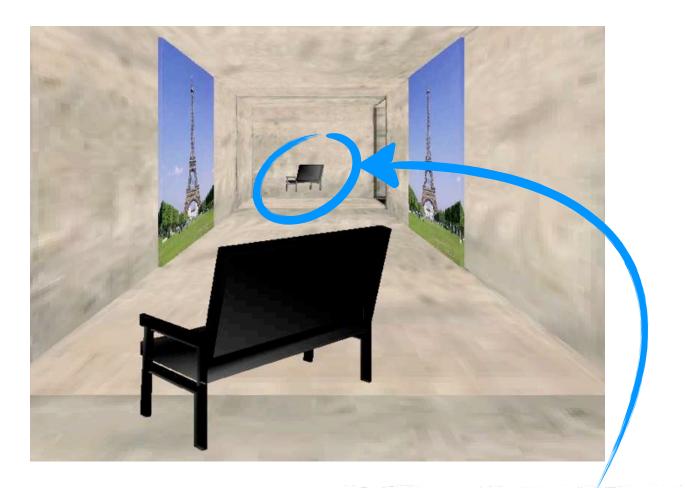
•

Resolve Referents



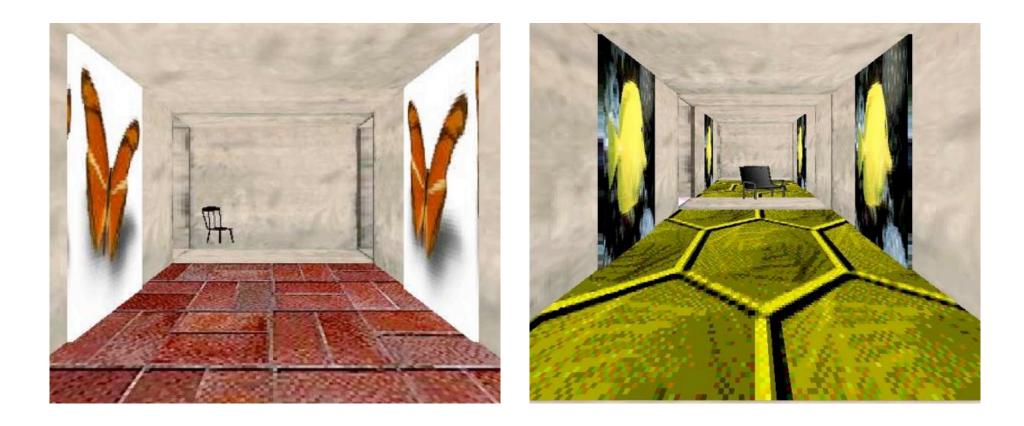


Resolve Referents



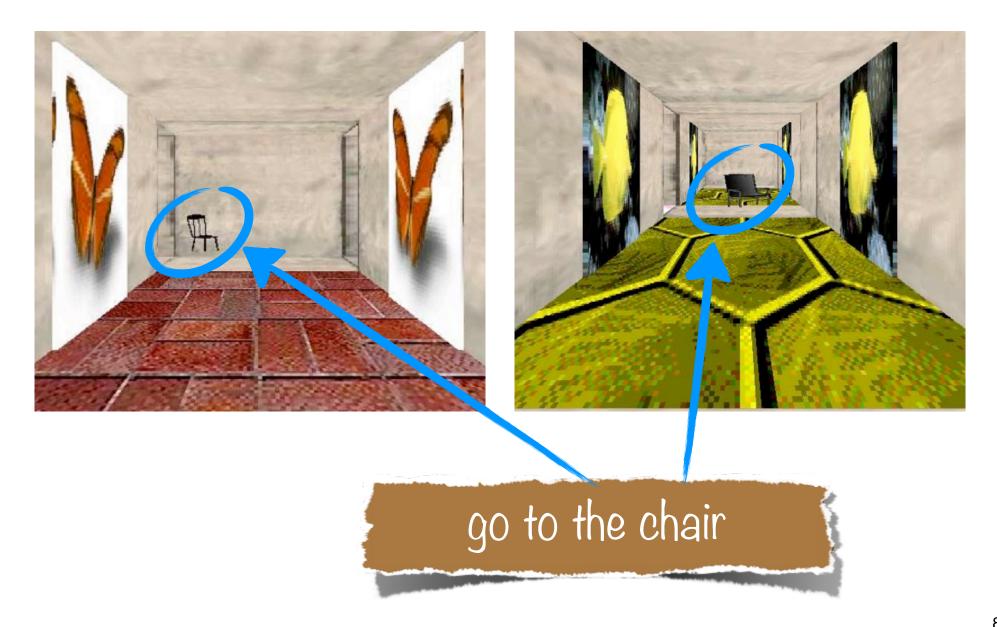
go to the next sofa

Disambiguate Word Sense

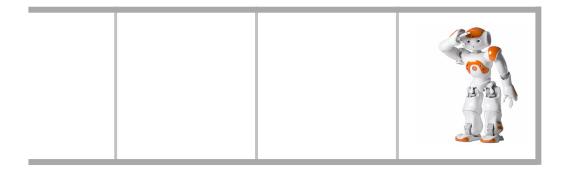




Disambiguate Word Sense

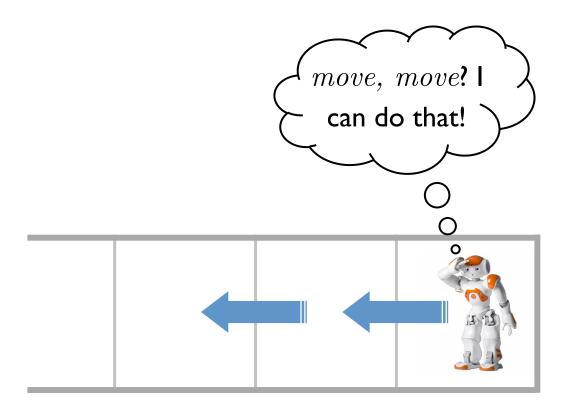


Identify Executable Actions



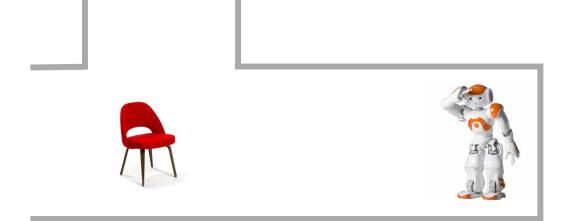


Identify Executable Actions



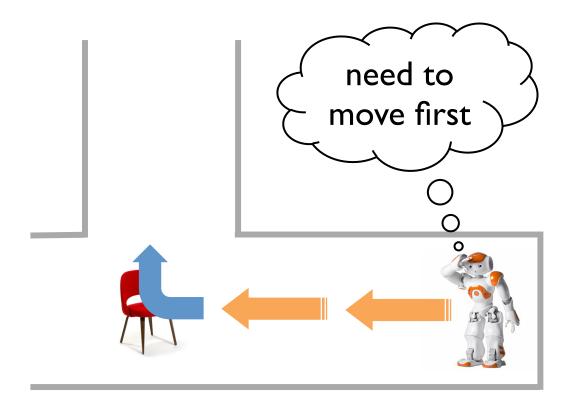


Understand Implicit Requests



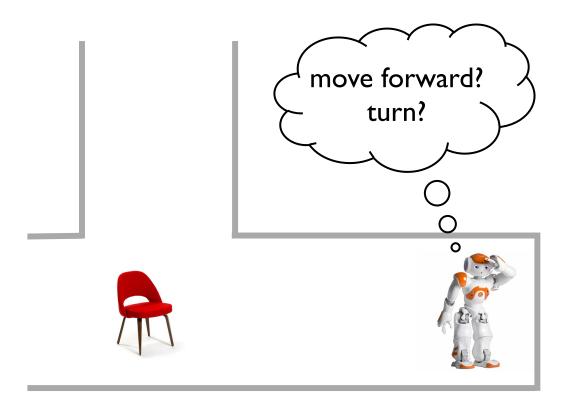


Understand Implicit Requests



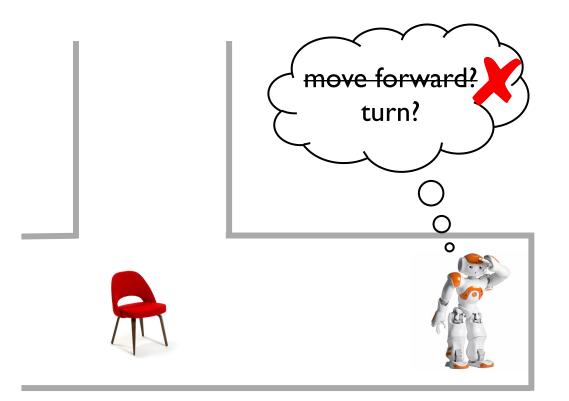


Grounded Learning





Grounded Learning





Learning Signal

Instruction:

at the chair, move forward three steps past the sofa

Demonstration:



Learning Signal

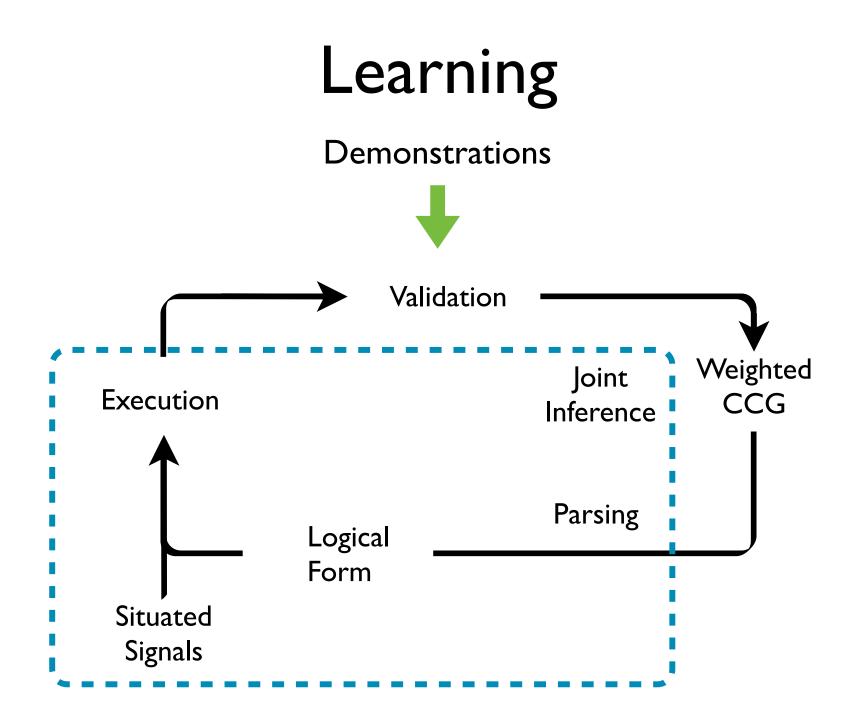
Instruction:

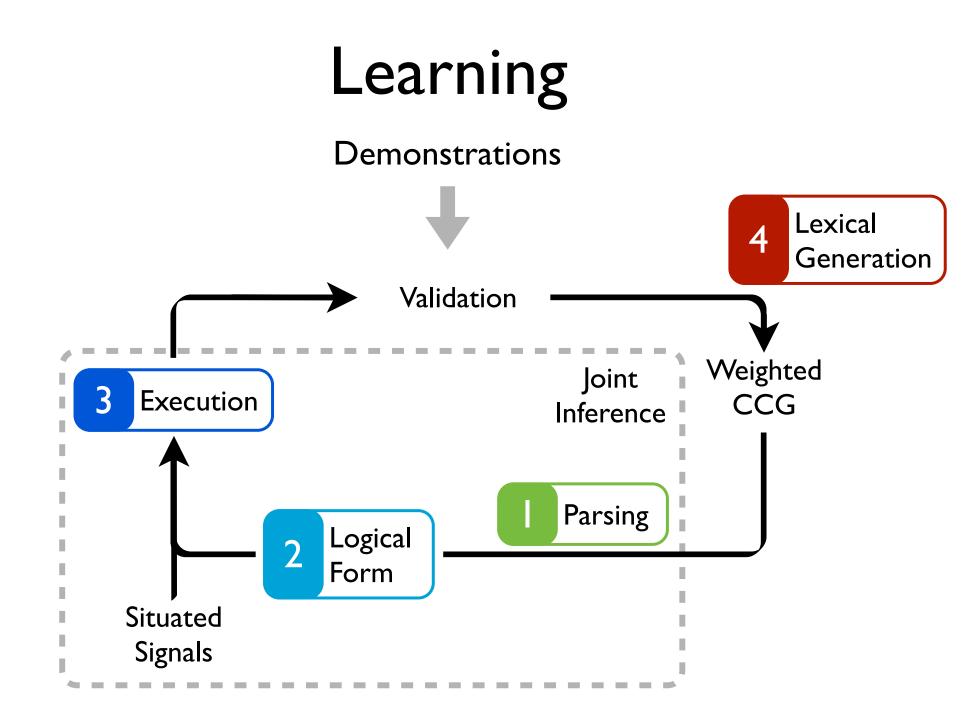
at the chair, move forward three steps past the sofa

Demonstration:

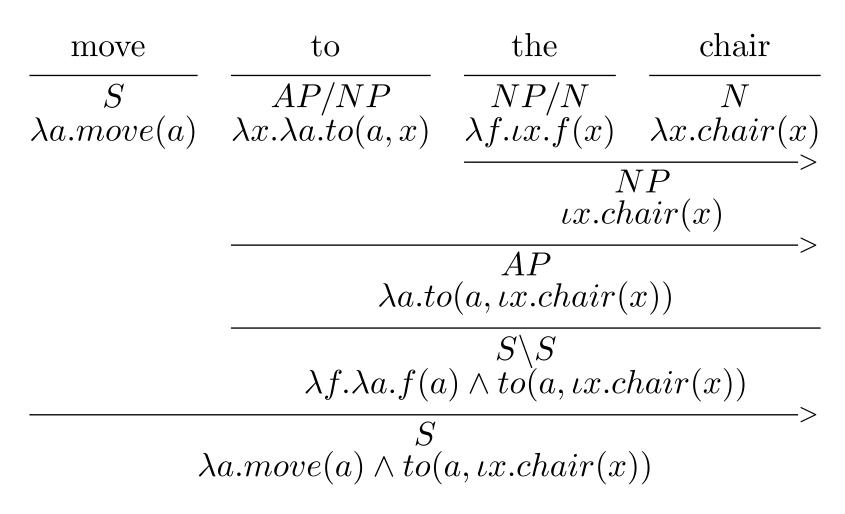


During learning: validate executions of different interpretations against demonstrations

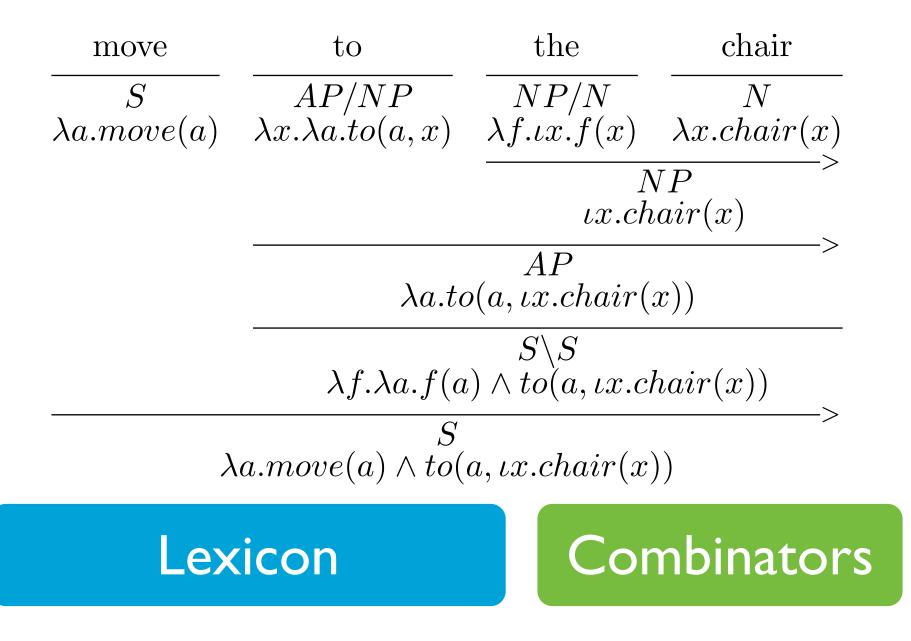




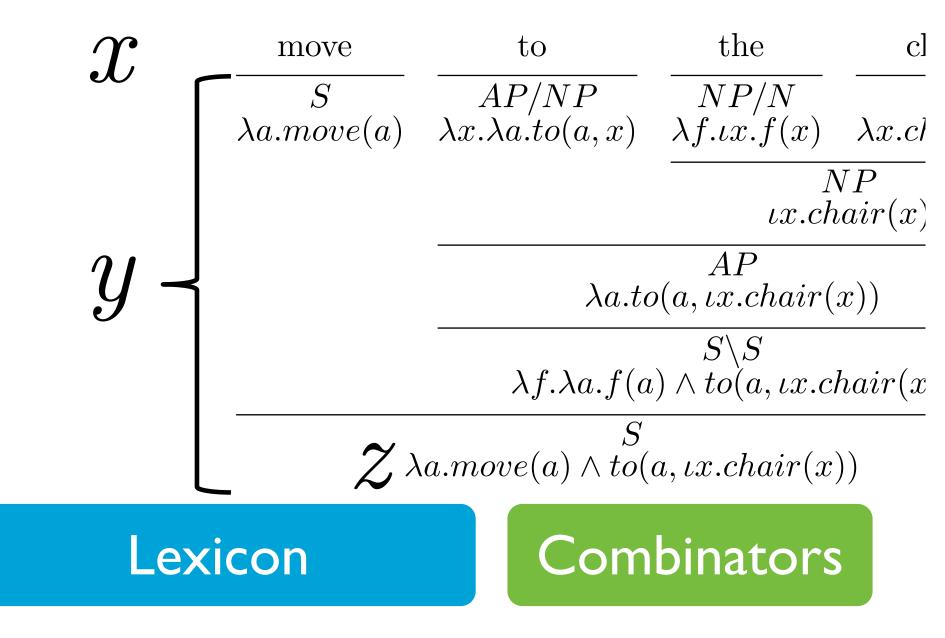
Combinatory Categorial Grammars



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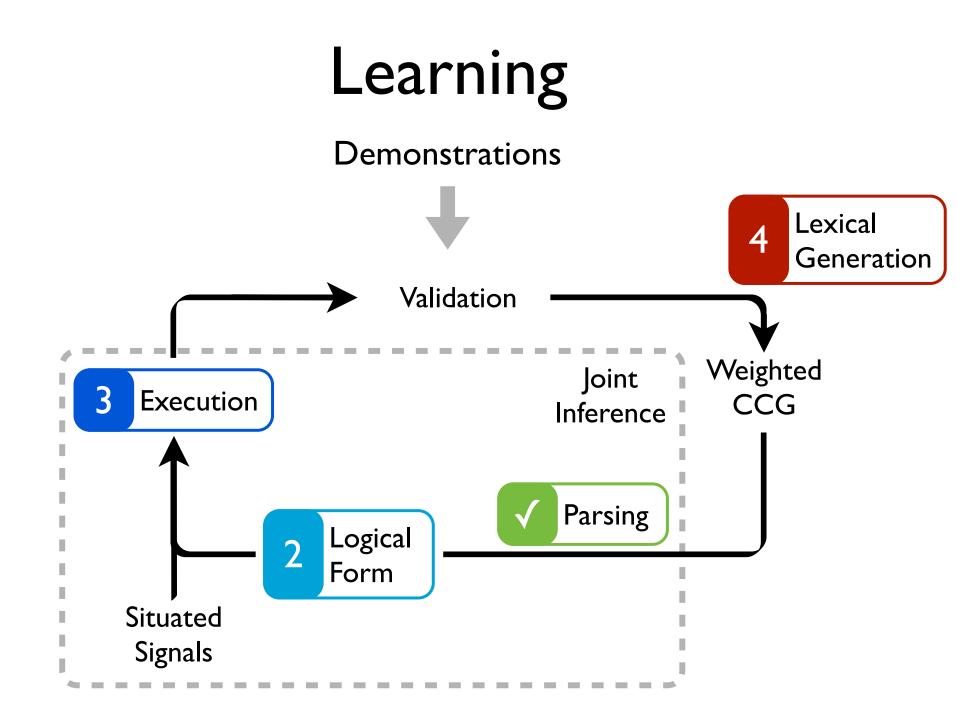


Weighted Linear CCGs

- Given a weighted linear model:
 - CCG lexicon Λ
 - Feature function $f: X \times Y \to \mathbb{R}^m$
 - Weights $w \in \mathbb{R}^m$
- The best parse is:

$$y^* = \operatorname*{arg\,max}_{v} w \cdot f(x, y)$$

• We consider all possible parses y for sentence x given the lexicon Λ



Nouns	Sets of entities
PPs and adjectives	Constrain sets
Noun phrases	Specific entities
Verbs	Relations between entities

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Works well for natural language interfaces for DBs

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Noun phrases	Specific entities
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Previous work on instructional language adopted procedural representation

[Matuszek et al. 2010; 2012; Chen, Mooney 2011; Chen 2012; Kim, Mooney 2012]

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Previous work on instructional language adopted procedural representation

How can we use this approach for instructions?

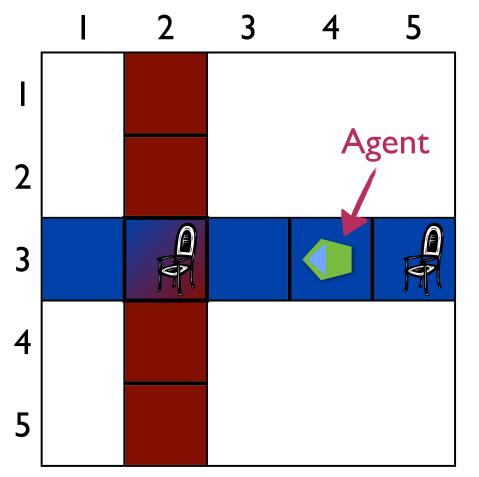
Name objects

Nouns	Sets of entities
PPs and adjectives	Constrain sets
Noun phrases	Specific entities

Instructions to execute

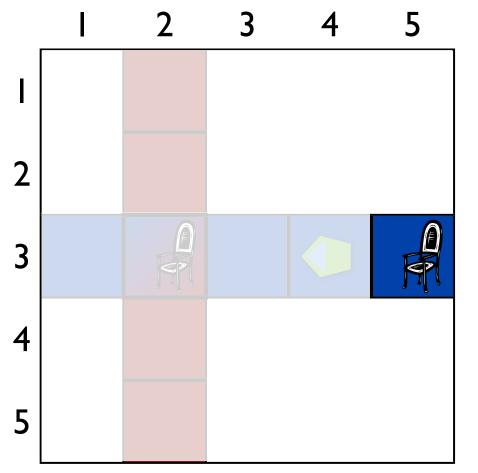
Verbs	Davidsonian Events
Imperatives	Sets of events

Spatial Environment Modeling



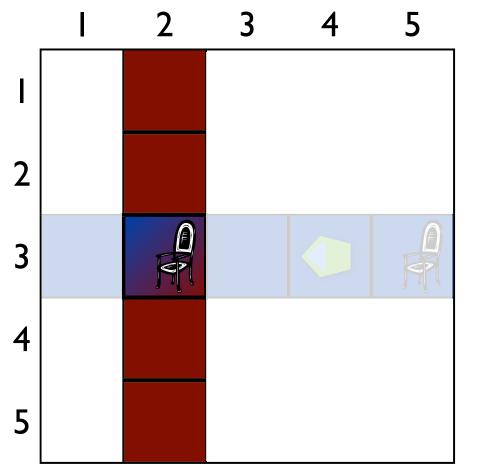
- Maps are graphs of connected positions
- Agent can move forward, turn right and turn left
- Agent perceives clusters of positions
- Clusters capture objects

Spatial Environment Modeling

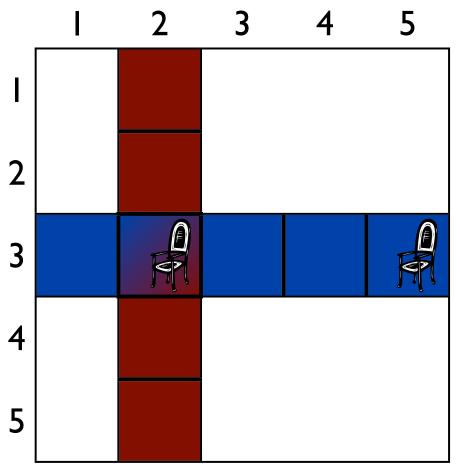


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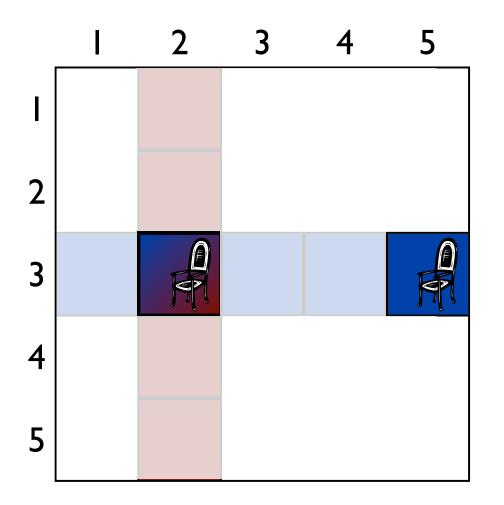
Spatial Environment Modeling



- Maps are graphs of connected positions
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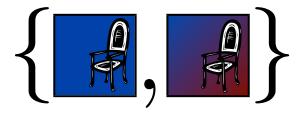
- Nouns
- Noun phrases
- Adjectives
- Prepositional phrases
- Spatial relations

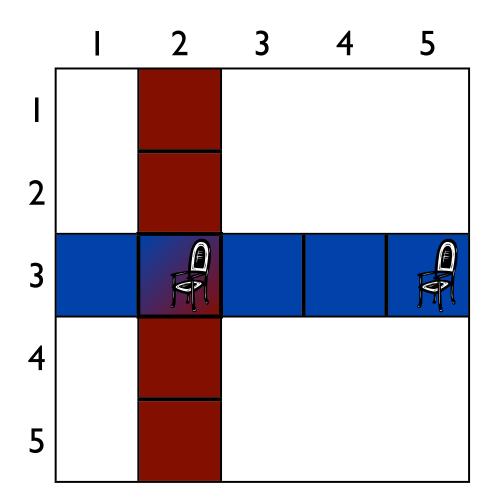


Nouns denote sets of objects

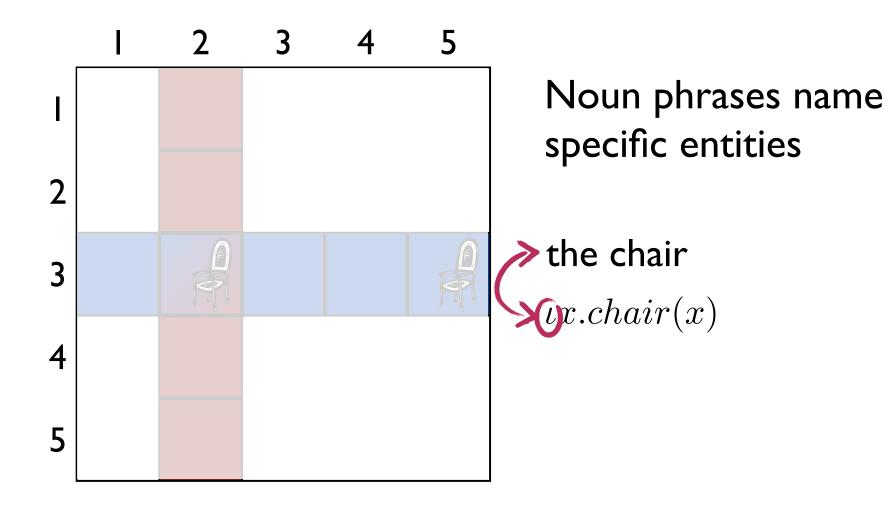
chair

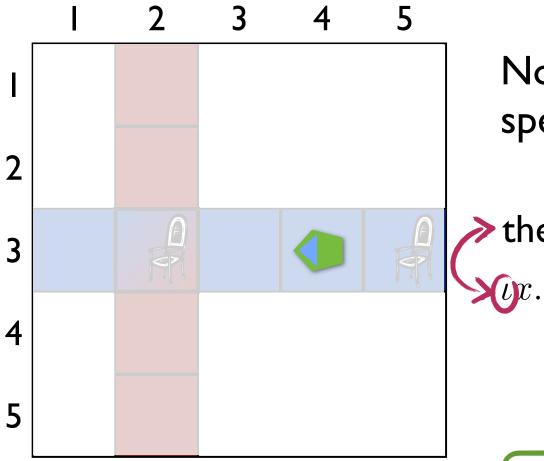
 $\lambda x.chair(x)$





Noun phrases name specific entities



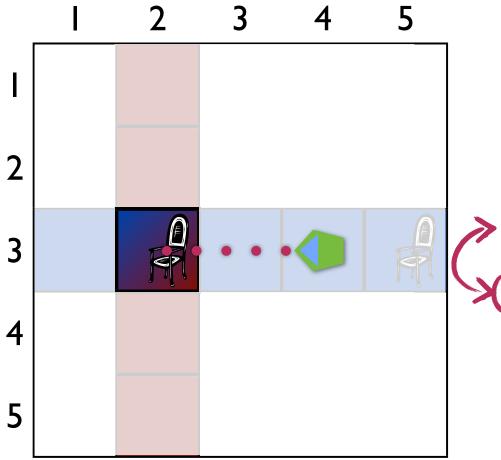


Noun phrases name specific entities

the chair x.chair(x)

> Definite determiner depends on agent state

Spatial Language



Noun phrases name specific entities

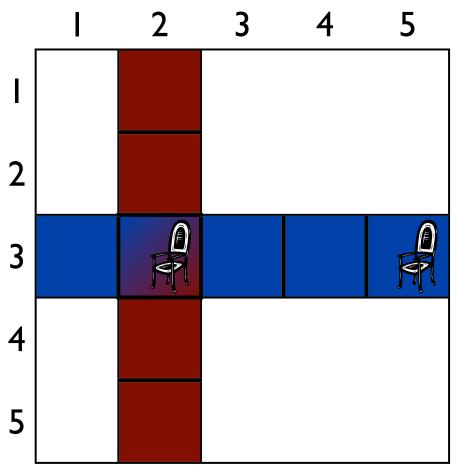
>the chair





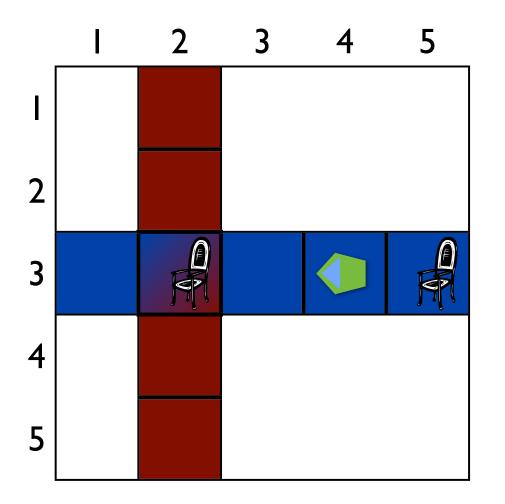
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Spatial Language



- Nouns
- Noun phrases
- Adjectives
- Prepositional phrases
- Spatial relations

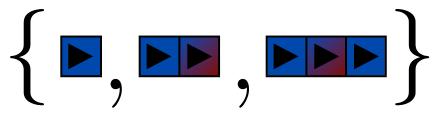
- Sequences of identical actions are events
- Use Neo-Davidsonian event semantics
- Represent imperatives as sets of events

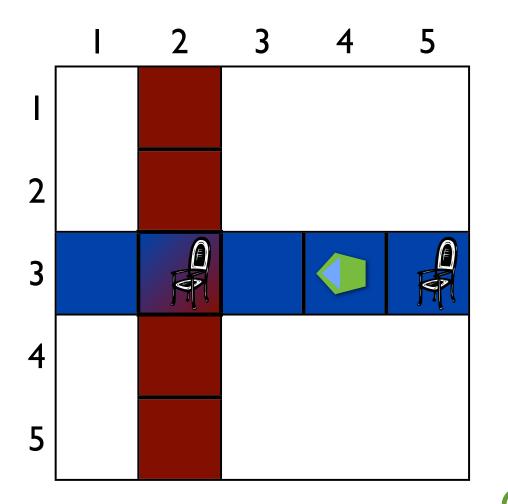


Imperatives are sets of events

move

 $\lambda a.move(a)$

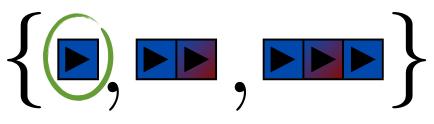




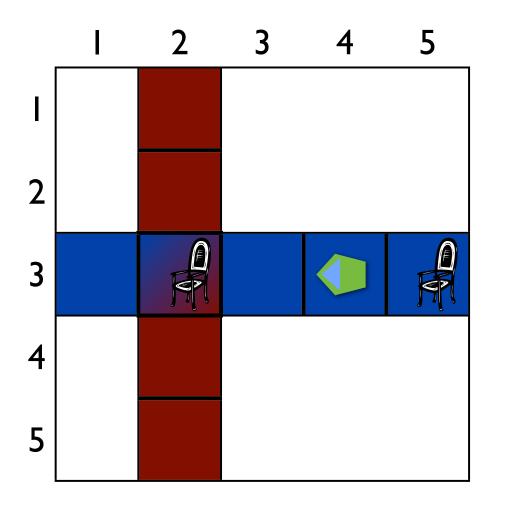
Imperatives are sets of events

move

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Disambiguate by preferring shorter sequences

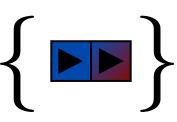


Events can be modified by adverbials

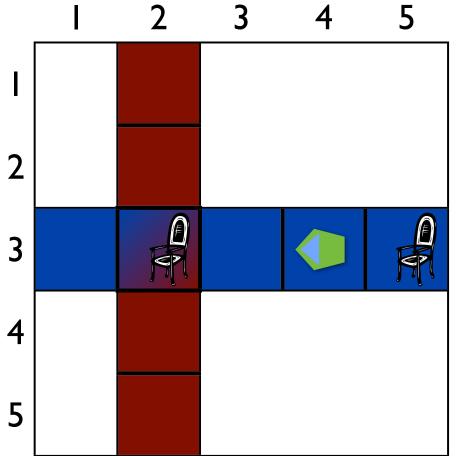
go to the chair

 $\lambda a.move(a) \land$

 $to(a, \iota x.chair(x))$

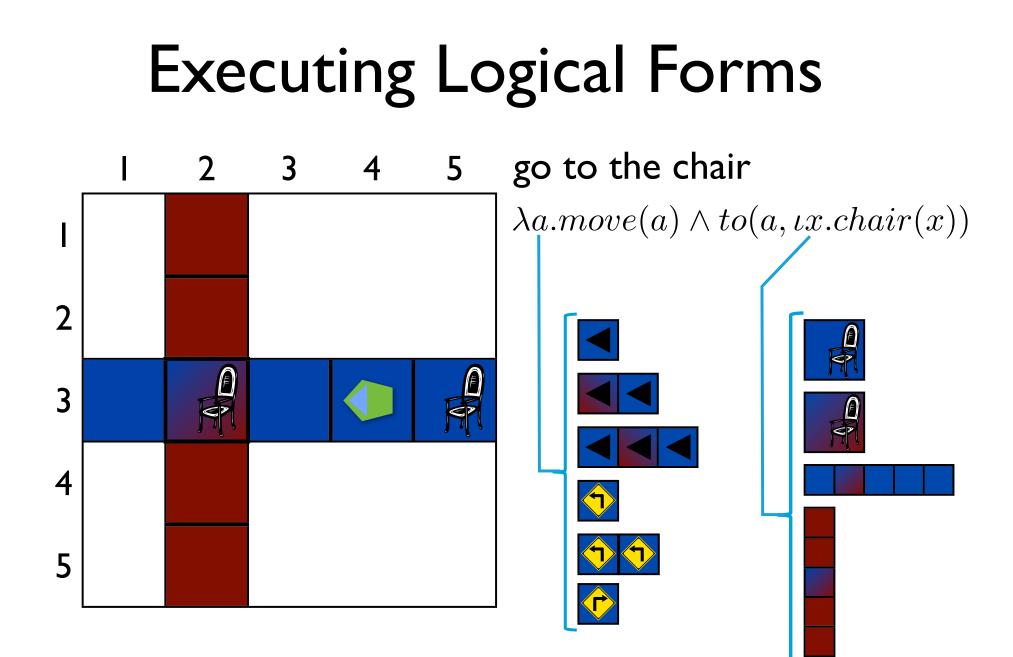


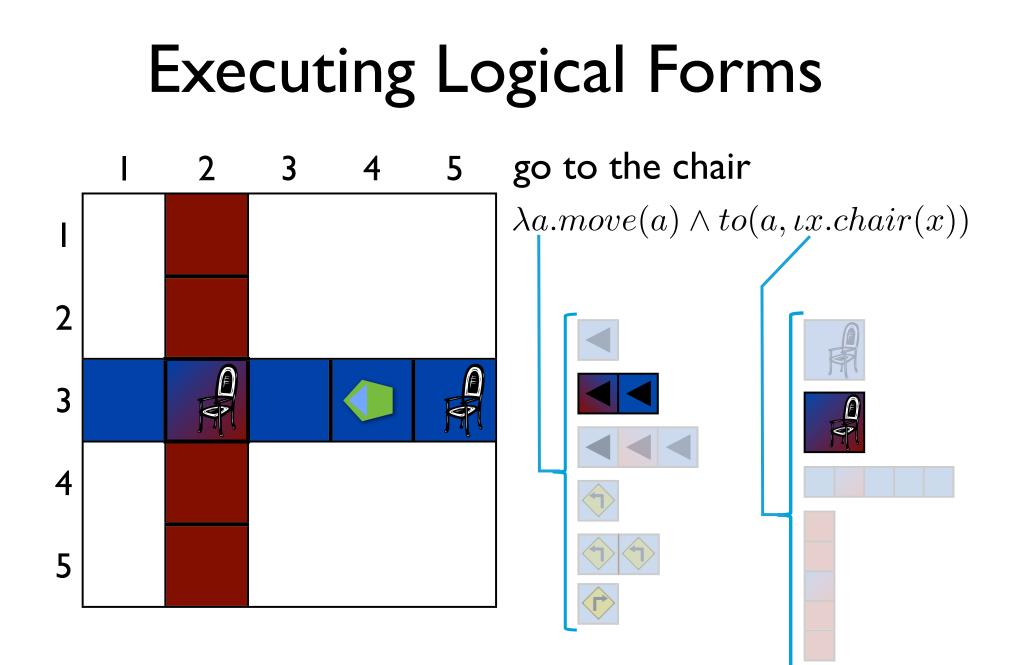
Executing Logical Forms

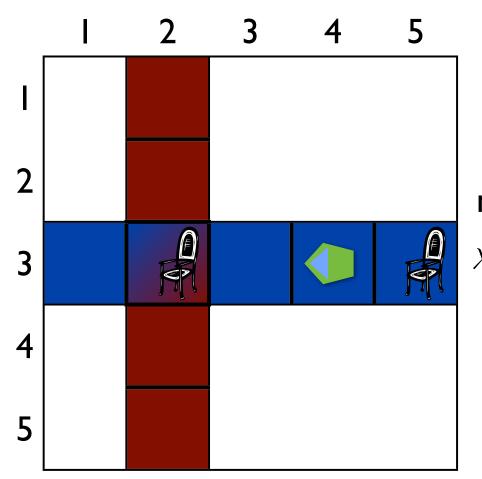


go to the chair $\lambda a.move(a) \wedge to(a, \iota x.chair(x))$

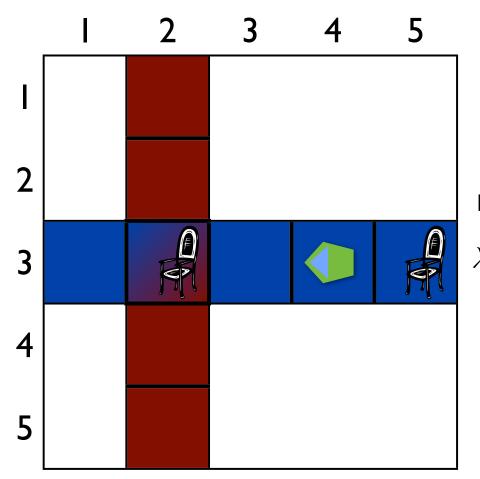
Consider all variable assignments to find satisfying ones



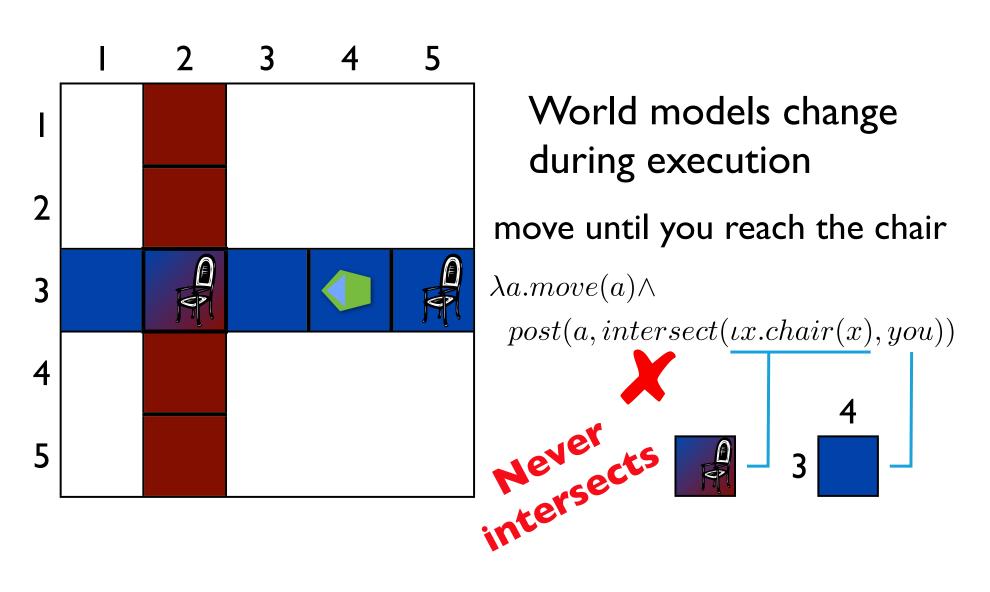


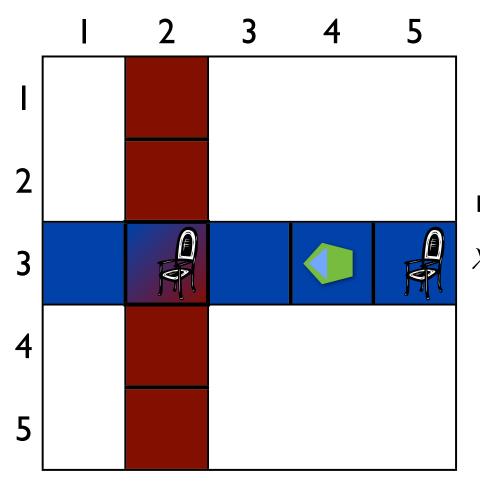


World models change during execution move until you reach the chair $\lambda a.move(a) \wedge$ $post(a, intersect(\iota x.chair(x), you))$



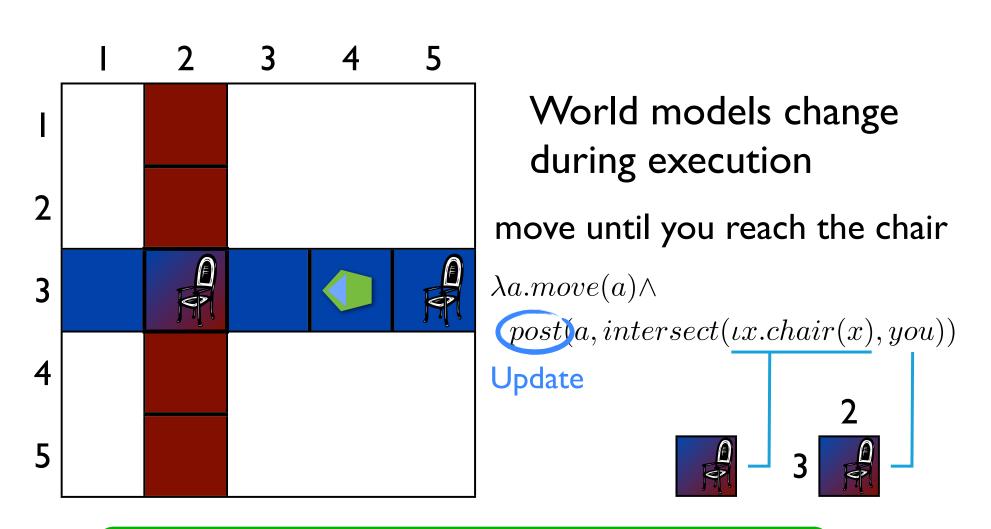
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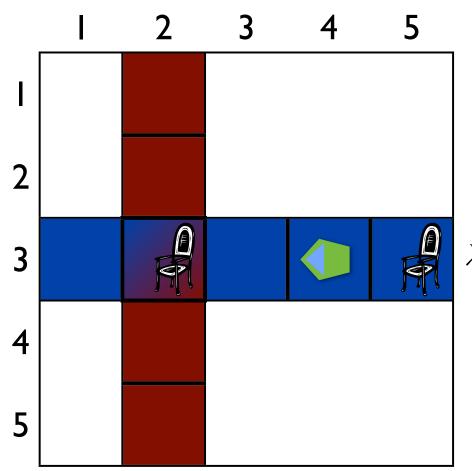
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Update model to reflect state change



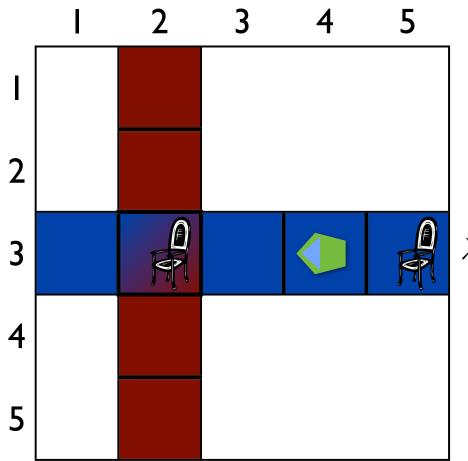
Update model to reflect state change

Implicit Actions



Consider actions with prefixed implicit actions at the chair, turn left $\lambda a.turn(a) \wedge dir(a, left) \wedge$ $pre(a, intersect(\iota x.chair(x), you))$

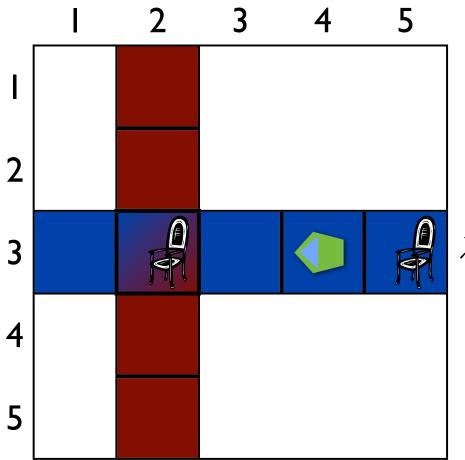
Implicit Actions



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Implicit Actions

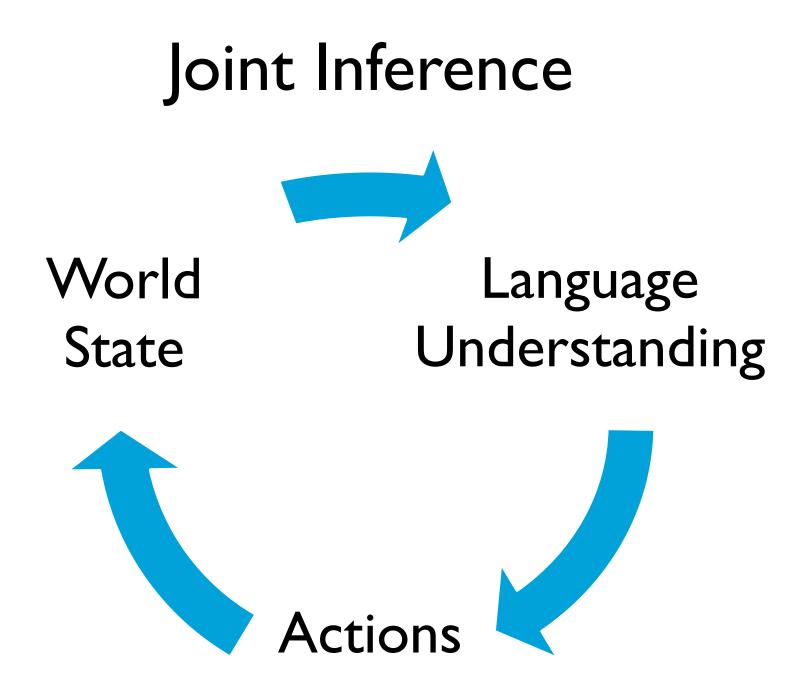


Consider actions with prefixed implicit actions

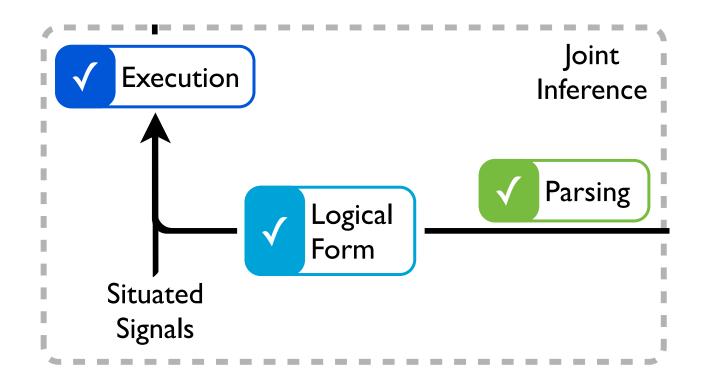
at the chair, turn left

 $\begin{array}{c} \lambda a.turn(a) \wedge dir(a, left) \wedge \\ pre(a, intersect(\iota x.chair(x), you)) \end{array} \end{array}$

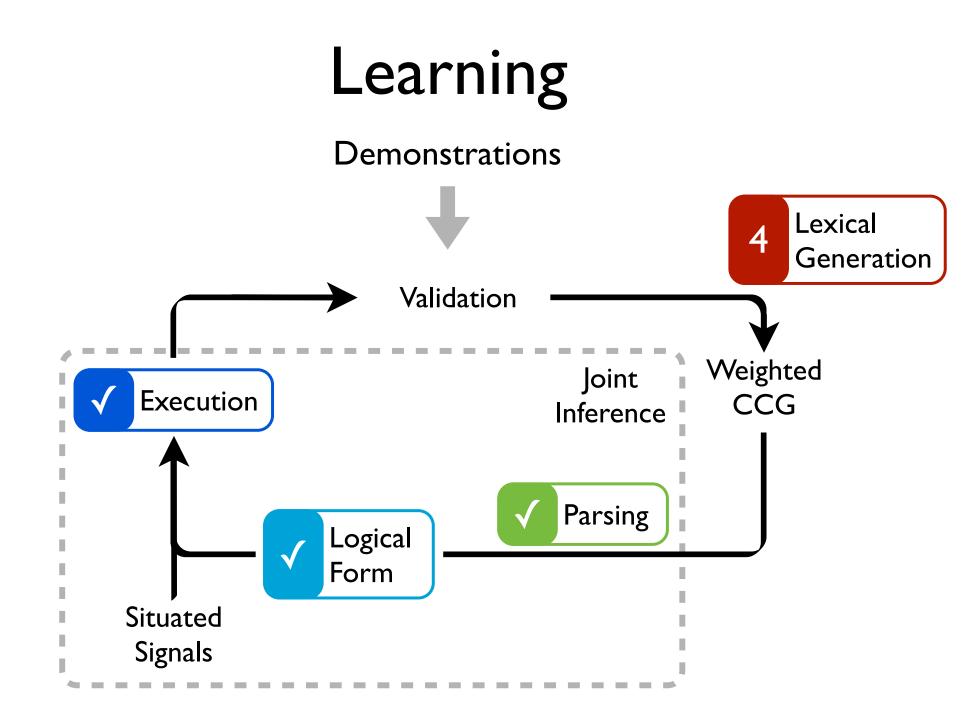




Learning

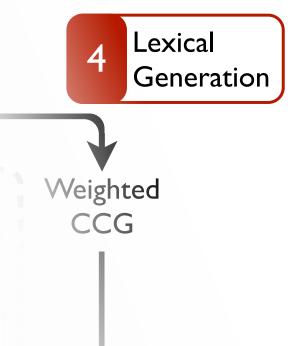


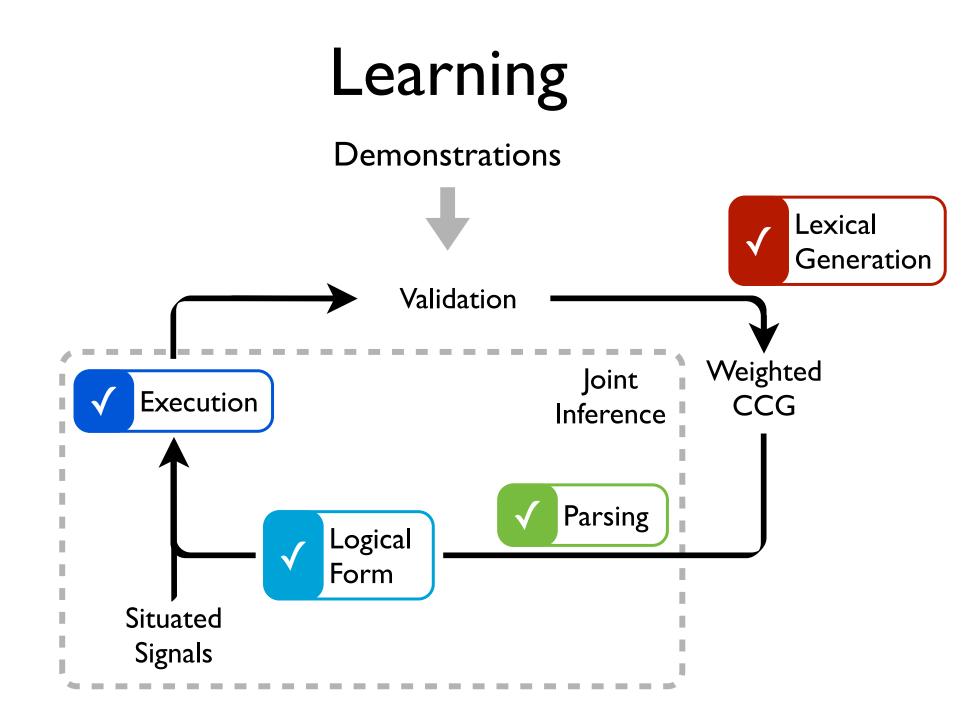
- Execution is integrated into parsing
- During parsing, execute logical forms and observe the result



Learning

- Based on a small set of seed templates
- New coarse-to-fine parsing algorithm to gradually prune the potential lexical entries
- Conservative approach to introduce new entries to the model





Validation-Driven Learning

- Online
- 2 steps:
 - Lexical generation
 - Parameter update
- Driven by a weak validation signal

Validation-Driven Learning

For T iterations, for each training sample:

- Step I: Lexical generation
 - Generate a large number of potential lexical entries
 - Parse with the generated lexicon using the model
 - Select the best valid parses from the k-best parses
 - Add their lexical items to the lexicon
- Step 2: Update parameters

Validation-Driven Learning

For T iterations, for each training sample:

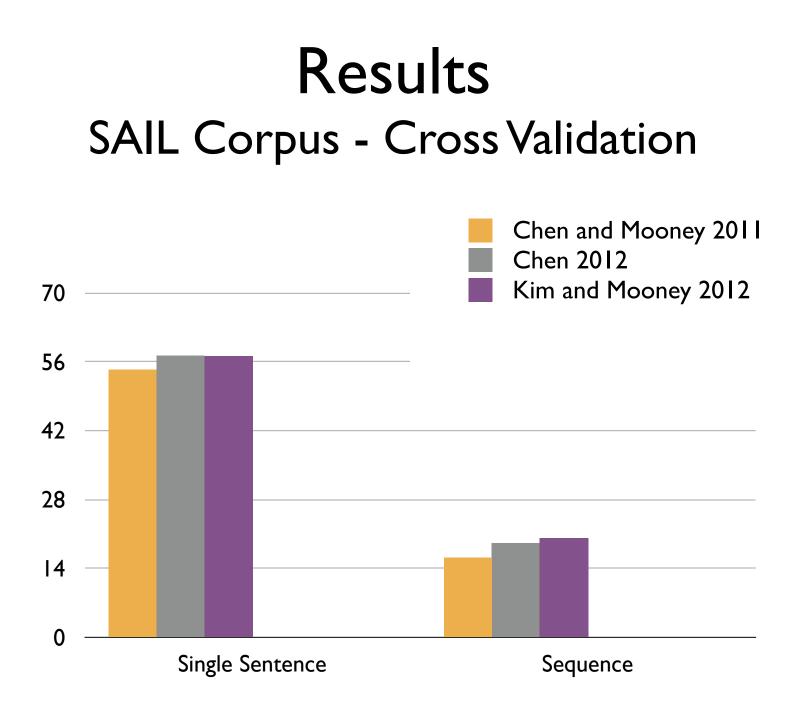
- Step I: Lexical generation
- Step 2: Update parameters
 - Parse using the model
 - Split all parses into two sets: max scoring valid and invalid
 - Find margin violating pairs between the sets
 - Do a perceptron-style update using these violations

Related Work

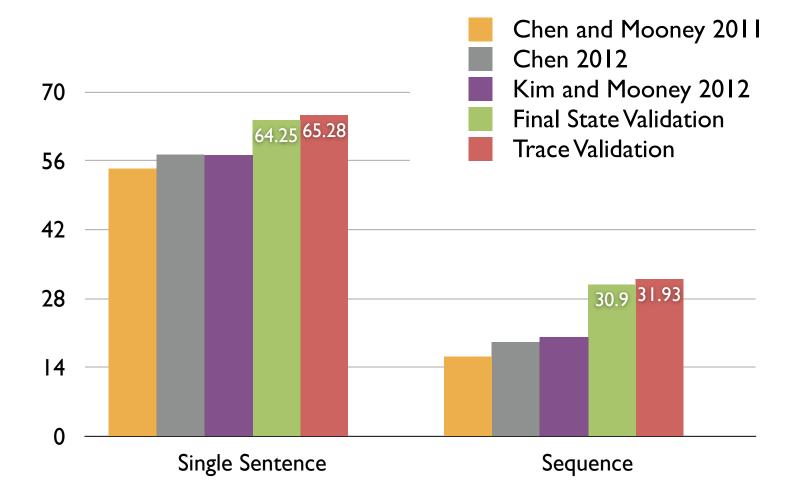
Supervised semantic parsing	[Kate, Mooney 2006; Wong, Mooney 2007; Muresan 2011]
with CCGs	[Zettlemoyer, Collins 2005; 2007; Kwiatkowski et al. 2010; 2011]
Weakly supervised semantic parsing	[Clarke et al. 2010; Goldwasser, Roth 2011; Liang et al. 2011; Kirshnamurthy, Mitchell 2012; Goldwasser et al. 2011]
Grounded Semantic Analysis	[Liang et al. 2009; Chen et al. 2010; Matuszek et al. 2012]
Executing Instructions with Shallow Representation	[Branavan et al. 2009; 2010; Vogel, Jurafsky 2010; Wei et al. 2009; Kollar et al. 2010; Tellex et al. 2011]
Non-joint Execution of Instructions	[Matuszek et al. 2010; 2012; Chen, Mooney 2011; Chen 2012; Kim, Mooney 2012]

Experimental Setup

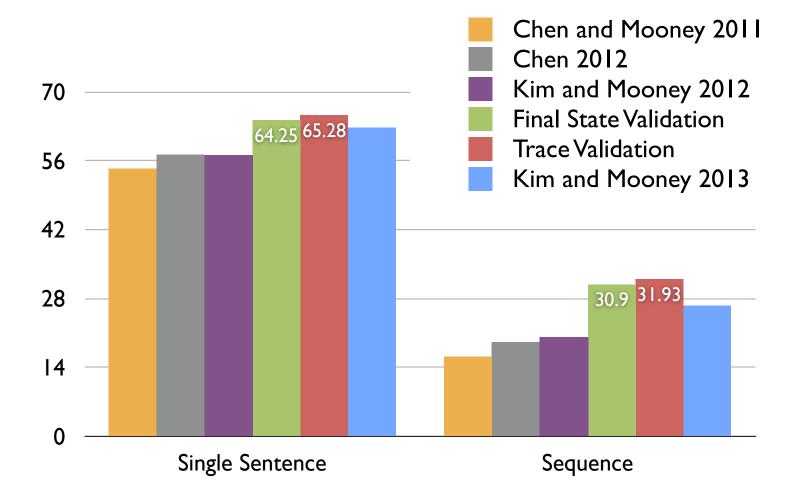
- Seed lexicon from an annotated randomly selected 12 instruction sequences
- Features: lexical, type-raising usage and repetitions in logical coordinations
- Consider only executable parses as complete



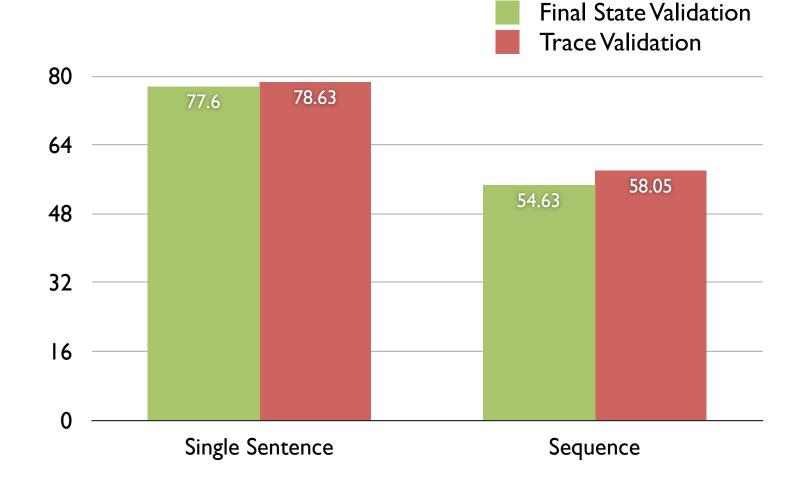
Results SAIL Corpus - Cross Validation



Results SAIL Corpus - Cross Validation



Results Oracle Corpus - Held-out Set



Contributions

- Linguistically-driven modeling of instructional language
- Joint inference for interpretation and execution of grounded language
- General weakly-supervised learning approach for semantic parsers

UW SPF

Open source semantic parsing framework

http://yoavartzi.com/spf

Semantic Parser Flexible High-Order Logic Representation

Learning Algorithms

UW SPF

Open source semantic parsing framework

<u>http://yoavartzi.com/spf</u>

Semantic Parser

Flexible High-Order Logic Representation

Learning Algorithms

Navigation code and data available online Coming up: ACL tutorial

http://yoavartzi.com/navi

[fin]