

SCI 52 – **Artificial Intelligence:**
An Introduction to Neural Networks and Deep Learning

Cognitive Architecture

and Conversational Agents

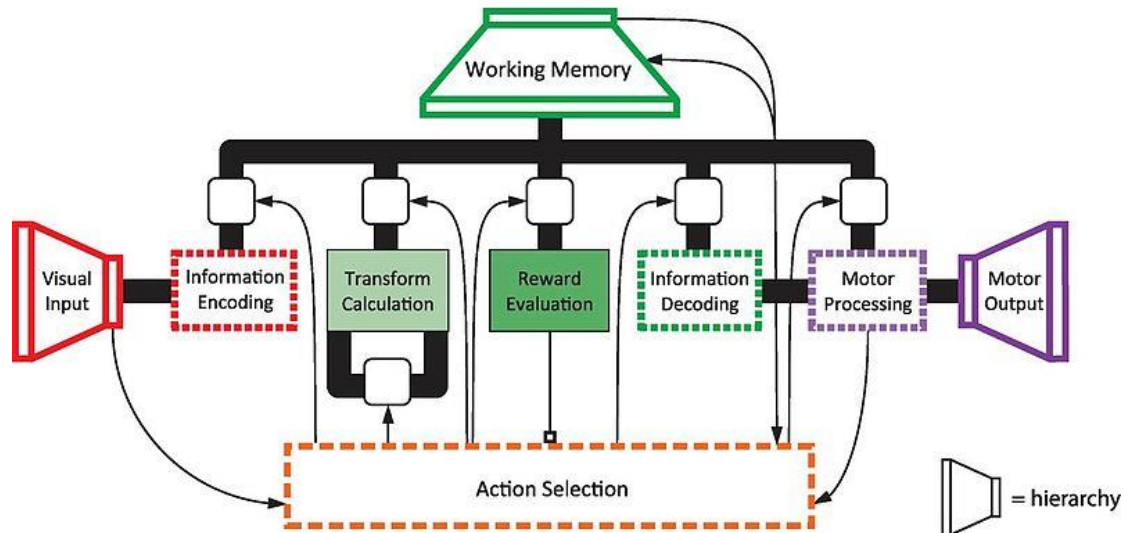
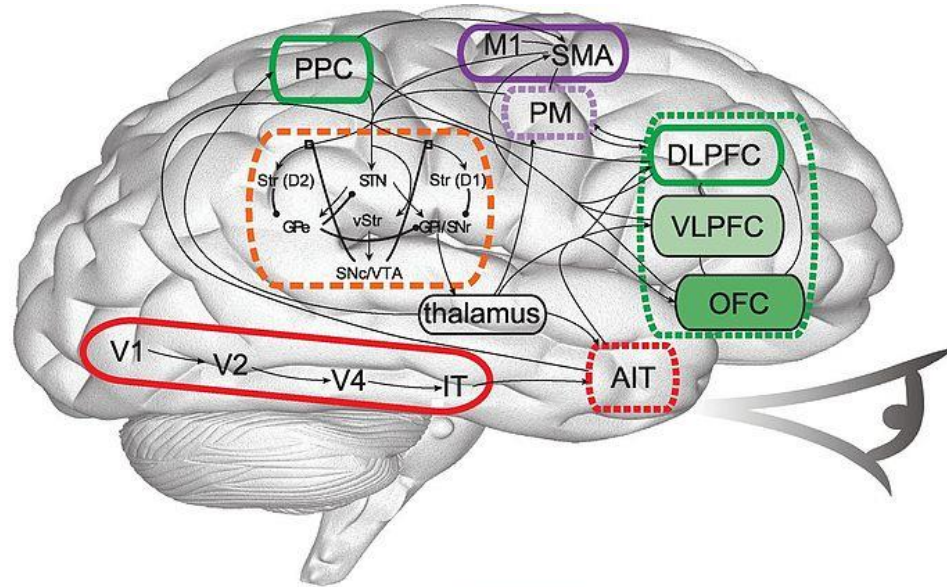
Eric Saund, Ph.D.
February, 2019

www.saund.org

The organization of structural and functional elements to achieve some purpose.



Brains and Minds



Cognitive Architecture: Outline

- Architecture in Information Systems
 - “An organization of elements to achieve some information processing purpose.”
- History
 - Computational theory of mind
- The Baseline/Generic Cognitive Architecture
- Examples
 - Soar, LIDA, CopyCat
- Architecture in NN / Deep Learning Networks
- Conversational Agents
 - NLP
 - Dialog Management
 - Knowledge Graphs
 - Deep Networks

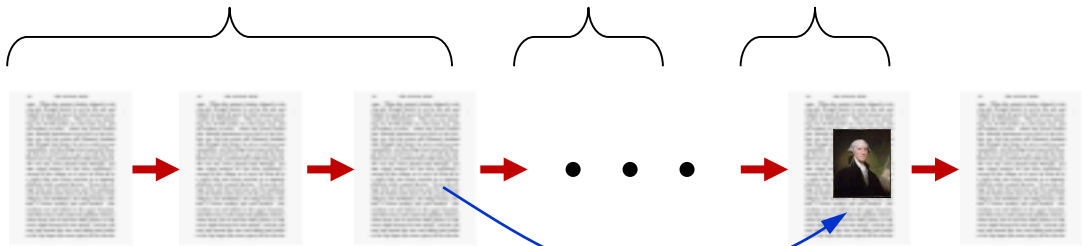
Information Architecture: A Book



Table of Contents

page numbers

logical sections



content Pages

cross-references

page numbers

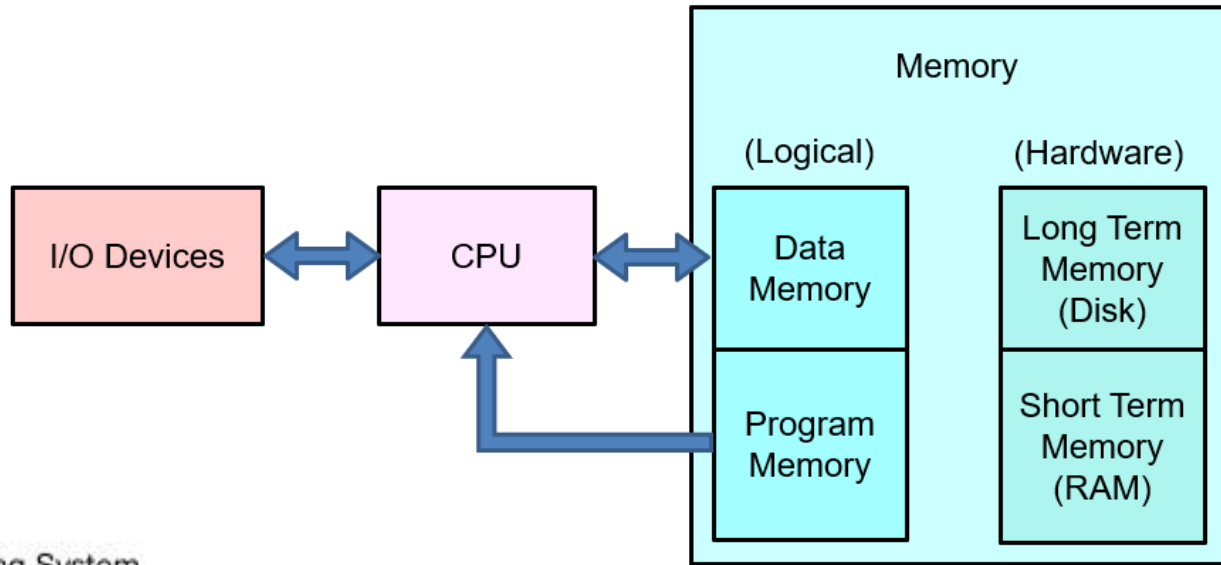
content items



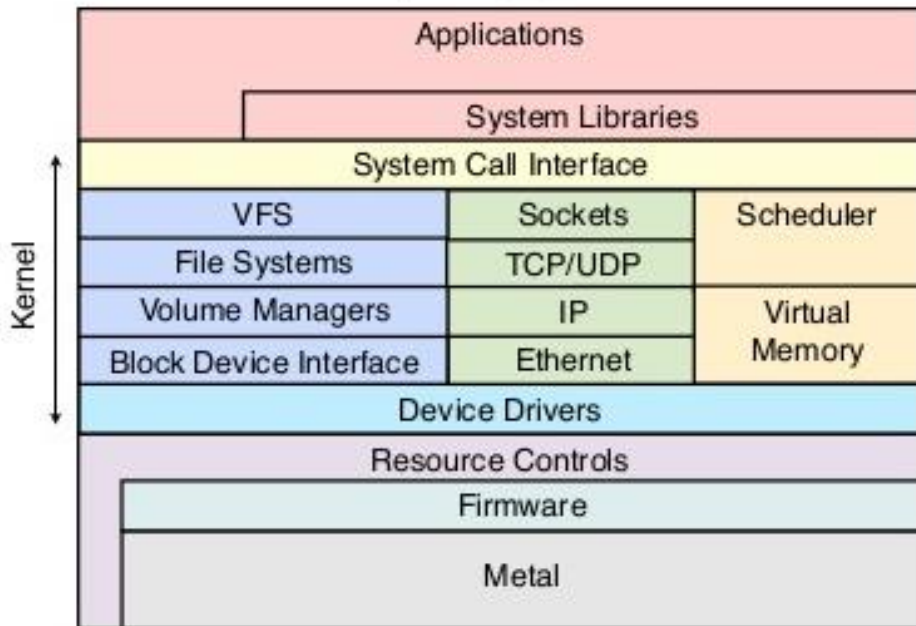
Index	
of areas, 10	with, 104-105
Alpha, 10, 11	gamma, 10
Alpha, 10, 11	hybrid, 104-105
Alpha, 10, 11	beta, 10
Alpha, 10, 11	delta, 10
Alpha, 10, 11	epsilon, 10
Alpha, 10, 11	zeta, 10
Alpha, 10, 11	eta, 10
Alpha, 10, 11	theta, 10
Alpha, 10, 11	iota, 10
Alpha, 10, 11	kappa, 10
Alpha, 10, 11	lambda, 10
Alpha, 10, 11	mu, 10
Alpha, 10, 11	nu, 10
Alpha, 10, 11	xi, 10
Alpha, 10, 11	omicron, 10
Alpha, 10, 11	pi, 10
Alpha, 10, 11	rho, 10
Alpha, 10, 11	sigma, 10
Alpha, 10, 11	tau, 10
Alpha, 10, 11	upsilon, 10
Alpha, 10, 11	phi, 10
Alpha, 10, 11	chi, 10
Alpha, 10, 11	psi, 10
Alpha, 10, 11	omega, 10
Alpha, 10, 11	various, 10
Alpha, 10, 11	etc., 10

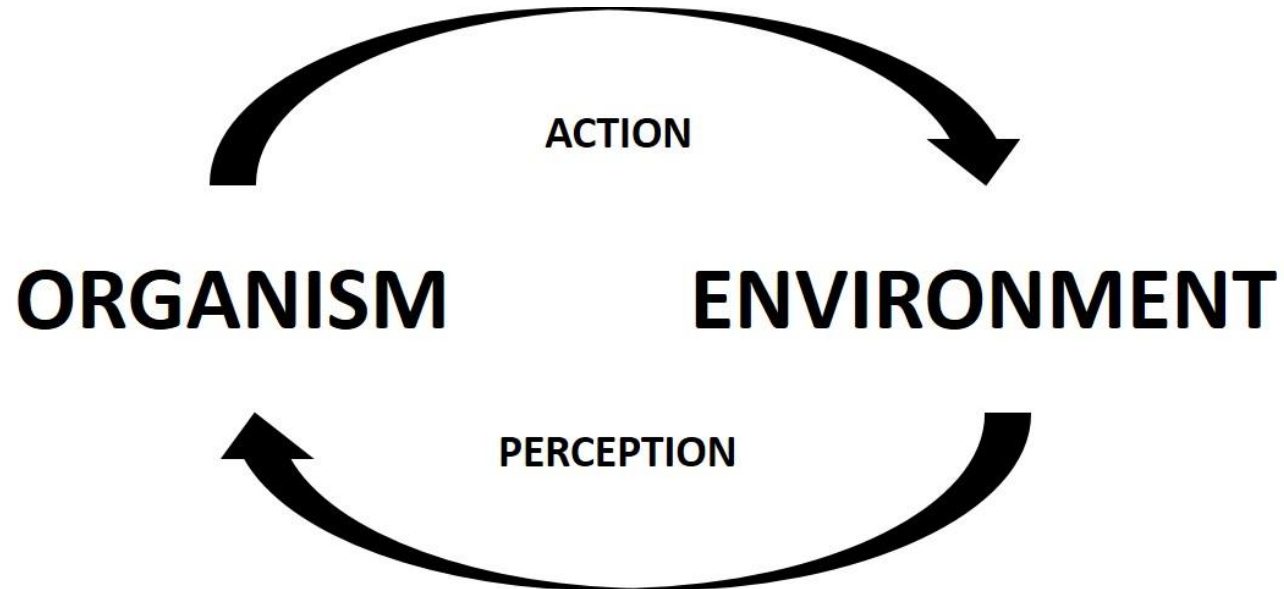
Index

Information Architecture: A Computer

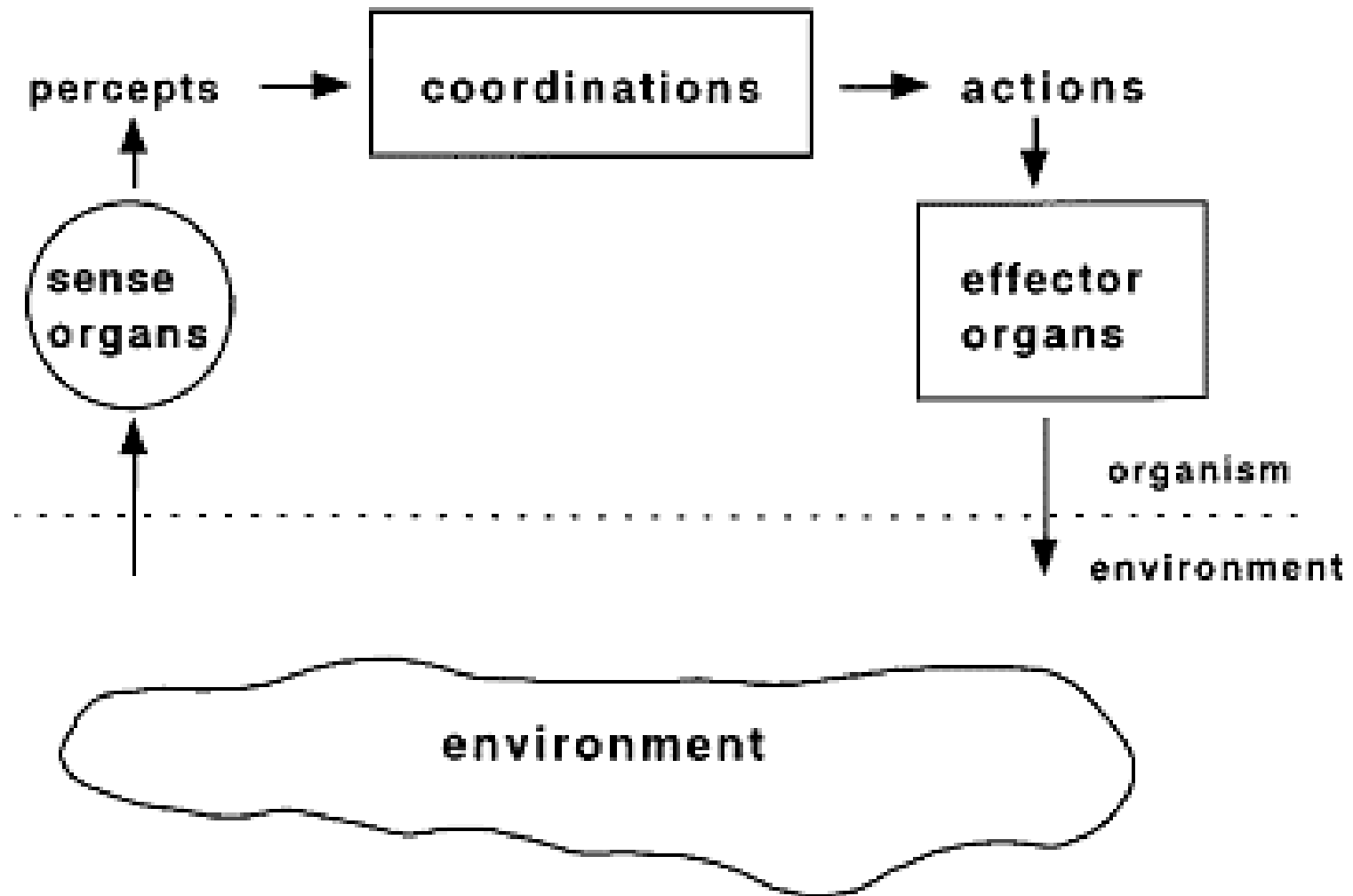


Operating System





Cognitive Architecture: Basic Agent



Cognitive Architecture: Historical Roots

1800s

1900s

Psychoanalytic Theories

Freud, Jung

Perceptual & Phenomenal Psychology

Helmholtz, William James

Behaviorism

Pavlov, B.F. Skinner

Cybernetics

Norbert Wiener

Theory of Computation

Turing, von Neumann

Artificial Intelligence

McCarthy, Minsky

Computational Theory of Mind

McCulloch and Pitts

Chomsky

Newell & Simon

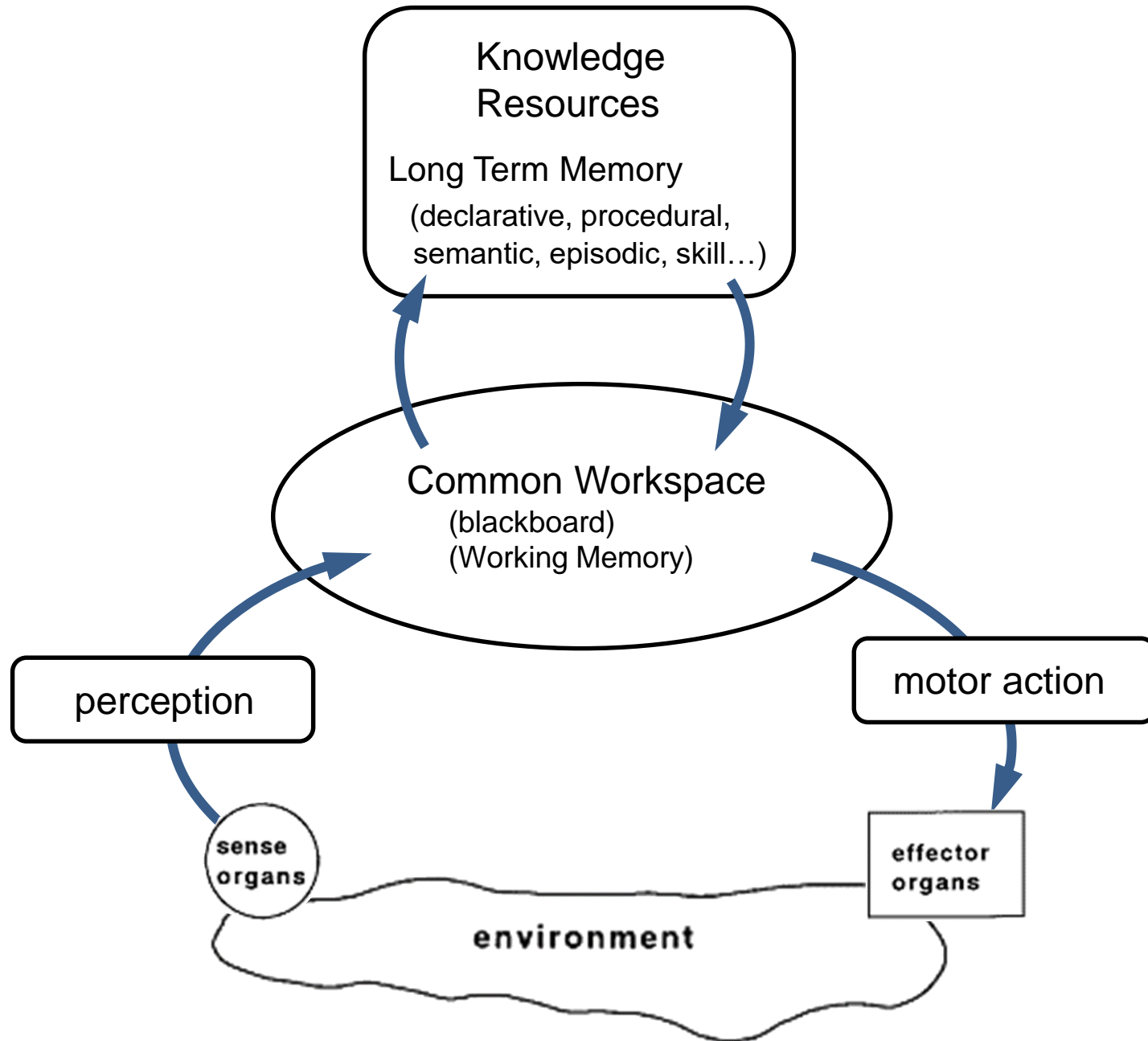
Guiding Metaphor

- engines and hydraulics
- signal transmission

- signal processing
- calculating machines

- computers

Cognitive Architecture: Standard Model



Cognitive Architecture: Big Questions

- What are the *types of content* held in the workspace?
 - percepts
 - beliefs
 - memories
 - goals
 - intentions & plans
 - emotions
 - attitudes
- What are the *representations* for state and knowledge?
 - activation patterns over fixed vectors
 - graphs of objects and relations
 - frequencies and phases of waveforms
- How is processing *controlled*?
 - automatic processes
 - conscious deliberation
 - selection of operators

Soar

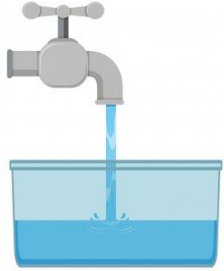
(Newell, Laird,
1983 -> present)

- Definition of intelligence:
 - problem states and transitions
 - solutions found through *search* in state space
 - Representation:
 - graphs of objects and relations
 - Control: production system
 - Working Memory blackboard
 - procedural knowledge
 - declarative knowledge
- } Long-Term Memory

Soar: Water Jug Problem Example

Definition of intelligence:

- problem states and transitions
- solutions found through *search* in state space



5 gal.

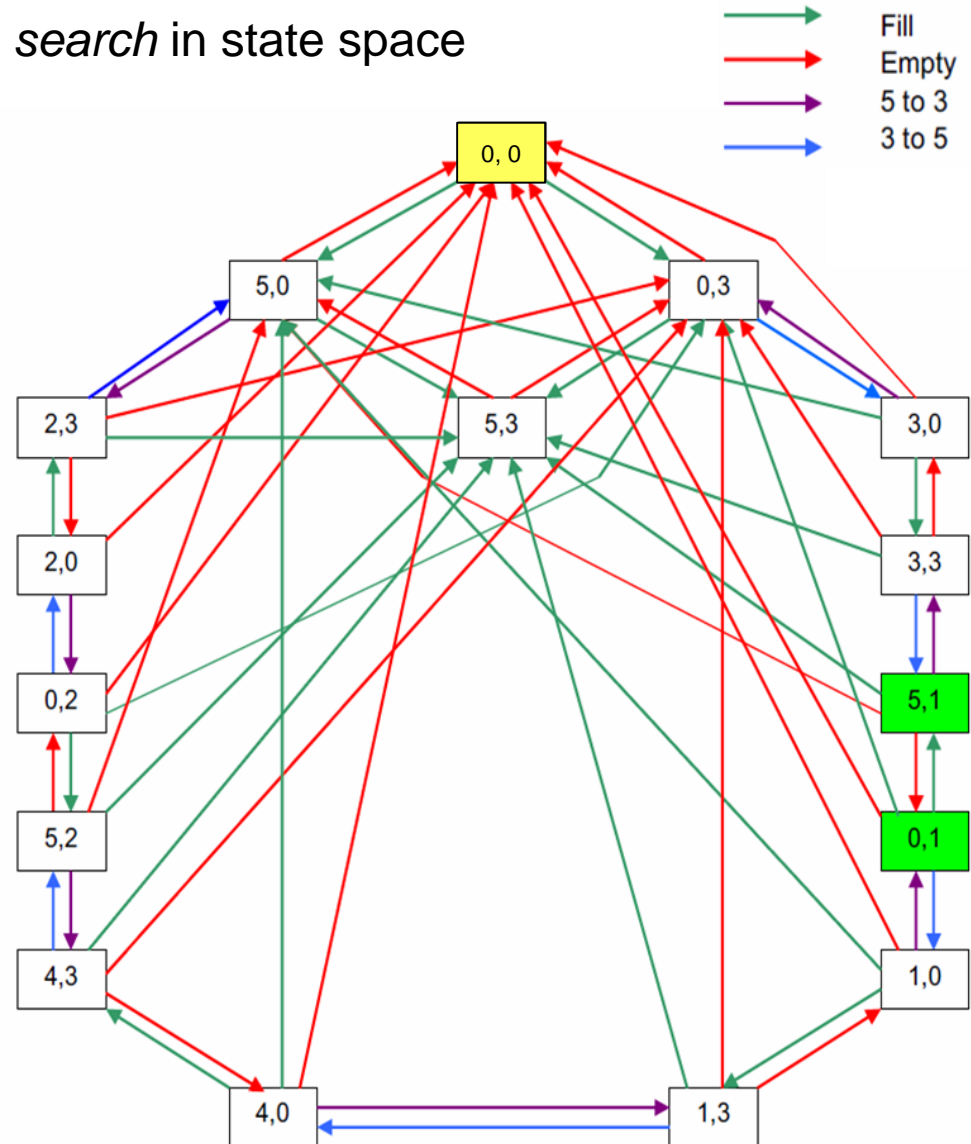
3 gal.



Start state:
both jugs
empty.



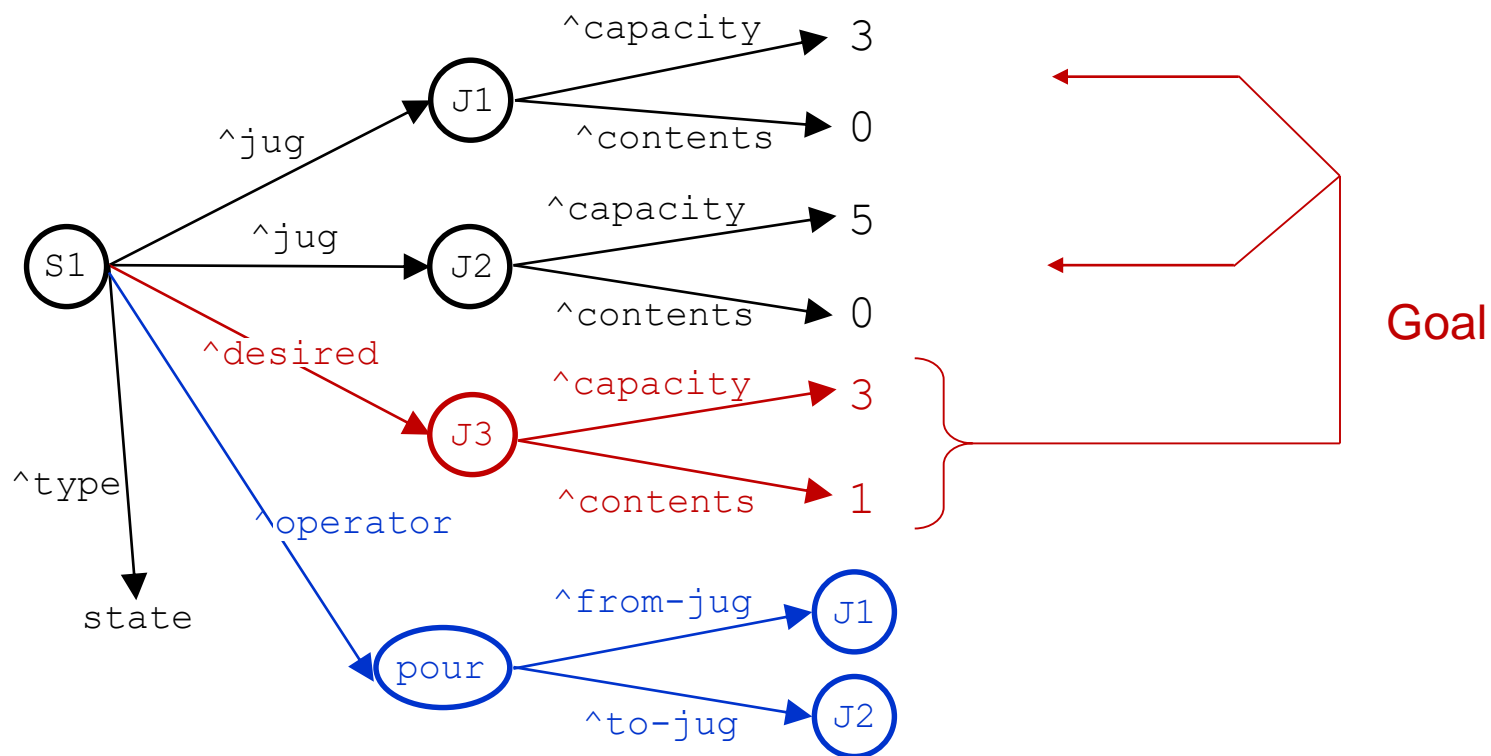
Goal state:
3-gallon jug
contains
1 gallon of water.



Representation in Soar

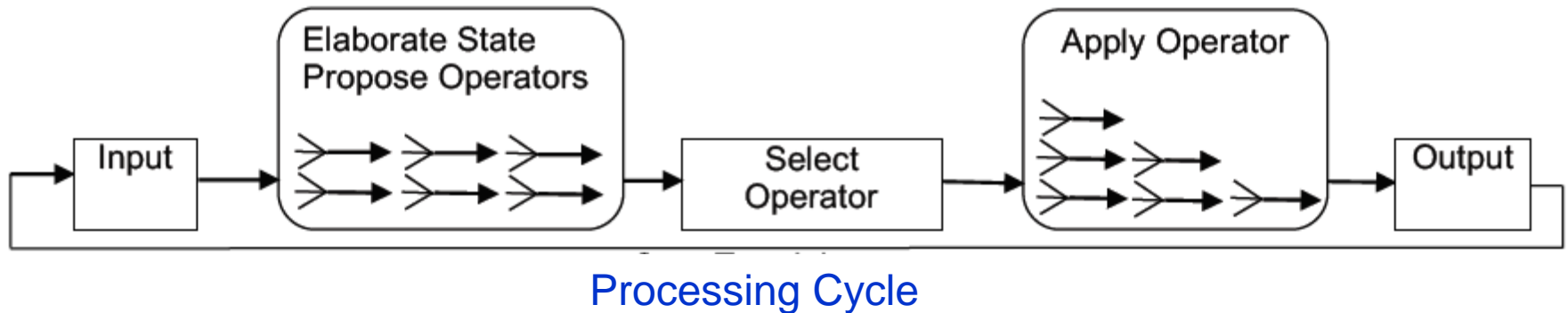
Graph

- data objects
- attributes & relations
- operators
- Working Memory (state)
- Long-Term Memory (knowledge)



Production System

- Working Memory blackboard
- declarative knowledge - what
- procedural knowledge – how
 - rules
 - operators
- subgoal states



Executive Function (Psychology, Cognitive Neuroscience):

- update Working Memory from sensory and Long-Term Memory resources
- focus attention, inhibit distractors
- shift task context





Cognitive Architecture: Outline

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Marr's Three Levels of Abstraction

David Marr: Theoretical Neuroscience  *Computational Intelligence*
what? why?

Example

- **Computational Theory**

What is the computation and by what principles is it accomplished?

Textbook:
ToC, body text, index

- **Algorithm**

What representations and algorithms are used to carry forth computation?

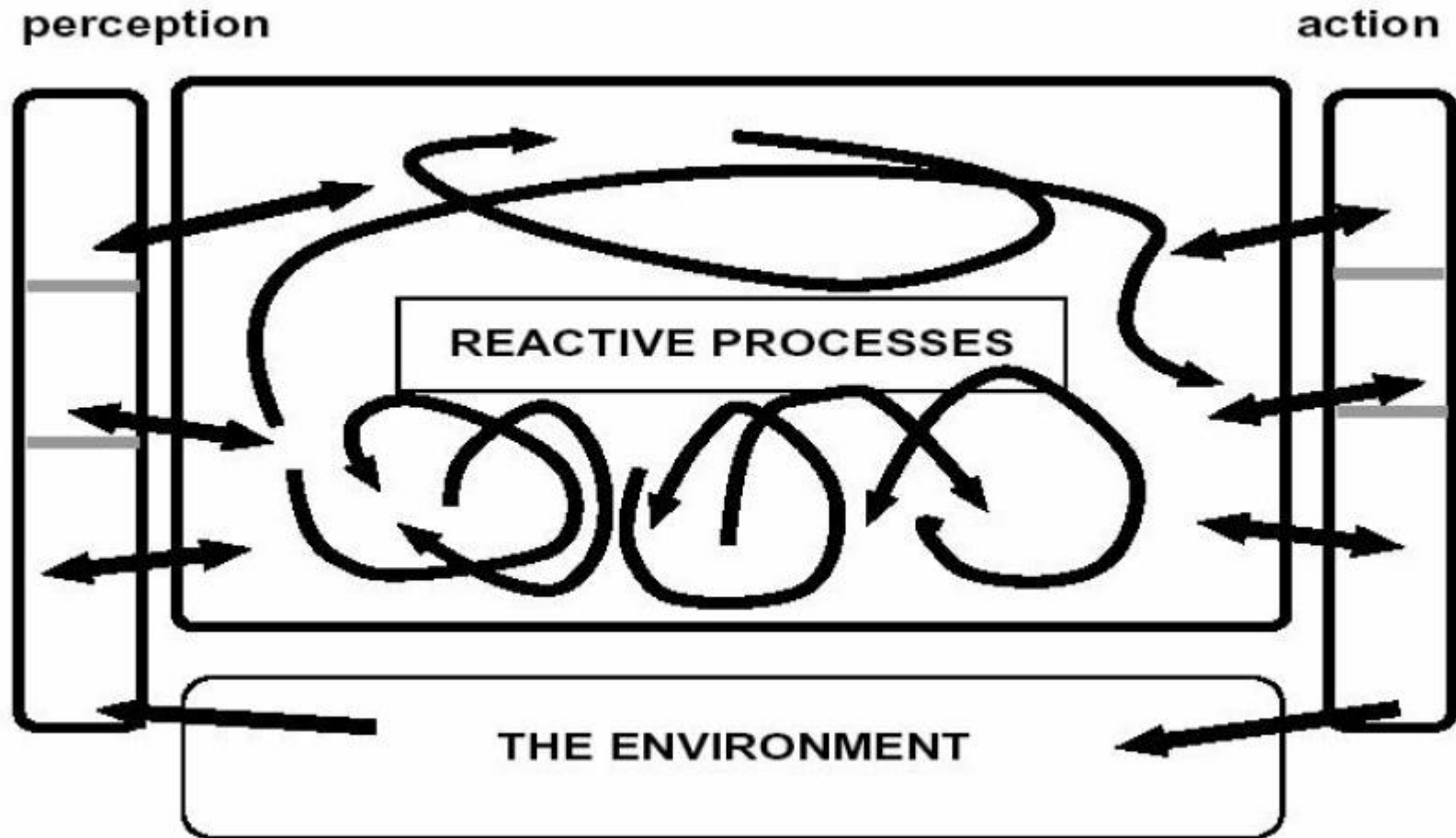
Scrolling text roll,
manual + automatic
positioning

- **Implementation**

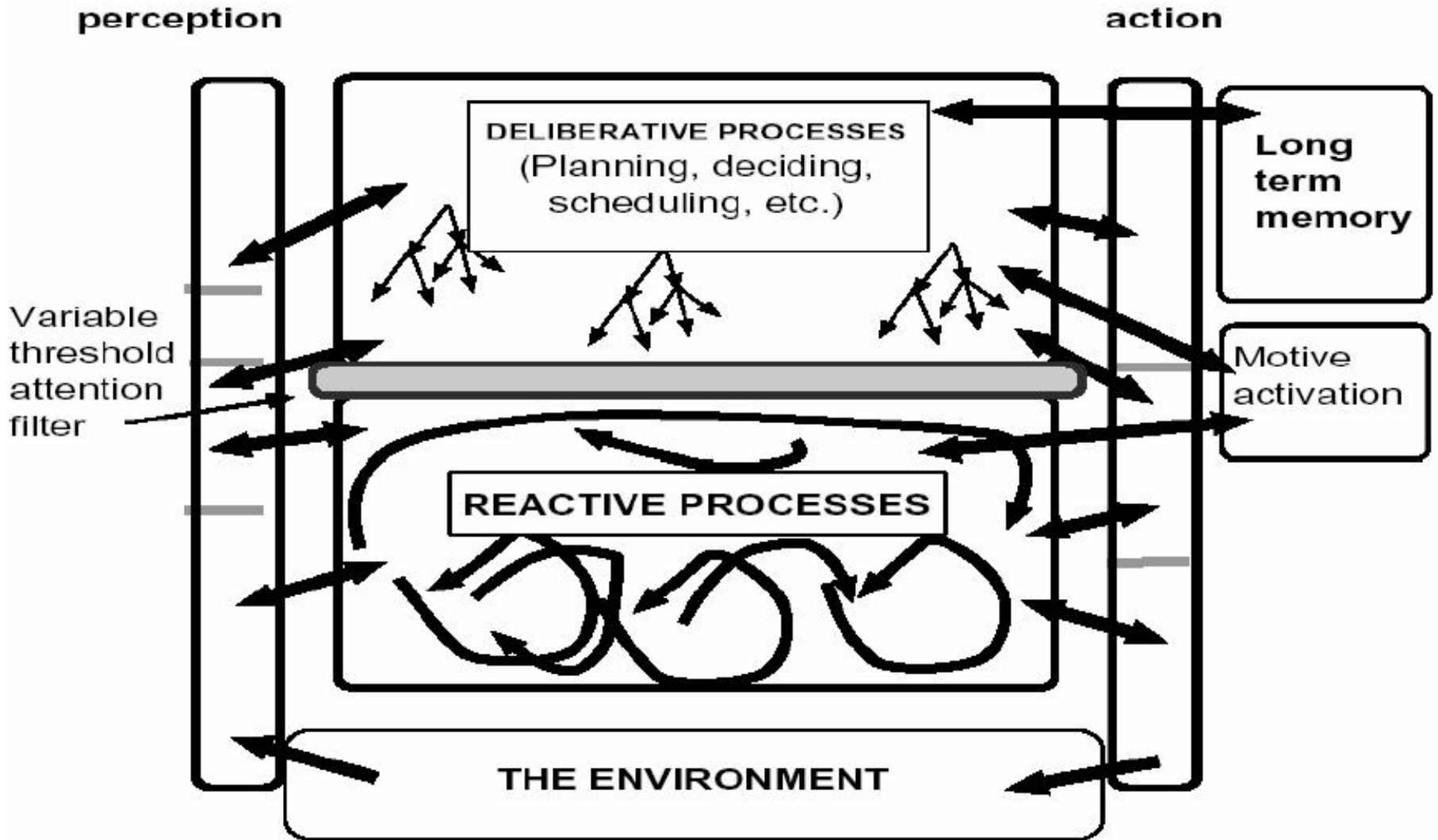
On what physical hardware and firmware is the algorithm run?

Projection of markings
printed on acetate film,
keyboard controls scrolling
per ToC and index.

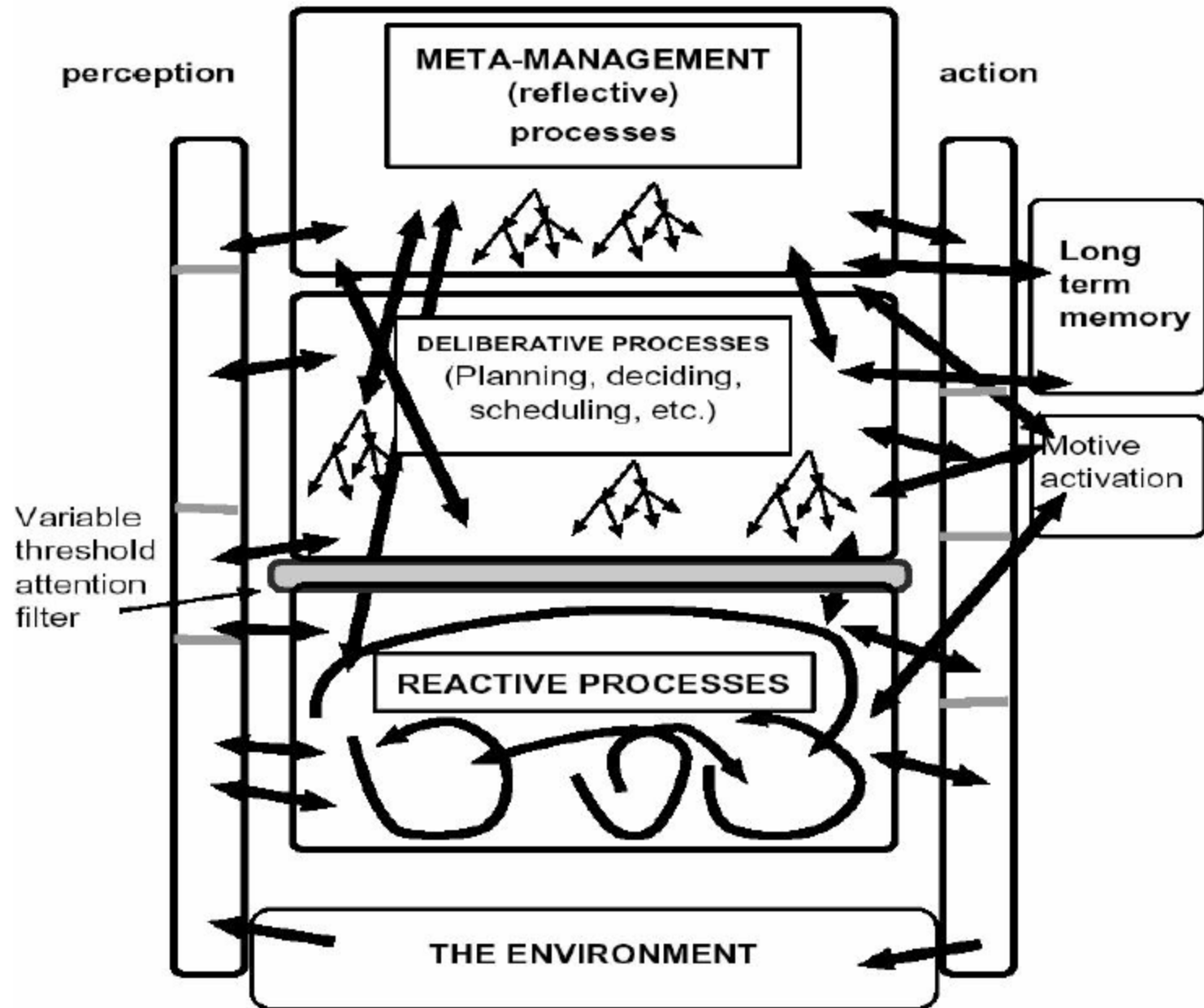
Cognitive Architecture: Reactive Agent



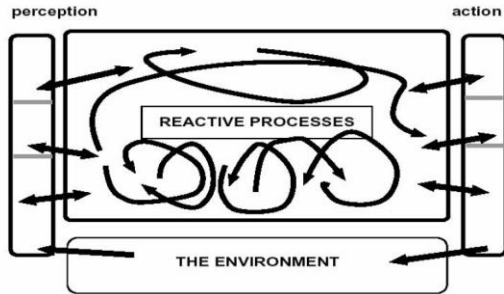
Cognitive Architecture: Deliberative Agent



Cognitive Architecture: Reflective Agent



Reactive vs. Deliberative



Reactive

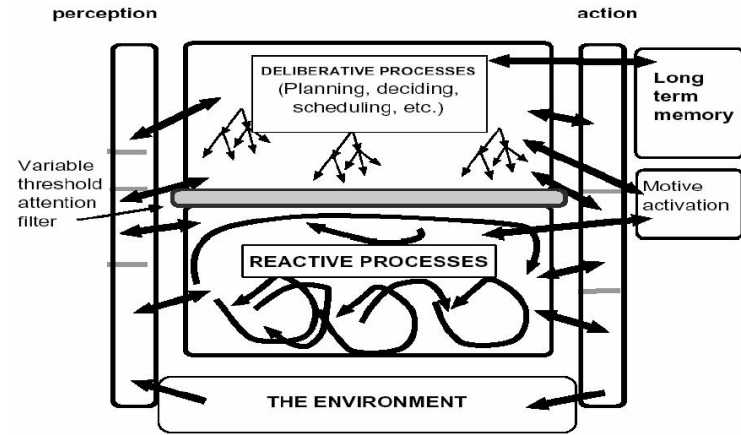
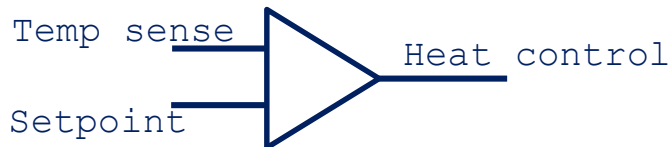
automatic & strictly determined

modest internal state

implicit representations

Kahneman System 1

Example: thermostat



Deliberative

makes choices

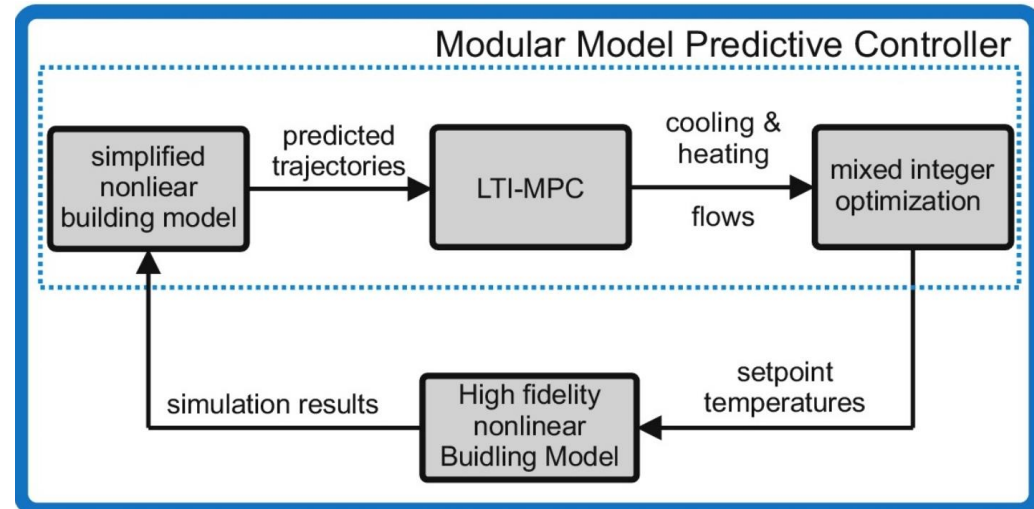
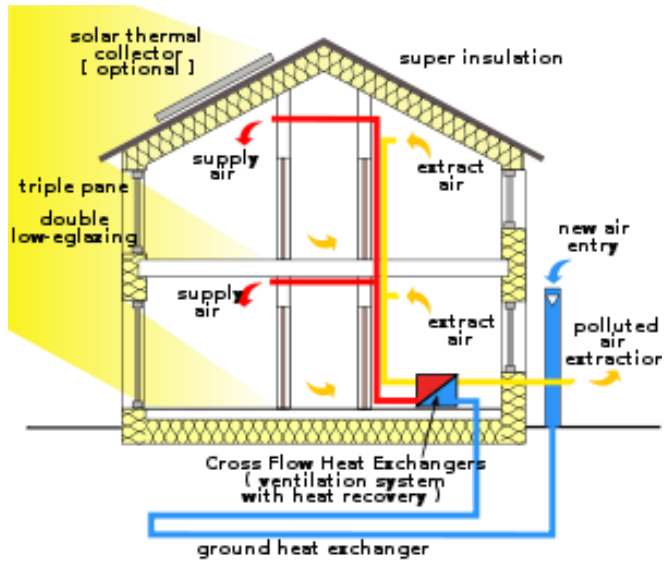
rich internal state

explicit world models

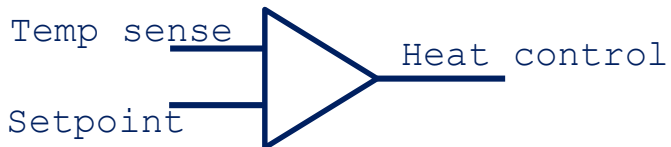
Kahneman System 2

Example: building temperature management system

Reactive vs. Deliberative Building Temperature Controller

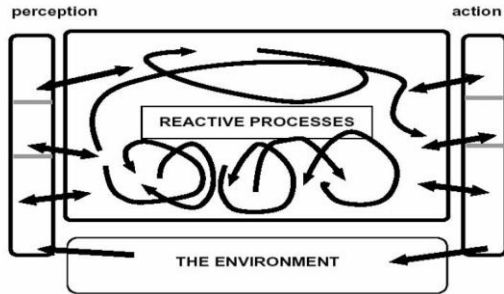


Example: thermostat



Example: building temperature management system

Reactive vs. Deliberative

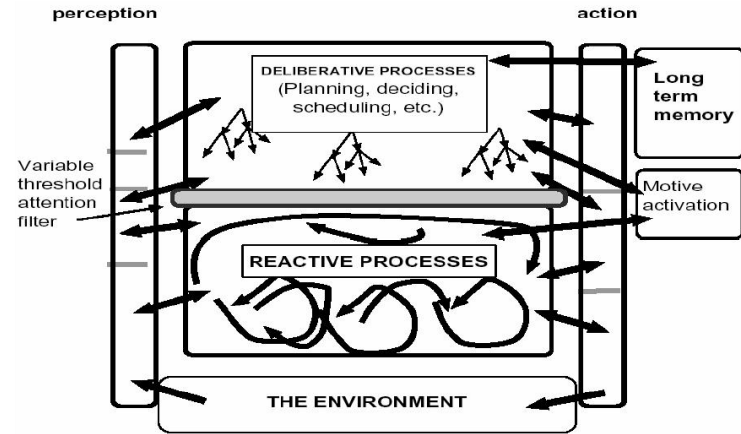


Reactive

automatic & strictly determined

modest internal state

implicit representations



Deliberative

makes choices

rich internal state

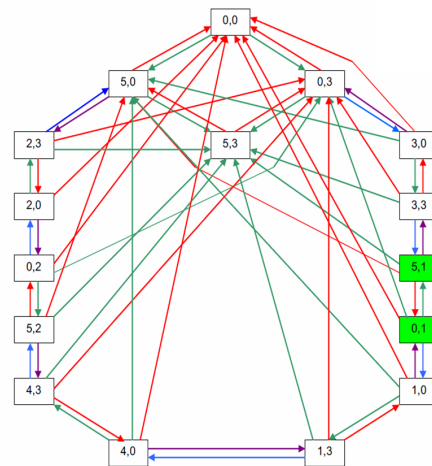
explicit world models

?

Soar

A Reactive Water Jug Solution in Soar

Deliberative search space



jug-5, jug-3

Reactive program rules

If $0, 0$ Then `FILL` (3) \Rightarrow $0, 3$

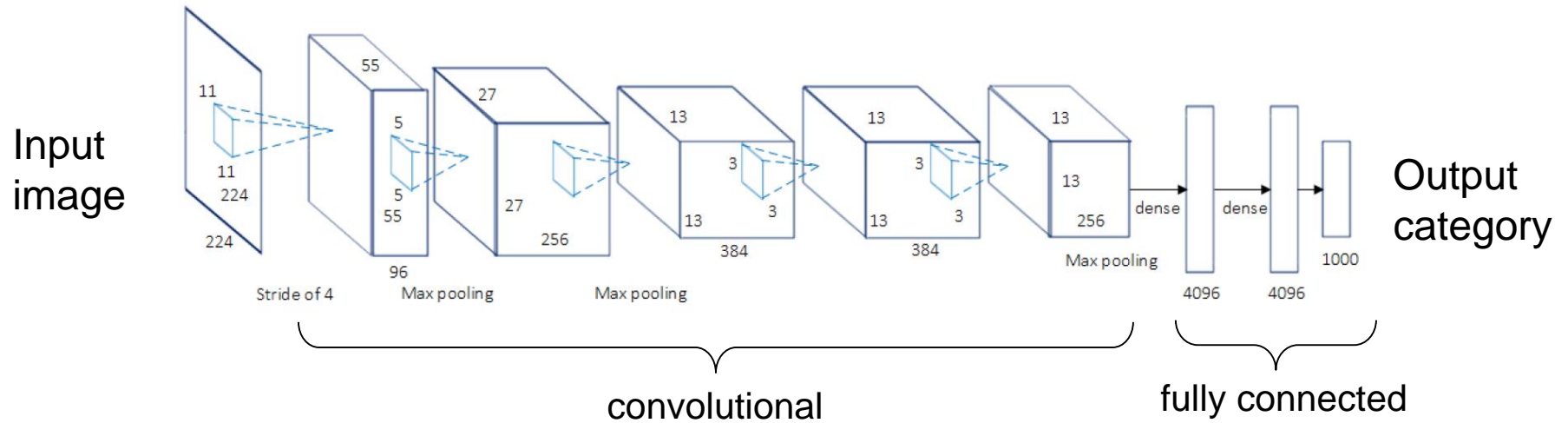
If $0, 3$ Then `Pour` (3, 5) \Rightarrow $3, 0$

If $3, 0$ Then `Fill` (3) \Rightarrow $3, 3$

If $3, 3$ Then `Pour` (3, 5) \Rightarrow $5, 1$



Neural Network Architectures



Alexnet

- Architectural Elements

Layer dimensions,
weights, nonlinearities

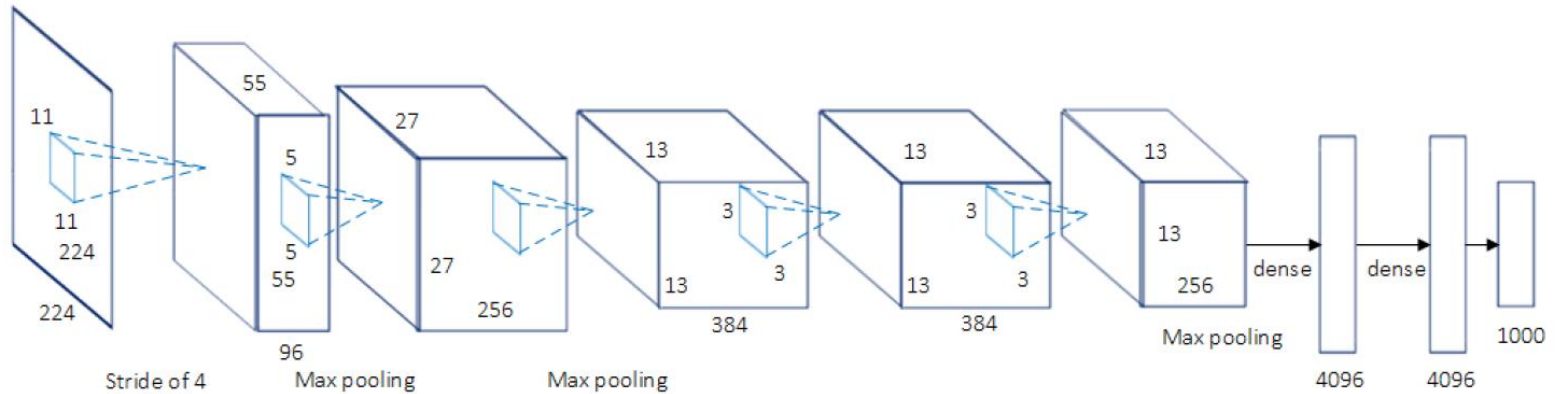
- Organization

Layer connectivity

- Purpose

Function approximation

Neural Network Architectures



Reactive or deliberative?

choices?
state?
implicit/explicit

Cognitive Architecture

- Naturally Intelligent Agent
- Artificially Intelligent Agent

Where does knowledge reside?

How is state represented and utilized in decisions?

Forms of Memory

Localist vs. Distributed Representations

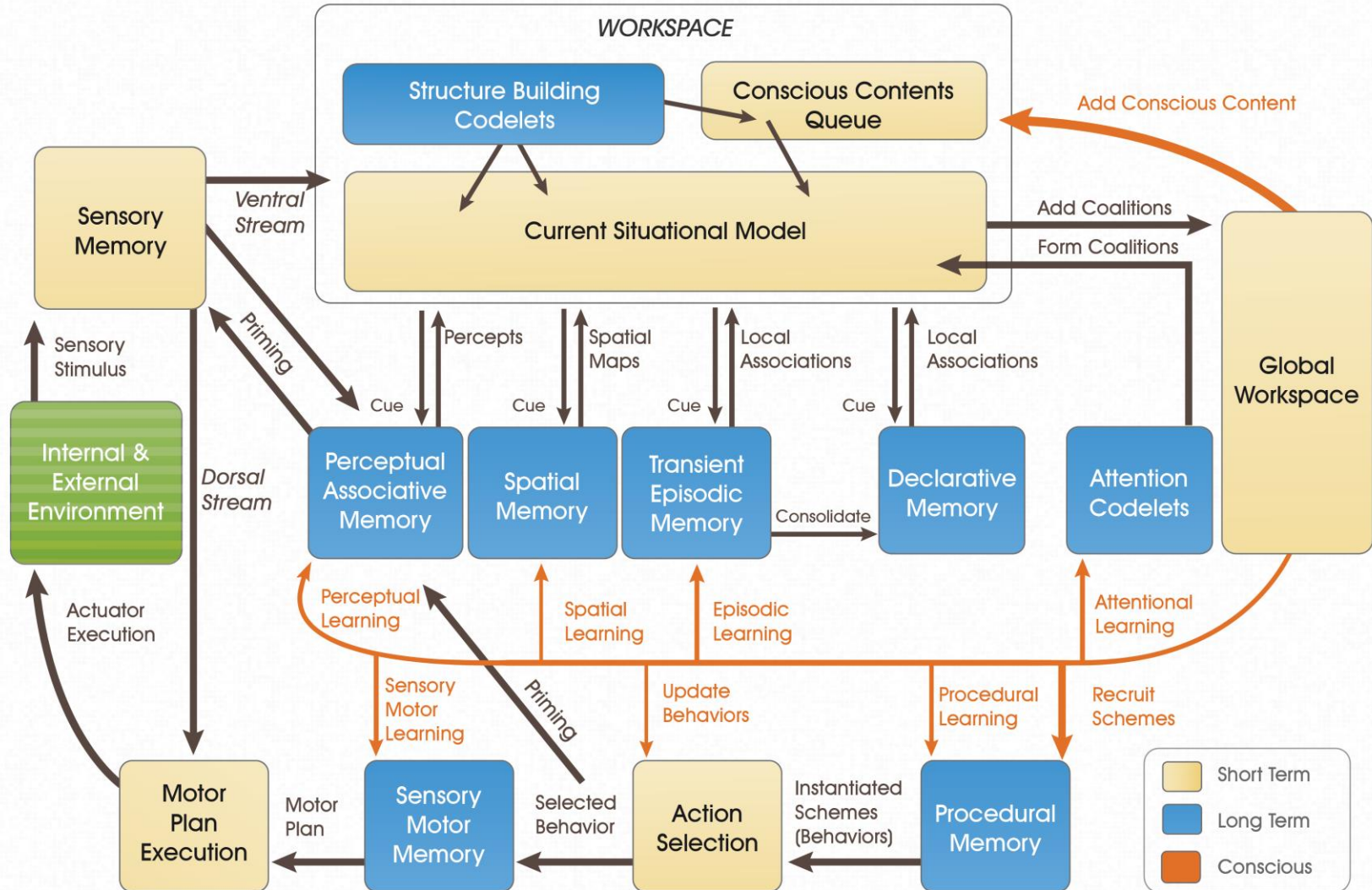
What is the control mechanism: what to do & think next?

Distinction between Program Control and Data

How Learning Happens

LIDA Cognitive Architecture

LIDA Cognitive Cycle



Task: Letter string analogy

abc \Rightarrow *abd*

mrrjjj \Rightarrow ?

Task: Letter string analogy

abc \Rightarrow *abd*

xyz \Rightarrow ?

Copycat Architecture

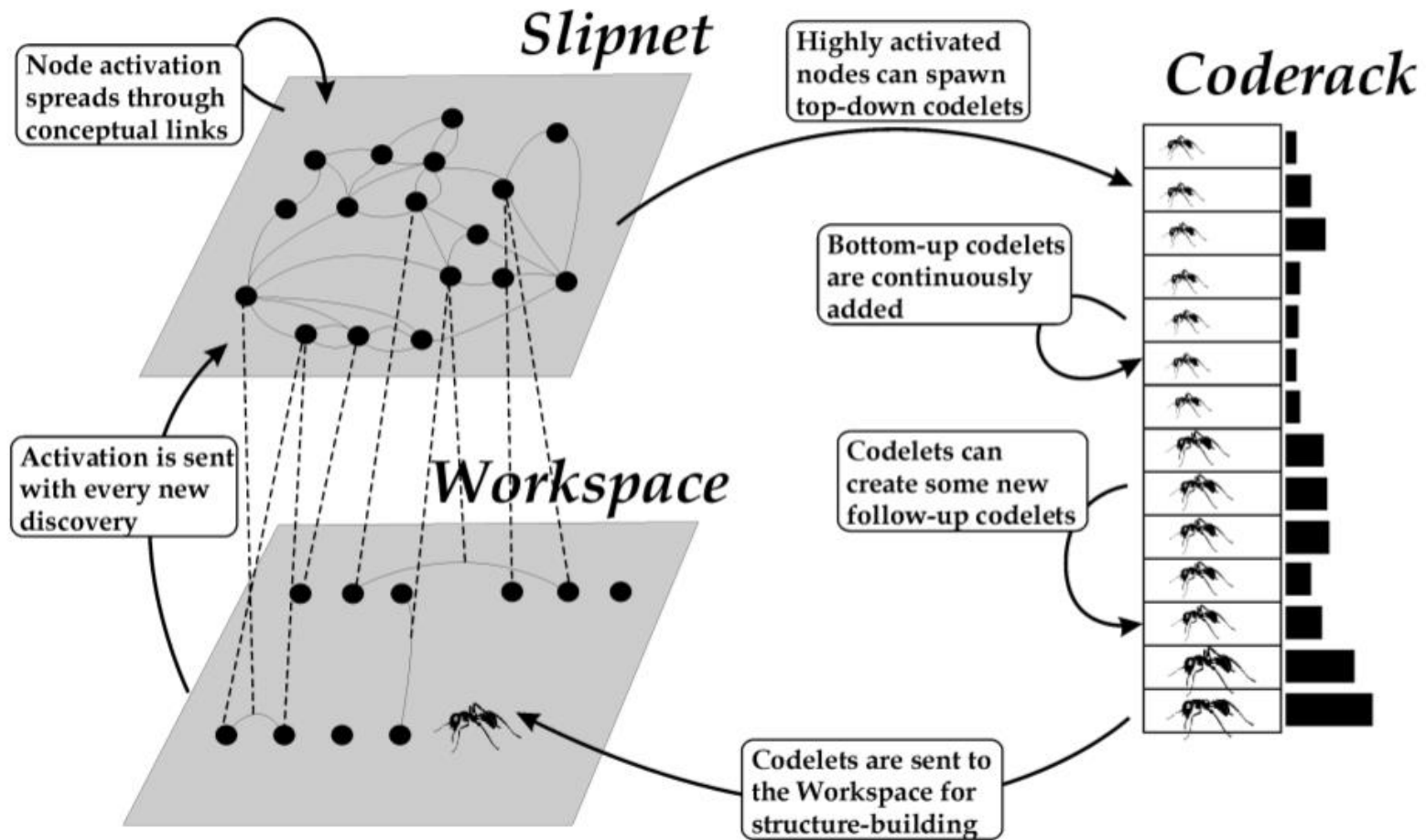
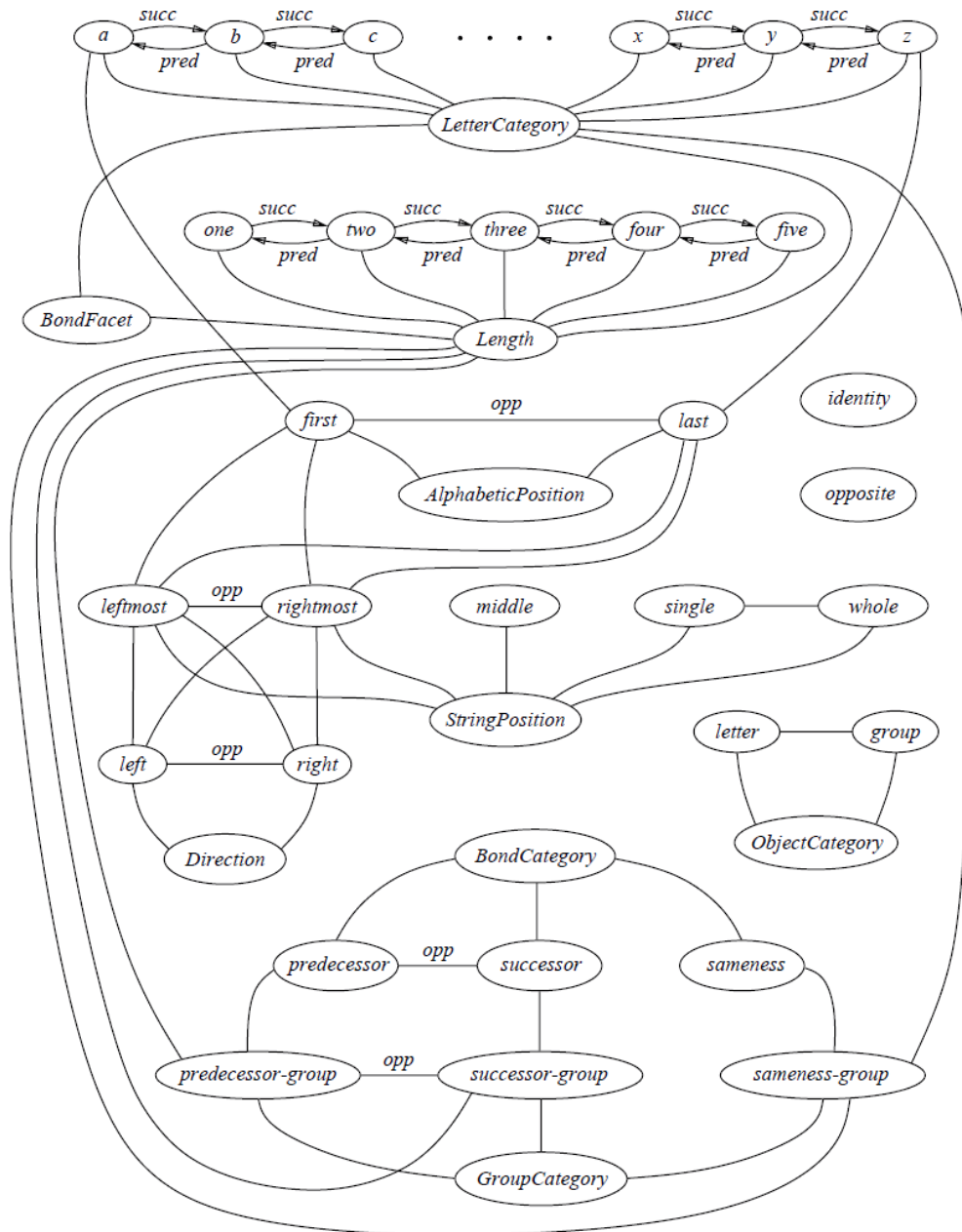


Figure 7: A feedback loop between perceptual and conceptual activity.

Copycat/Metacat Slipnet



Running Metacat

The screenshot displays the Metacat software interface with several windows and a workspace diagram.

Top Panel: Shows "Top Rule" and "Answer mrrjjk".

Episodic Memory: A window showing the transformation: `abc -> abd, mrrjj -> mrrjjk`.

Coderack: A list of codelet types and their selection probabilities. The "Answer finders" codelet is highlighted.

Slipnet Activation: A grid of nodes representing different object categories. The nodes are: Opposite, StringPos, bmost, middle, rmost, whole, single, ObjectCtgy, letter, group, AlphaPos, first, Identity, Direction, left, right, BondCtgy, pred, succ, same, GroupCtgy, predgrp, succgrp, samegrp, LetterCtgy, a, b, c, d, e, f, g, h, i, j, k, l, m, n, o, p, q, r, s, t, u, v, w, x, y, z, Length, one, two, three, four, five, BondFacet.

Workspace: A diagram showing the transformation of the string "abc" to "abd". The letter 'c' is highlighted in red, and a red box indicates the rule: "Change letter-category of rightmost letter to successor". Below this, the string "mrrjj" is shown with boxes around the 'r's and 'j's. A blue box indicates the rule: "Change letter-category of rightmost letter to successor".

Metacat Control Panel: A window with a menu (Help, Demos, Windows, Options, Clear Memory), the text "abc -> abd; mrrjj -> ? seed: 187793084", and buttons for "Step", "Go", "Stop", and "Reset".

Text Panel: A pink box containing the text: "Okay, if 'abc' changes to 'abd', what does 'mrrjj' change to? Hmm... The answer 'mrrjjk' occurs to me. I think this answer is pretty good."

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Conversational Agent

Question answering task

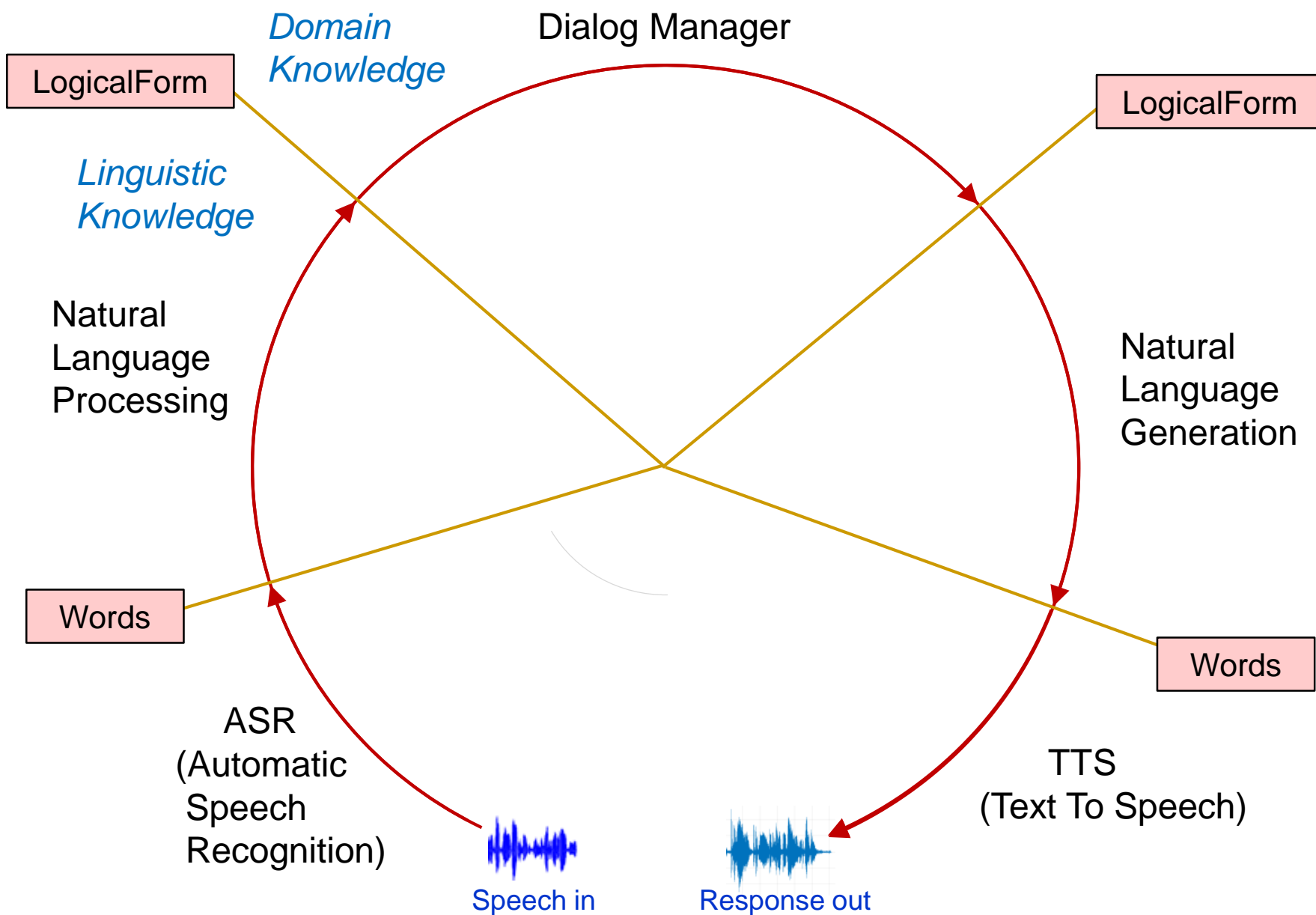
“Alexa, who won the 1934 world series?”

“The Saint Louis Cardinals beat the Detroit Tigers 4-3 in the 1934 World Series.”

“Alexa, what is Sam Taylor’s phone number?”

“Sam Taylor’s telephone number in Boston is area code six-five-eight, one two three, four five six seven.”

Architecture of a Conversational Agent



NLP: From Words to LogicalForms (Meaning)

“What is Sam Taylor’s phone number?”

“What is the number for Sam in Boston?”

“Tell me Sam Taylor’s telephone number.”

“What do I dial to reach Sam Taylor?”

Question-word

Command-word

Person-word

Location-word

Address-book-item

Intent terms

Entity terms

Syntactic Pattern:

{Question (WhatIs)} {Person}’s {X} ?

LogicalForm:

Query (AddressBookItem (Person (name “Sam Taylor”),
(telephoneNumber ?)))

NLP Pipeline Elements

Theoretical,
Comprehensive

Co-Reference Resolution

Part-of-Speech Tagging

Sentence Parsing

Stemming

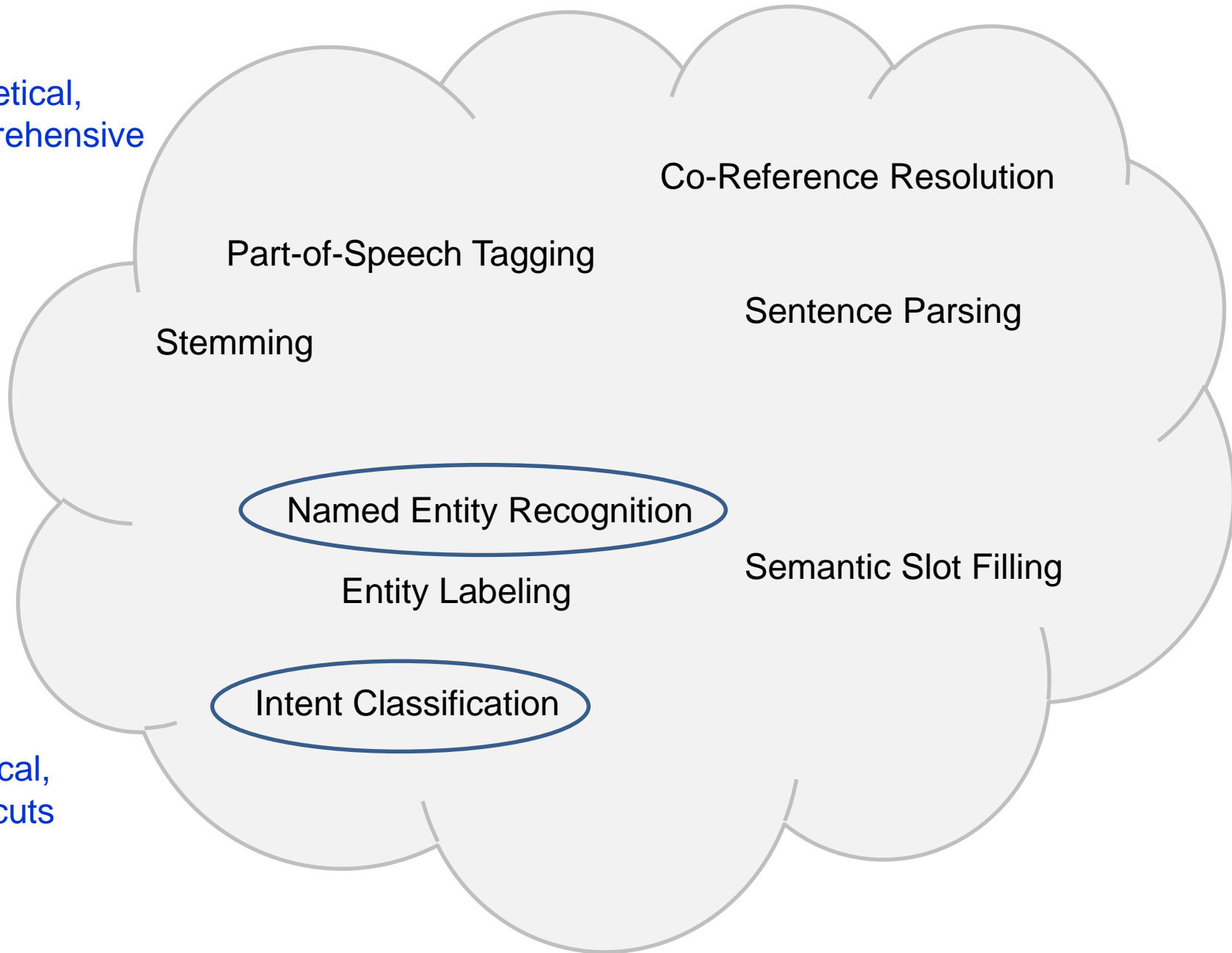
Named Entity Recognition

Semantic Slot Filling

Entity Labeling

Intent Classification

Practical,
Shortcuts

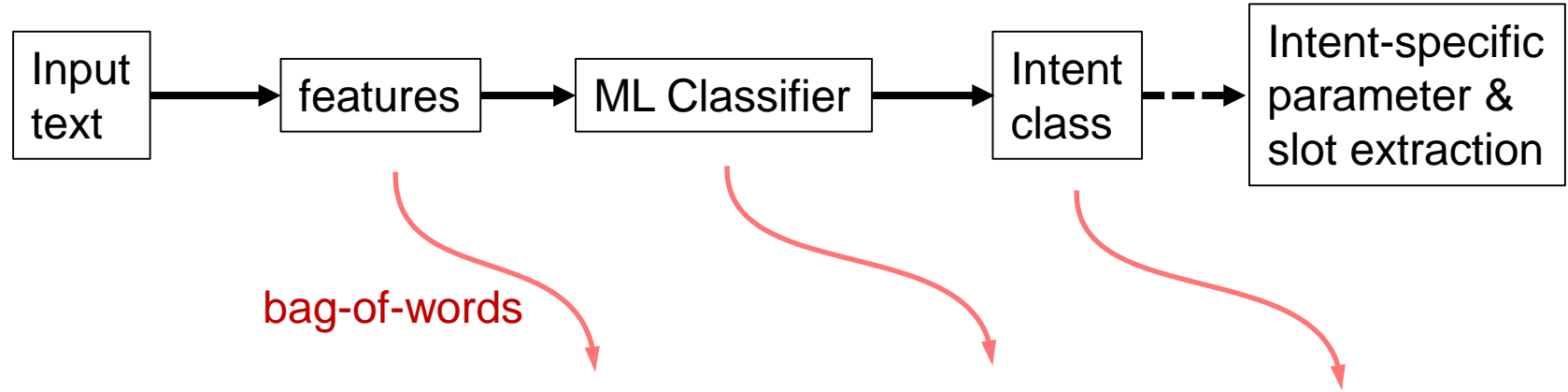


Intent Classification

Practical but Limited

What is Sam Taylor's phone number?

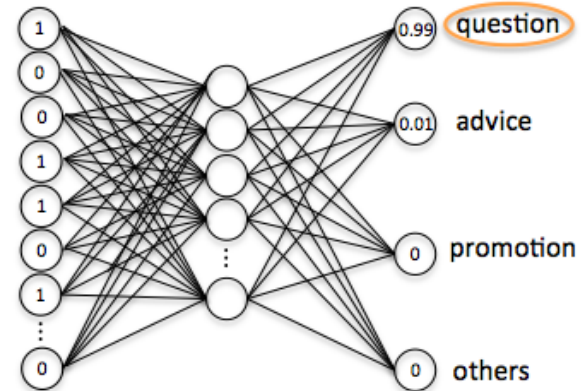
Data-Value-Question



bag-of-words

Mau tanya master.. Pupuk organik cair apa yang n lengkap menurut pengalaman teman2.?

pupuk	1
bercak	0
padi	0
pengalaman	1
apa	1
air	0
lengkap	1
...	...
pestisida	0



NER: Named Entity Recognition

“What is John’s address in Garden Grove?”

1. common nouns:

garden

grove

2. but uncommon to see the bi-gram,
.....“garden grove”...

3. City database:

“Garden Grove, California”

4. Lexical environment

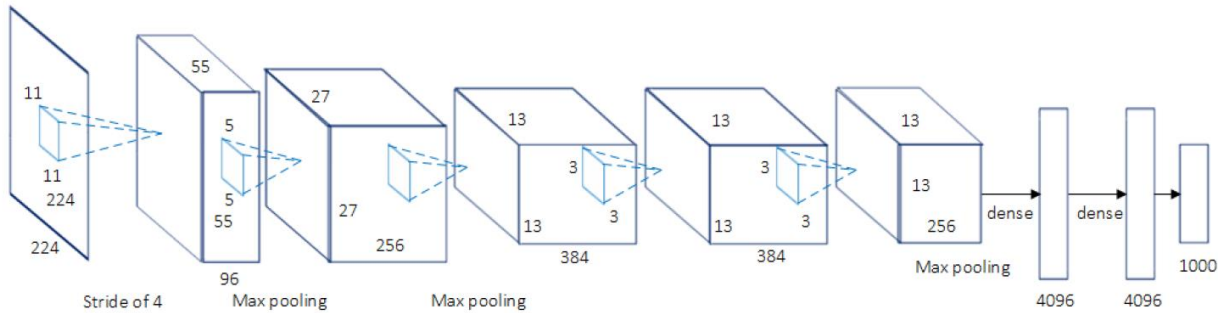
address in { location }



- rules
- standard ML

Symbol Processing in Neural Networks

Input type:
vector



Output type:
vector

Can Neural Networks Produce Symbolic Structures from Sentences?

Input type:

Words = sequence of symbols

“What is Sam Taylor’s phone number?”

Output type:

Logical Form = hierarchical tree of symbols

Query (AddressBookItem (Person (name “Sam Taylor”),
(telephoneNumber ?)))

Word Vector Embedding

Purpose:

- lower dimensionality reduces data sparsity for learning
- similar words :: similar vectors leads to improved generalization

word vocabulary size

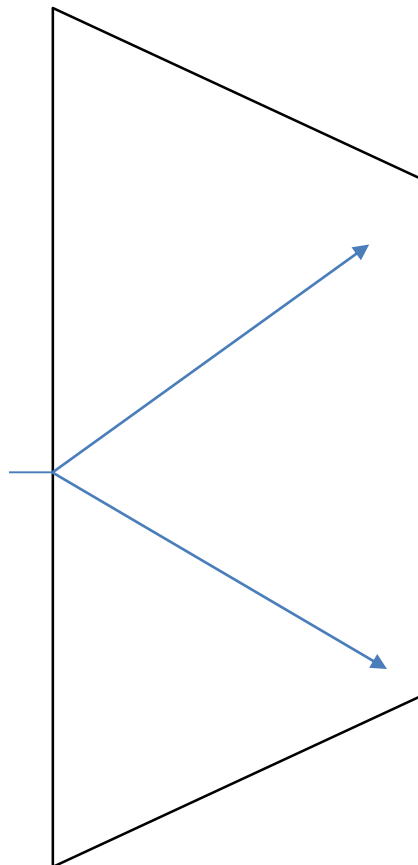
e.g. 10k

embedding vector size

e.g. 300

what
why
who
.
.
is
was
are
.
.
phone
telephone
.
.
<person>
<location>
.
.
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0
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1
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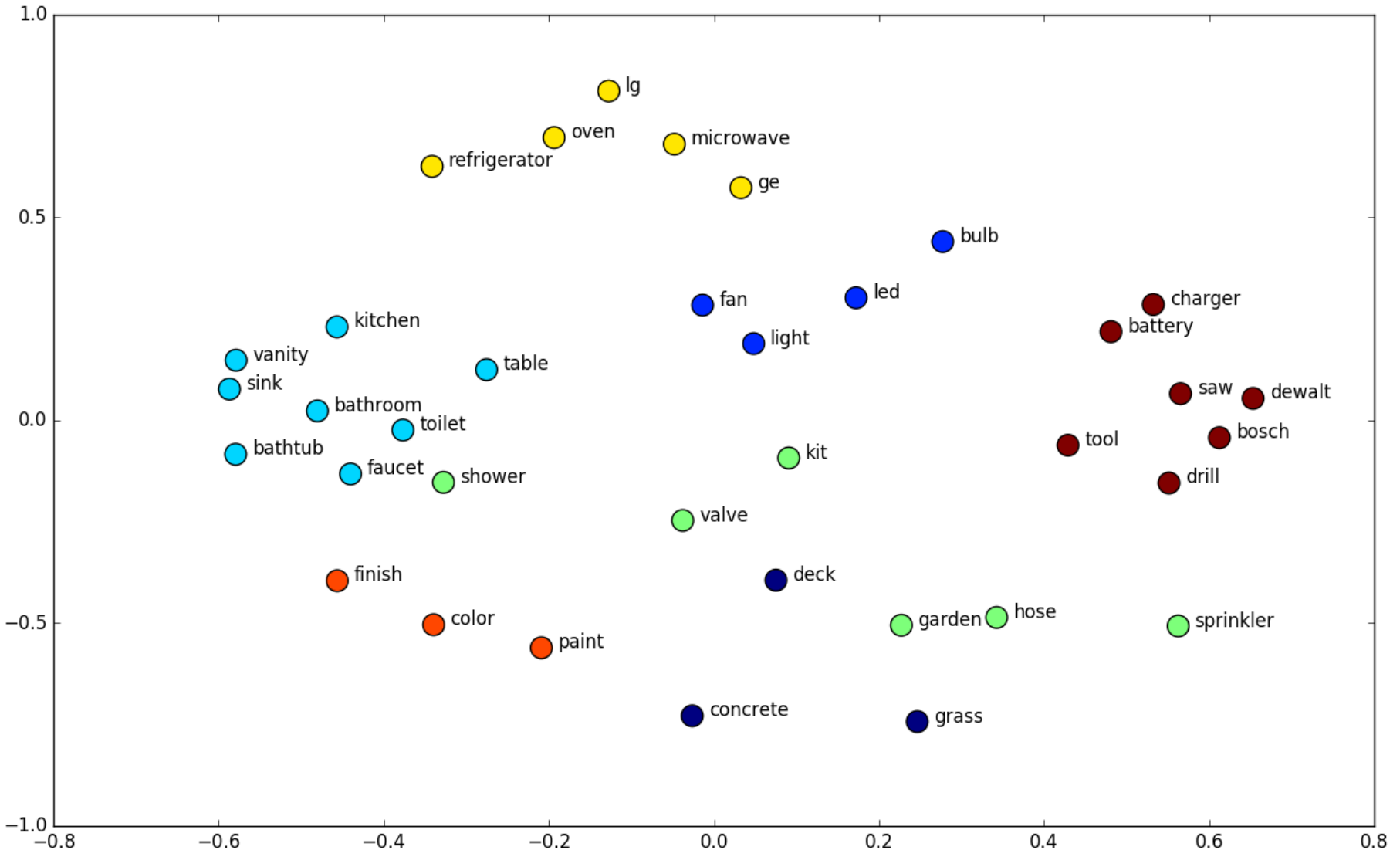


.2
-.3
.1
.04
-.6
-.7
.15
.
.

Word Vector Embedding

Purpose:

- lower dimensionality reduces data sparsity for learning
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NLP: From Words to LogicalForms (Meaning)

“What is Sam Taylor’s phone number?”

“What is the number for Sam in Boston?”

“Tell me Sam Taylor’s telephone number.”

“What do I dial to reach Sam Taylor?”

Question-word

Command-word

Person-word

Location-word

Address-book-item

Intent terms

Entity terms

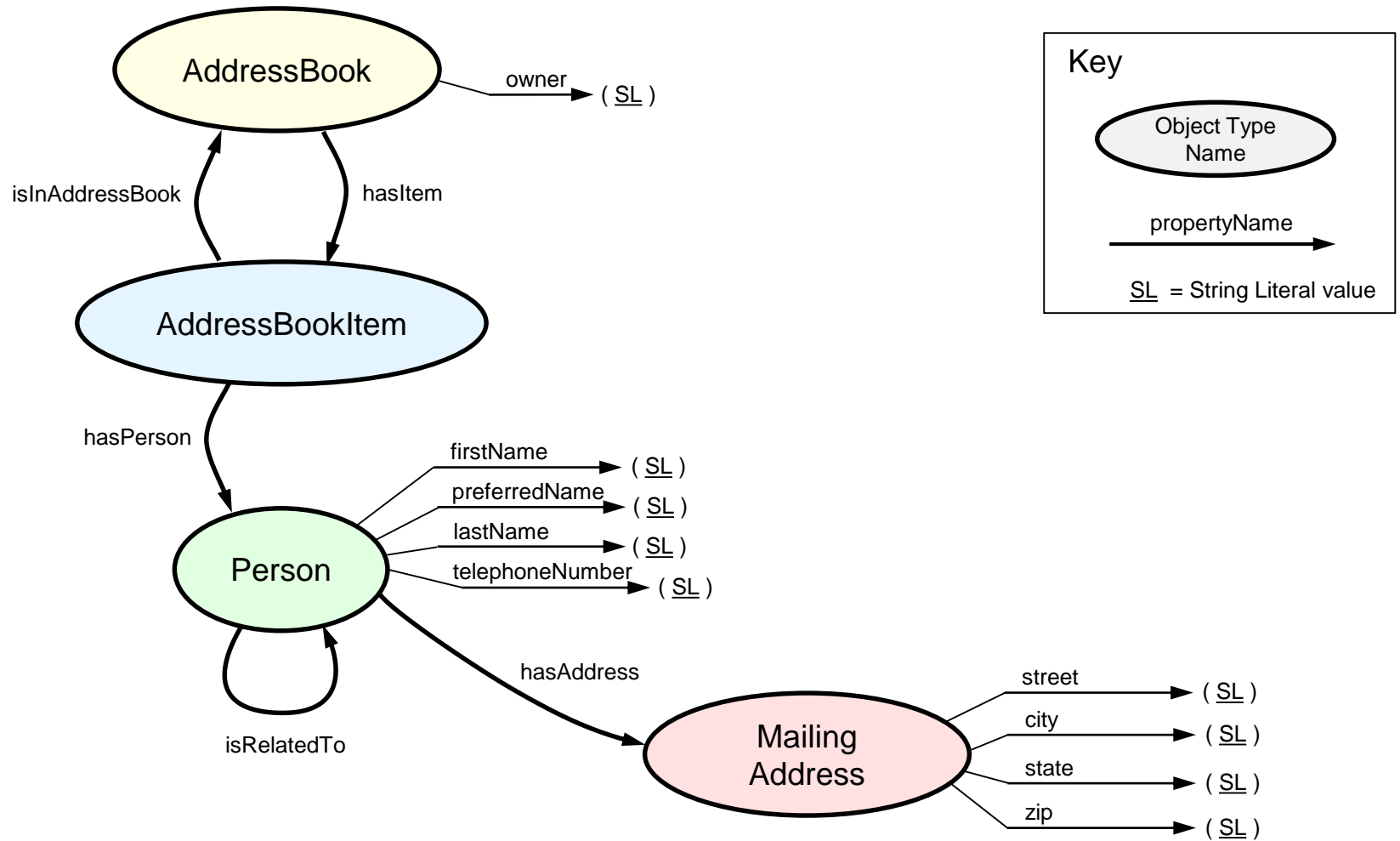
Syntactic Pattern:

{Question (WhatIs)} {Person}’s {X} ?

LogicalForm:

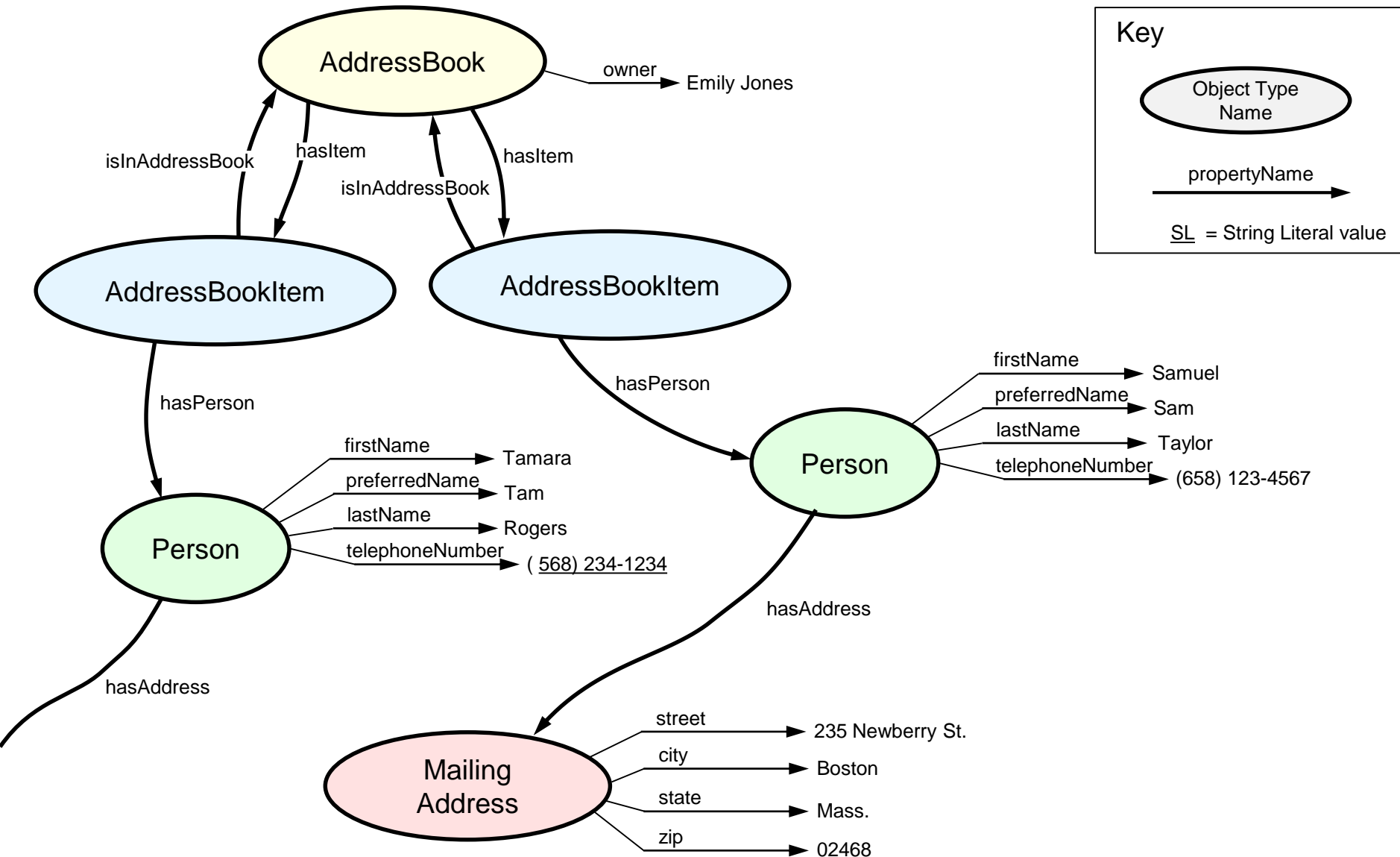
Query (AddressBookItem (Person (name “Sam Taylor”),
(telephoneNumber ?)))

Address Book Knowledge Base: Ontology Schema



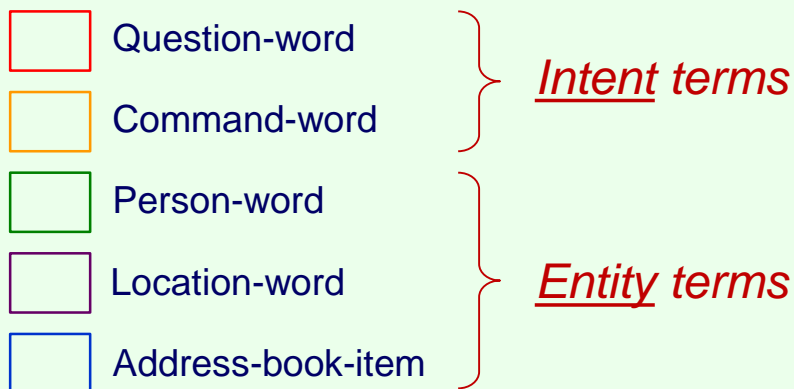
Address Book Knowledge Base: Data Example

Knowledge Graph



NLP: From Question to KB Query

“What is Sam Taylor’s phone number?”



Syntactic Pattern:

{Question (WhatIs)} {Person}'s {X} ?

LogicalForm:

Query (AddressBookItem (Person (name “Sam Taylor”),
(telephoneNumber ?)))

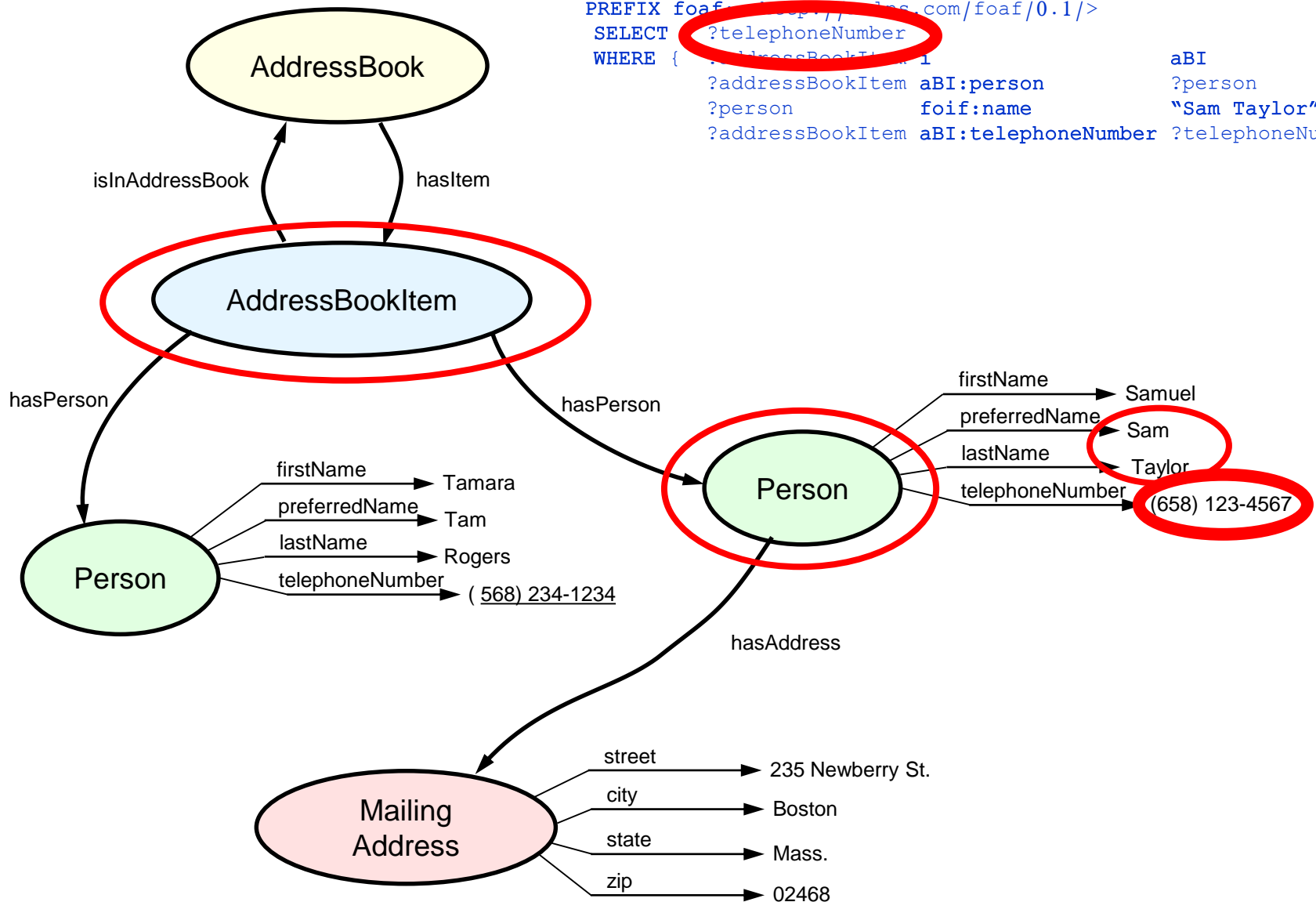
SPARQL KB Query:

```
PREFIX foaf: <http://xmlns.com/foaf/0.1/>
SELECT ?telephoneNumber
WHERE {
  ?addressBookItem i aBI .
  ?addressBookItem aBI:person ?person .
  ?person foif:name "Sam Taylor" .
  ?addressBookItem aBI:telephoneNumber ?telephoneNumber }
```

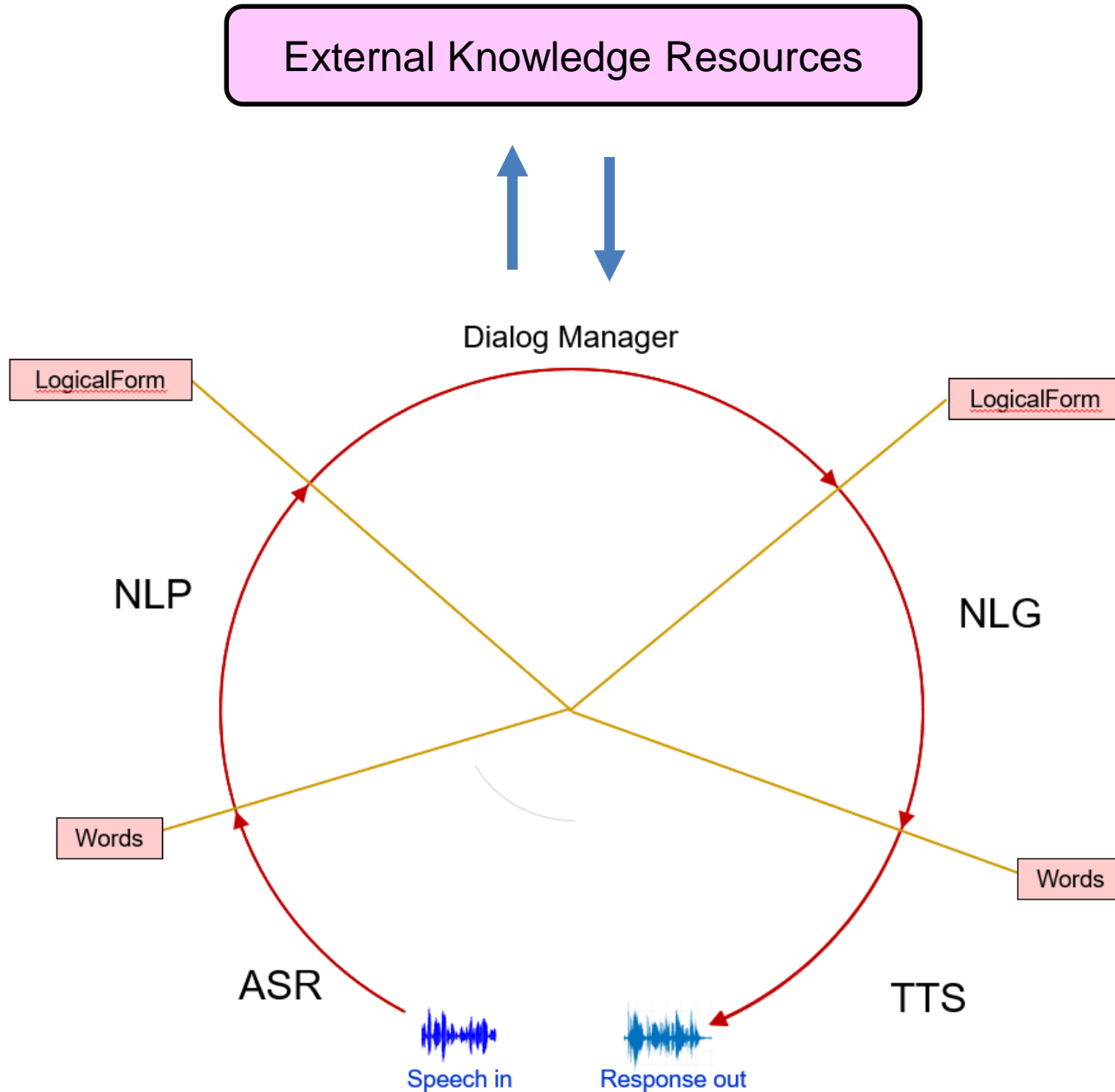
Query Match to Address Book Knowledge Base

SPARQL KB Query:

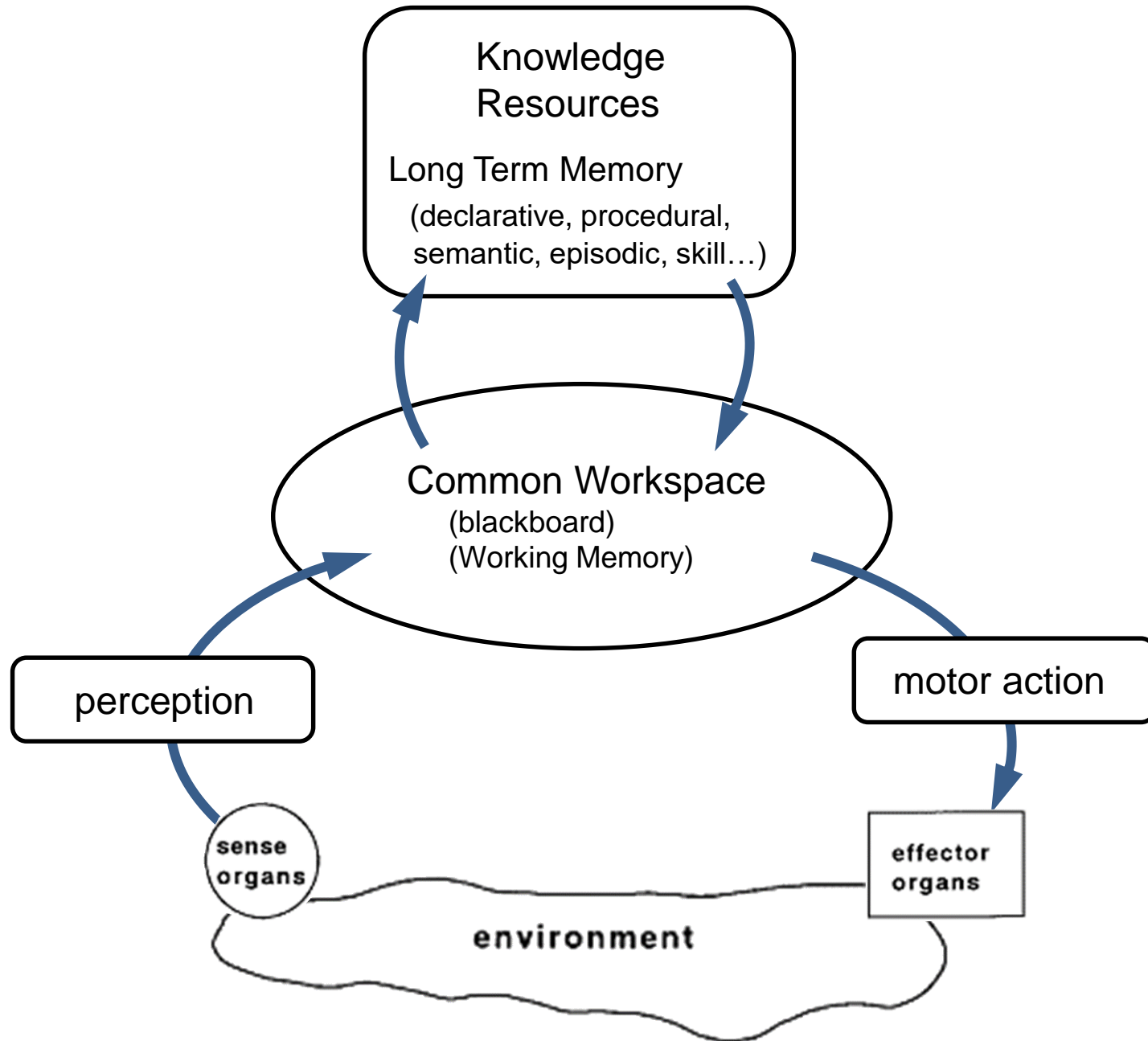
```
PREFIX foaf <http://xmlns.com/foaf/0.1/>
SELECT ?telephoneNumber
WHERE {
  ?addressBookItem foaf:person aBI:person .
  ?person foif:name "Sam Taylor" .
  ?addressBookItem aBI:telephoneNumber ?telephoneNumber .
}
```



Architecture of a Conversational Agent



Cognitive Architecture: Standard Model



Why Are Conversational Agents So Dumb?

“Alexa, who won the 1934 world series?”

“The Saint Louis Cardinals beat the Detroit Tigers 4-3 in the 1934 World Series.”

“Alexa, who was the starting pitcher for the Tigers?”

“The probable pitcher for the Tigers on Saturday February 23 has not been announced yet.”

“Alexa, who won the 1934 world series?”

“The Saint Louis Cardinals beat the Detroit Tigers 4-3 in the 1934 World Series.”



“Alexa, who was the president then?”

“This might answer your question. The president of the United States is Donald Trump.”

Needed: A cognitive architecture with contextual memory.

Contextual Memory in Conversational Agents (2019)



- Remember past Intents, Entities

- Query terms:

- baseball
 - world series
 - 1934

- Response terms:

- Saint Louis Cardinals
 - Detroit Tigers
 - 4-3

 Alexa
 Google Home

“Who was the manager of the Cardinals?” (St. Louis Cardinals, in 1934)

- Dialog State Tracking

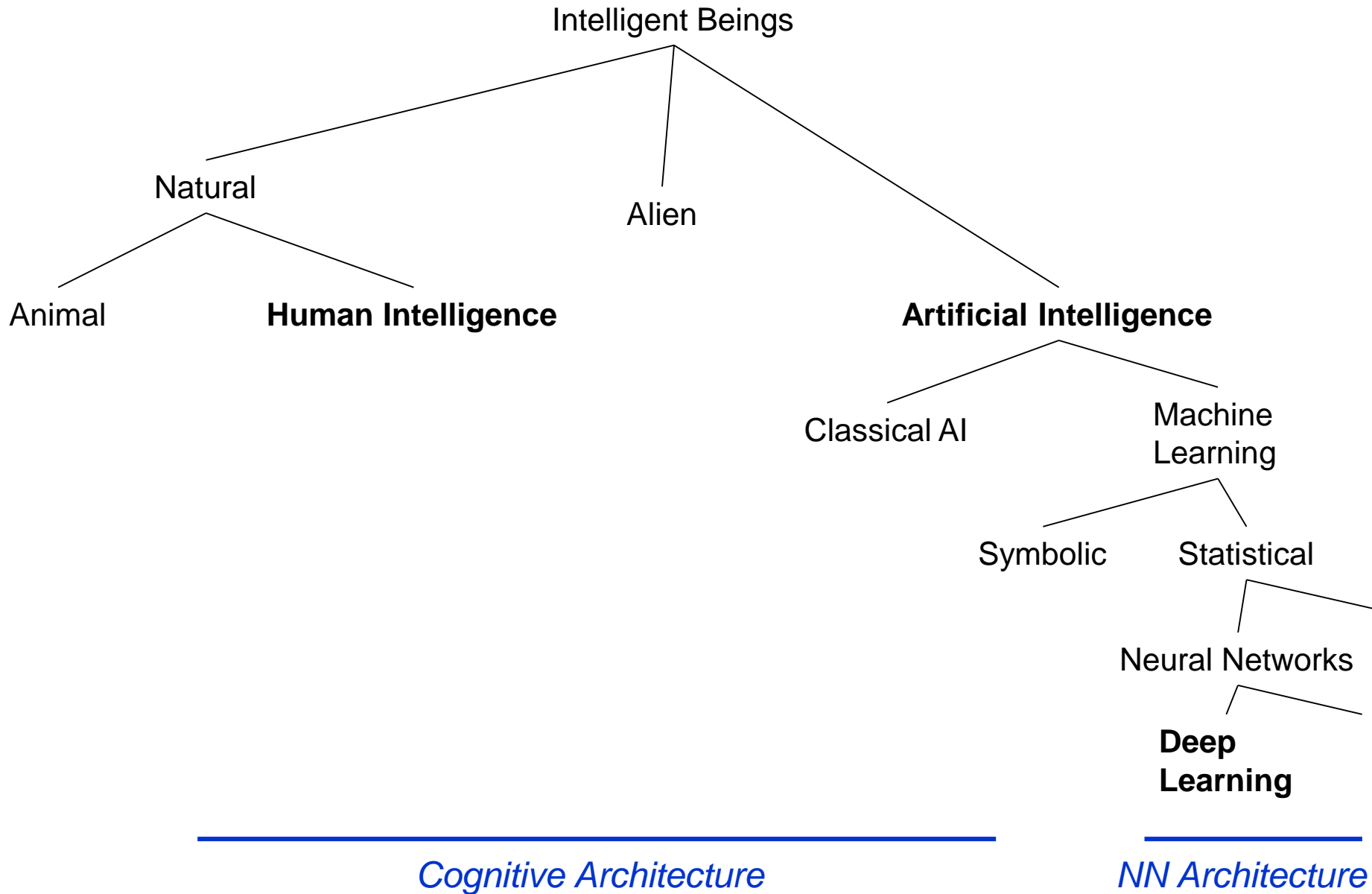
-slot filling for restricted tasks, e.g. ordering pizza

“Pizzabot I want a pepperoni pizza with tomato sauce”

size	?
sauce	tomato
topping1	pepperoni
topping2	X
topping3	X

“Sure, what size?”

Summary: Taxonomy





Eric Saund

- *Research scientist in Cognitive Science and AI.*
- *Conversational Agents, Visual Perception, Cognitive Architectures.*
- *I build stuff.*

Projects

Papers

Curiosities

Links

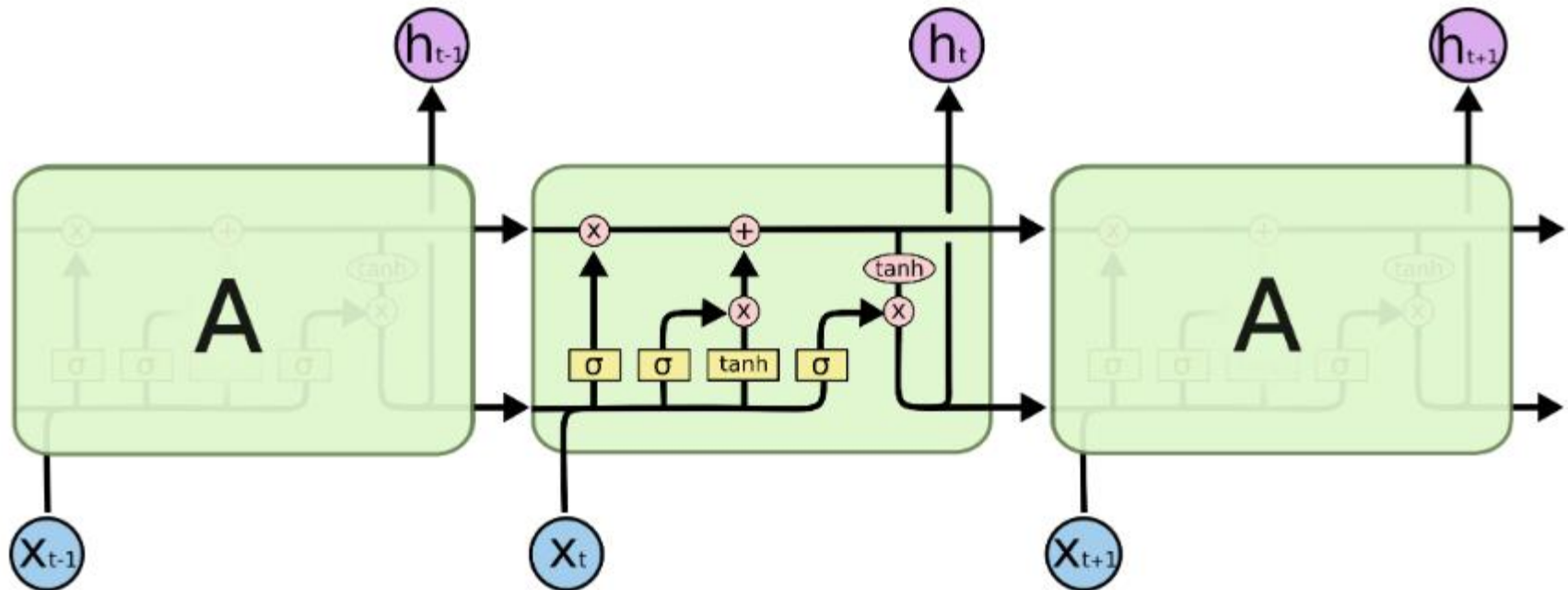
Contact

<http://www.saund.org>

saund@alum.mit.edu

Conversation

LSTM Module



Deep Learning Architecture for Parsing to Frame Graph

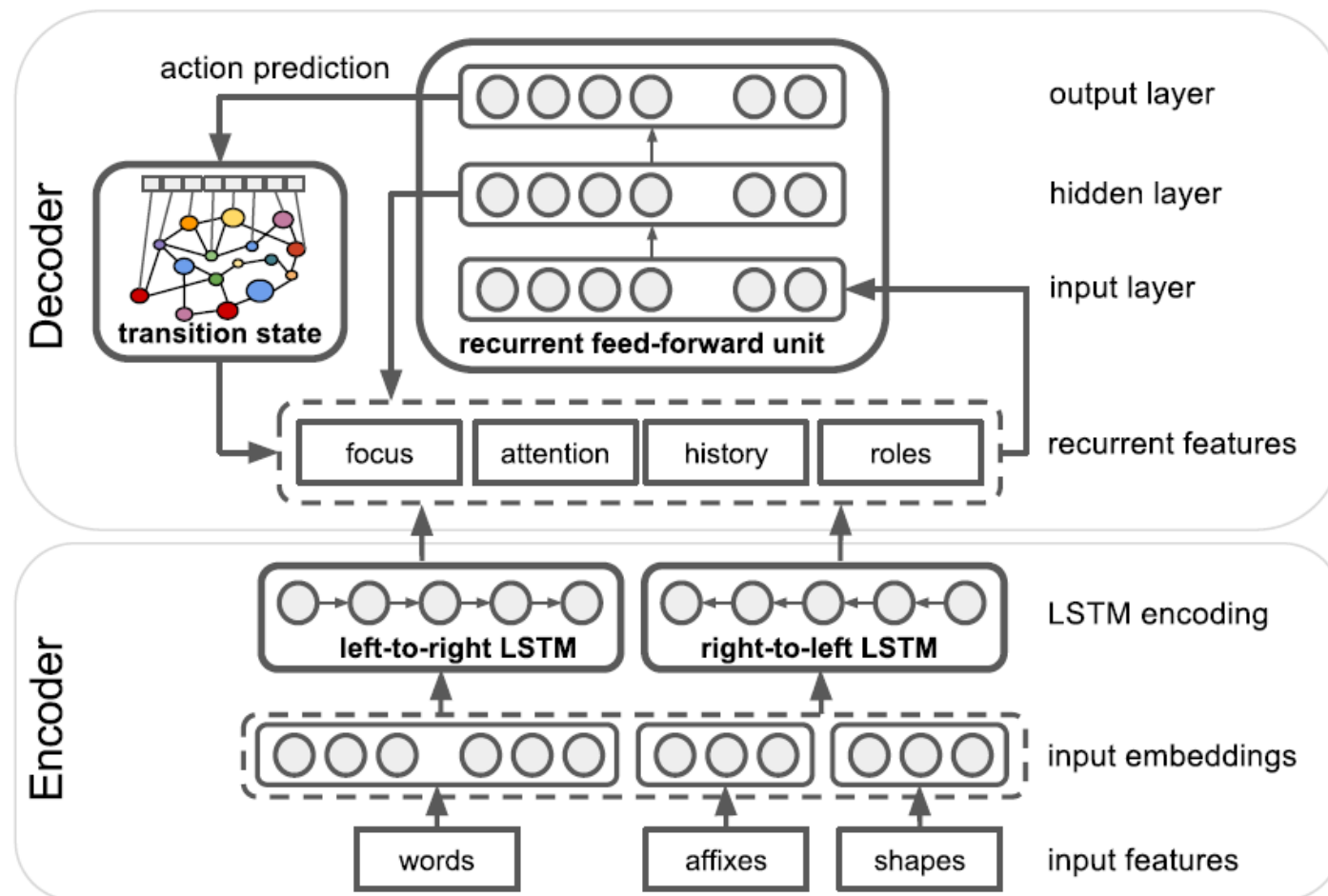


Figure 1: Neural network architecture of the SLING parser. The input is encoded by a bi-directional LSTM and fed into a recurrent feed-forward (FF) unit that proposes transition system actions. The hidden layer activations and the transition system state are combined to create the input feature vector for the next step. The FF unit is run repeatedly until the transition system has reached a final state.