

Situated Comprehension of Action Commands

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Outline

- ① Introduction
- ② Grounded Comprehension in BOLT
- ③ Agent Design
- ④ Discussion
- ⑤ Conclusions

Linguistic Agents

- Use natural language for task-oriented communication.
 - Work collaboratively with humans
 - Learn from natural language communication in a complex environment

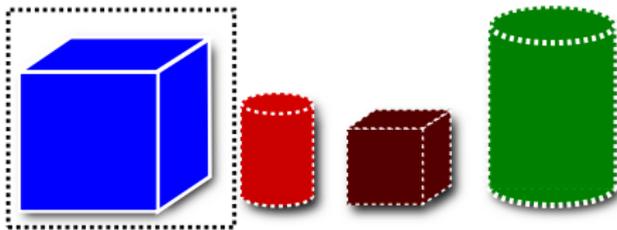
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- Focus on
 - Comprehension of actions commands
 - Move the blue, large cube to the table.
 - Acquire new knowledge (linguistic/extra linguistic)
 - How actions can be performed?
 - How language is used?

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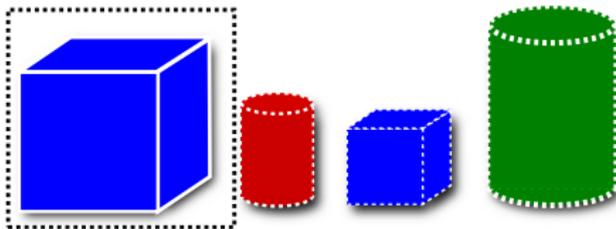
Pick up the blue cube.



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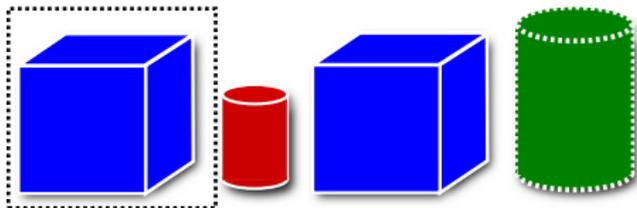
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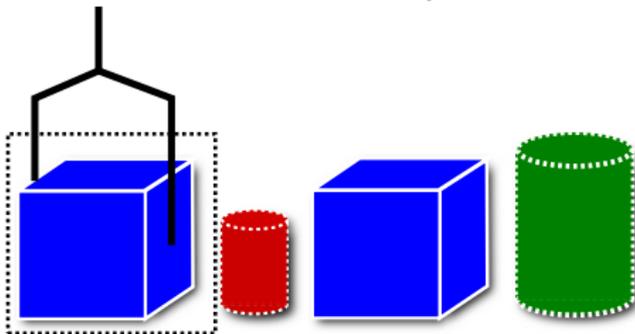
Pick up the cube on the left of the red cylinder.



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Put down the object.



The Indexical Hypothesis for Comprehension

Glenberg and Robertson (1999)

- Sentences become meaningful by grounding their interpretation in situated action
 - *index* words and phrases to referents.
 - derive *affordances* from these referents.
 - *mesh* these affordances under syntactical constraints, physical constraints of the environment

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- Our contribution
 - Formalization
 - Implementation in Soar

Using Language for Indexing

Barsalou (1999)

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- Immediate Indexing
 - *"the white board"*
 - Participants are simultaneously embedded in the environment
 - Language is used to refer to objects and event in the current situation.
 - Used in learning nouns/adjectives and spatial relationships

Using Language for Indexing

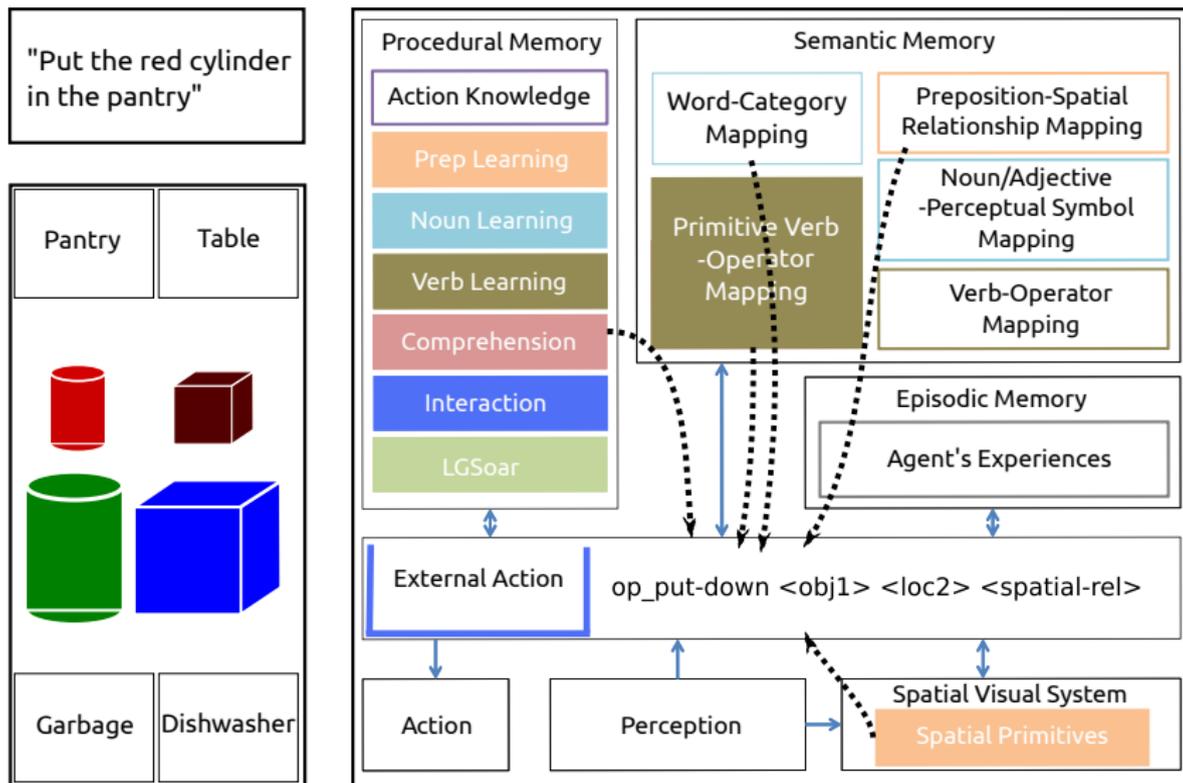
Barsalou (1999)

- Immediate Indexing
 - *"the white board"*
 - Participants are simultaneously embedded in the environment
 - Language is used to refer to objects and event in the current situation.
 - Used in learning nouns/adjectives and spatial relationships
- Displaced Indexing
 - *"the parking lot in front of BBB"*
 - Referents are not currently present
 - Language is used to refer to objects and events from prior experiences with the environment
 - *shared, componential, future*

Situated Comprehension

← Interaction Management

Behavior Execution →



Environment and Representation

- (limited) Partial Observability
 - Distance limited sensing
 - Complete value assignment is known, if perceptible

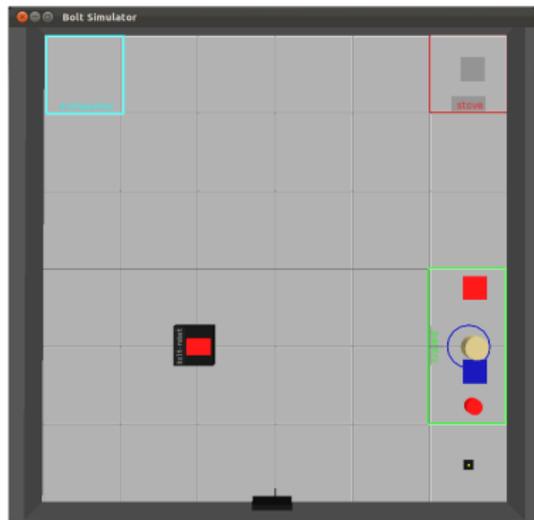


Figure: BOLT++

Environment and Representation

- (limited) Partial Observability
 - Distance limited sensing
 - Complete value assignment is known, if perceptible
- Primitive actions
 - `goto`, `pick-up(obj)`,
`put-down(obj)`
 - `put-down(obj, loc)`,
`put-down(obj, obj)`
 - `known`
proposal/application/termination
 - Affordance based proposals
 - `pick-up` proposed for all 'perceptible' objects
 - `put-down` proposed for objects in the gripper
 - ...

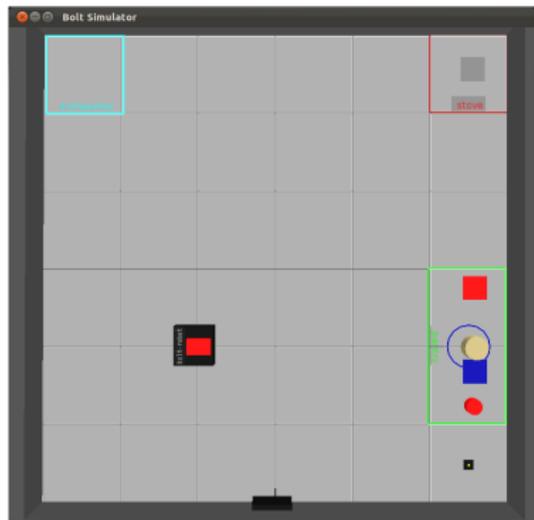
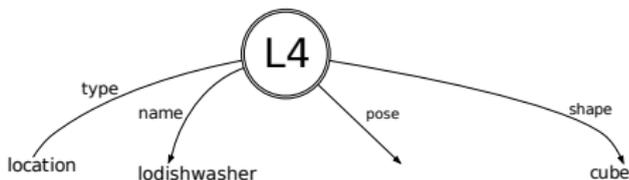


Figure: BOLT++

Background Knowledge

(Noun/Adj:Perceptual Symbols, Preposition:Spatial Relationship)

- Domain Semantic Knowledge

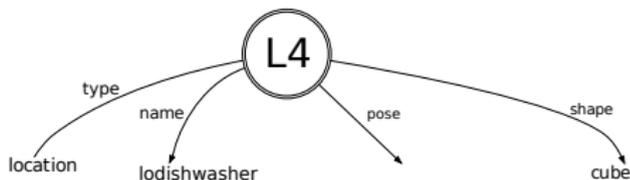


- May be acquired from previous experiences
- Allows the agent to communicate non-perceptible locations

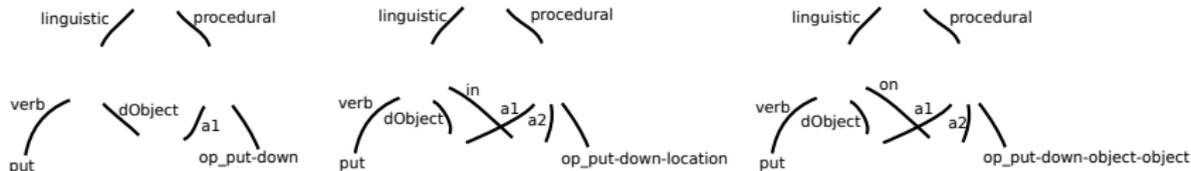
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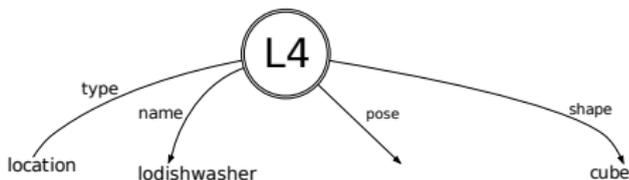
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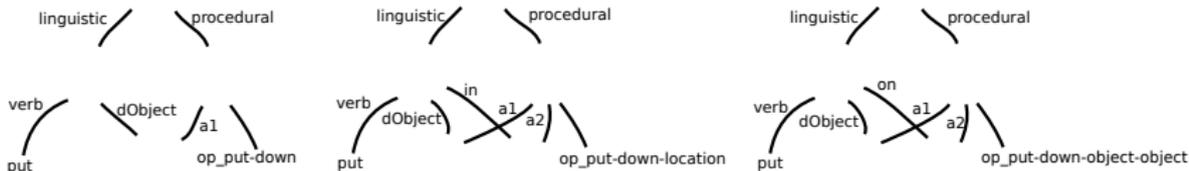
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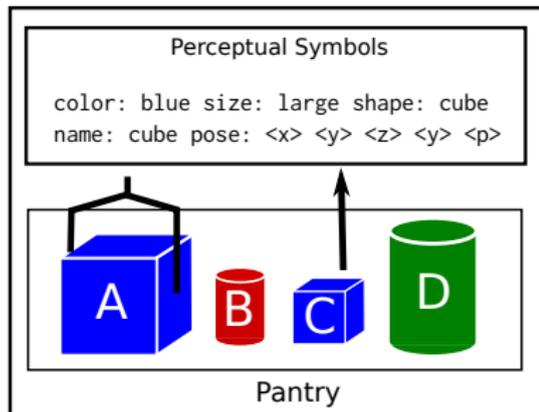


- Action Models
 - Changes in the world for primitive actions

Comprehension of Action Commands

"Put a blue cube in the dishwasher"

Phase: **Index Arguments**

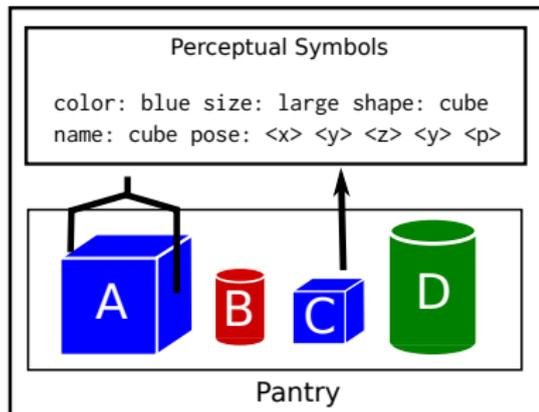


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 - add all matching objects to arg-candidate set $CA_{dO} = \{A, C\}$
 - if nothing matches, displaced indexing

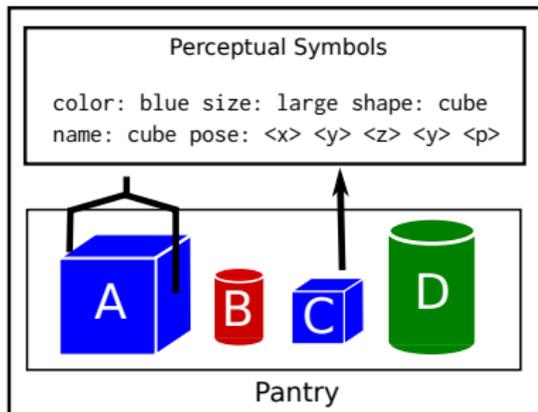


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 - match description to semantic objects
 - add all to arg-candidate set $CA_{in} = \{dishwasher\}$

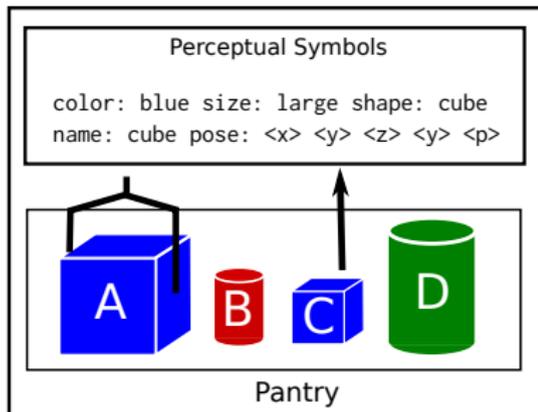


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- If $CA = \phi$, communicate

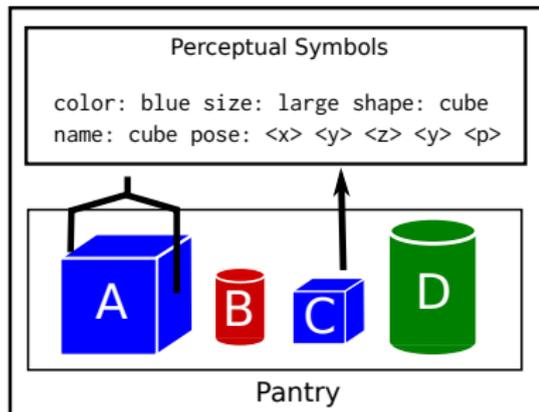


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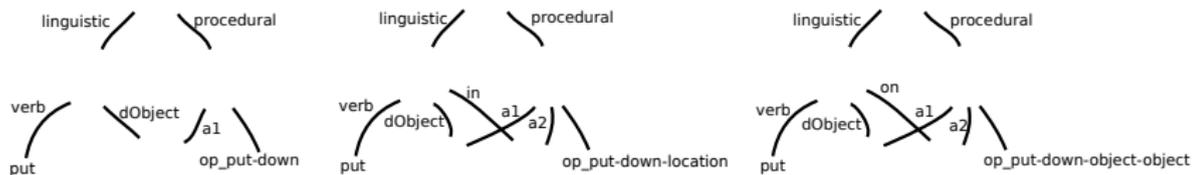
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- Displaced indexing into episodic memory?



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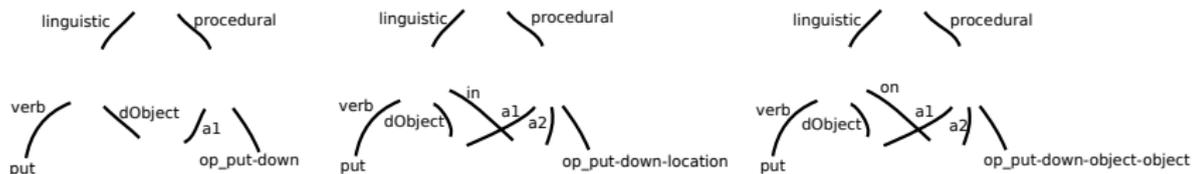
Phase: **Index Verb**



Comprehension of Action Commands

"Put a blue cube in the dishwasher"

Phase: **Index Verb**



- Possible Interpretations *CI*

op_put-down-object-location [A] [dishwasher]

op_put-down-object-location [C] [dishwasher]

Comprehension of Action Commands

"Put a blue cube in the dishwasher"

Phase: **Meshing**

- Generate all possible interpretations *CI*
 - Using all candidate objects, *CA* for all arguments of the verb

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 - Using affordances, physical constraints
 - Soar proposals
 - `op_put-down-object-location [A] [dishwasher]`

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- Excute action if only one element
 - Ground preposition: predicate projection, tracking

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- Communicate if empty or multiple elements

Linguistic Capabilities

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- Situated Referent Resolution
 - Using perceptual information
 - Use of most distinctive description given the perceptual state
 - Using semantic knowledge
 - mapping *the dishwasher* to a semantic object
 - Using procedural knowledge
 - Can only put down the object in gripper

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- Situated Referent Resolution
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 - Using procedural knowledge
 - Can only put down the object in gripper
- Situated Action Resolution
 - Using the argument structure

Future Work

- Resolve PP phrase ambiguities using context
- Use episodic memory for displaced indexing
- Co reference resolution
- Exploiting other context
 - linguistic, interaction, procedural, perceptual, semantic, episodic

Nuggets and Coal

- Nuggets
 - Proposed a scheme for comprehending action commands
 - Uses real-world context
- Coal
 - Only action-commands!
 - Evaluation?
 - Information theoretic analysis