

Matching utterances with visual scenes: neurocomputational investigation of the language-vision interface $\overline{ ext{USC}}$

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Introduction

We propose a framework that addresses the question of how language and visual processes are integrated, allowing us to communicate about the world we perceive.

In particular, we ask the question: what type of neurocomputational architecture could simulate a sentence-picture matching task (spmt)?

Multiple cooperating functional routes:

Neuropsychology: (1) Aphasics that are impaired in using grammatical cues can still use world knowledge (WK) cues during spmt¹. (2) Some aphasics The man kicked the bucket seem to be specifically impaired in their capacity to process grammatically relevant semantic cues².

Abstract Neural Net

Hidden units OOO copy

Output units OOOO

Input units OOOO

Psycholinguistics: Studies using the Visual World Paradigm (VWP) showed that WK, visual, and linguistic processes are dynamically and incrementally coordinated³.

Sentence-Picture Matching Task

Many *neurolinguistic models* have put forward multi-streams architectures for the language system with a tendency to distinguish between (1) a more semantic/heuristic stream from (2) a more grammatical/algorithmic stream. Most do not offer an explicit computational account of how processes distributed in those streams dynamically cooperate to support our linguistic performances.

Goals

Neural Circuit

Offer an explicit computational model of how visual, world knowledge, and grammatical processes, distributed over a multiple functional routes architecture, can dynamically cooperate during a sentence-picture

Localist Neural Net

Output CAT FIT MAT FAT

ut /k/ /v/ /m/ /f/ /t/ /a/ /i/

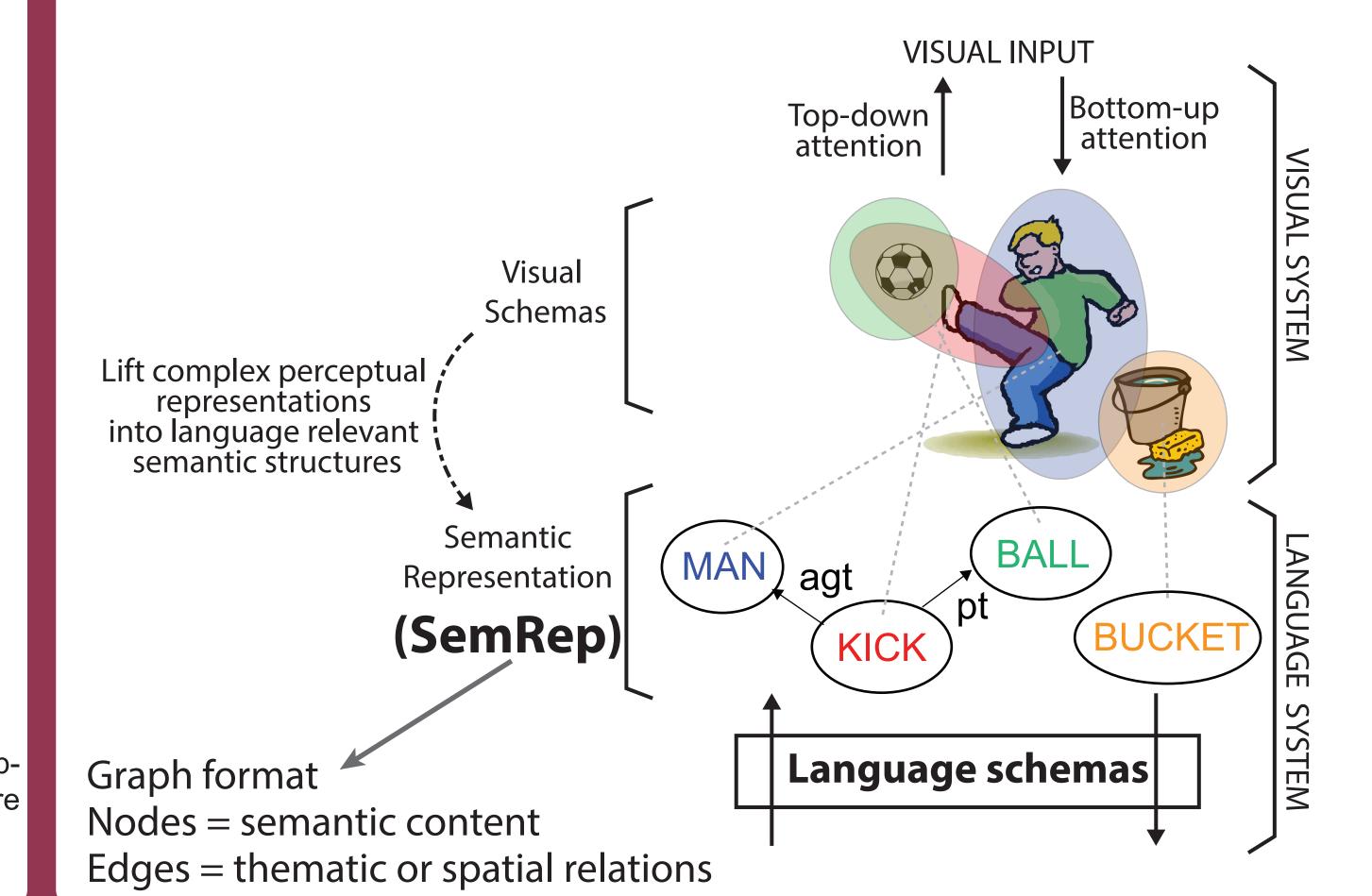
Schema model

Neural Schema model

Explicit symbolic computations

Distributed dynamical systems

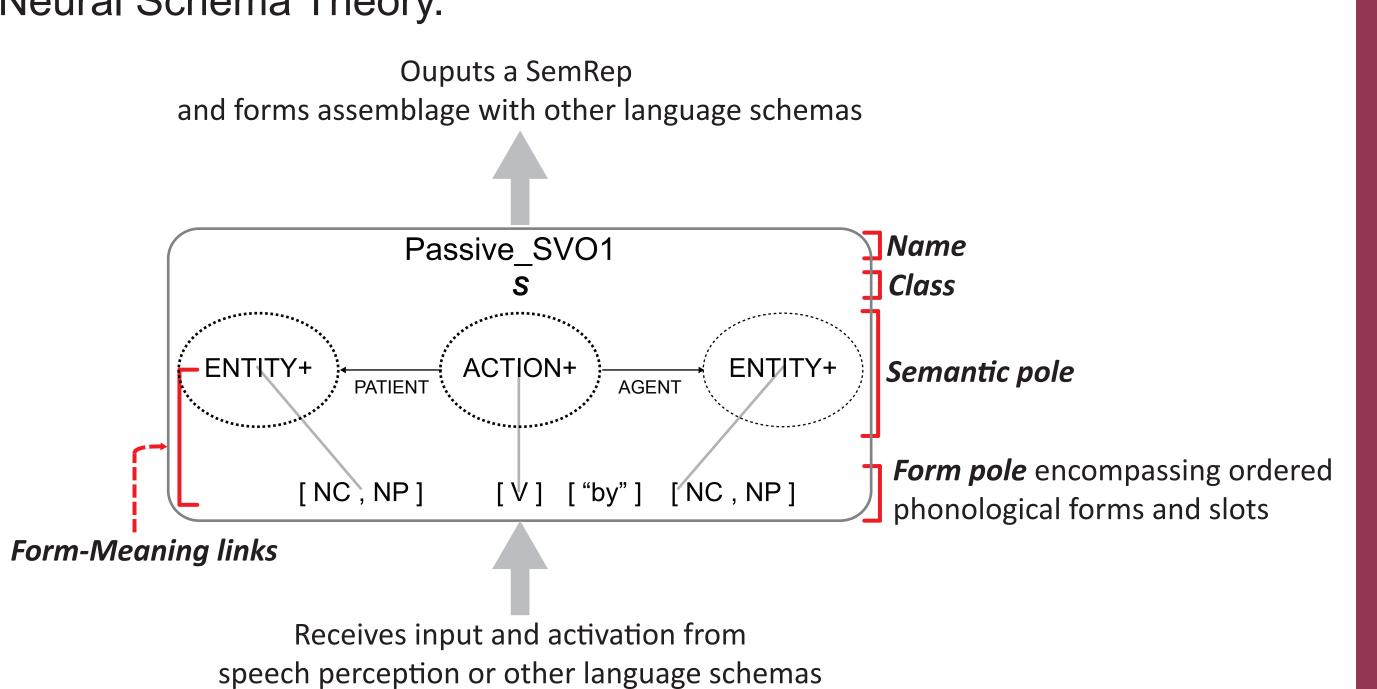
SemRep: handling the vision-language interface



Language schemas

In the Template Construction Grammar framework (TCG) language schemas are defined as **constructions** mapping form onto

Language schemas inherit the computational properties defined by Neural Schema Theory.



Grammatical Knowledge

Network

World Knowledge

Event/Object/Action

Low-level visual processing

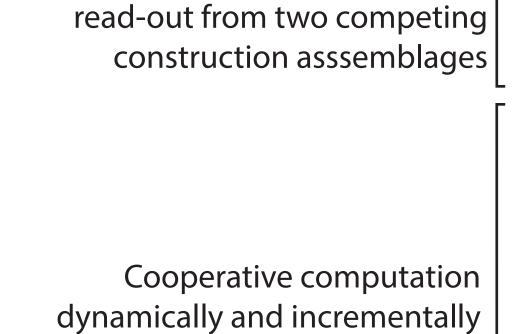
Attention-guidance systen

Open-class words

Visual

An explicit model of grammatical processing

Construction assemblages compete to interprete the expression "kick the bucket" (transitive construction vs. idiom). A visual scene could however set semantic expectations that would strongly bias the competition in favor of one of the alternative interpretations and disambiguate the SemRep early on in the comprehension process



Lexically anchored constructions matchine

MAN1 KICK1 Semantic representations (SemRep) <u>man</u> <u>kicked</u> generates construction assemblages DIE2 that map form onto meaning BUCKET1 (processing of articles not shown) KICK BUCKET

Incrementally received input

the ... man ... kicked ... the ... bucket

matching task.

Compare the effect of simulated lesions against the degradations of performances in agrammatics.

Modeling Framework: Neural Schema Theory

Brain Anchored Neural Net

Language schemas are defined as constructions mapping form onto

A two-route model of language comprehension

Comprehension is modeled as a cooperative process involving multiple functional routes that dynamically and incrementally build the SemRep.

Grammatical route (G)

Processes all linguistic inputs by incrementally building construction assemblage.

Cooperative computation between constructions involves both form and meaning (light semantics) matching.

Heavy Semantics route (HS)

Processes open-class words which are directly associated with SemRep nodes without relying on grammatical knowledge. The world knowledge (heavy semantics) dynamically and incrementally enriches the SemRep.

Conclusions and future directions

We offer an explicit model of how multiple functional routes can dynamically cooperate to interpret linguistic inputs. Our model, if it incorporates an explicit model of grammar, focuses on the question of how the language system generates semantic representations that can be matched against those generated by the visual system. The model can be used to quantitatively reframe competing theories of agrammatism (in particular thematic role assignment deficits⁶ and working memory deficits⁷ accounts). However the model also insists on the core role that complex coordinated dynamics play in comprehension and its degradation, an issue that is often insufficiently captured by non-computational "multi-streams" accounts.

Future work will address in more details the role of the visual route, harnessing existing neurocomputational models of bottom-up and top-down visual attention.

By varying the computational resources available to each routes and time pressure, we will also investigate how our model can simulate "good-enough" comprehension in normal subjects.

References

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[4] Piaget, J. (1965). The stages of the intellectual development of the child

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adults perform like aphasic patients. Acknowledgement

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meaning. **Neural Schema Theory**

Expanding on the notion of schema put forward by Piaget⁴, schema theory⁵ offers a computational framework to explicitly simulate how schemas can cooperate to organize the behavior of an organism. Schema

- Encapsulates a function (learned or innate) - Receives inputs and generates outputs

- As an activation level that reflects the relevance of its function given the current (internal and external) state of the organism. - Can in theory be broken down into subschemas when more is known about its function (system-of-systems approach). LTM and instantiation

Learned perceptual, motor, and semantic knowledge is represented as a schema network. When activated, schemas are instantiated in working memory where they remain active as long as they are relevant to the ongoing behavior.

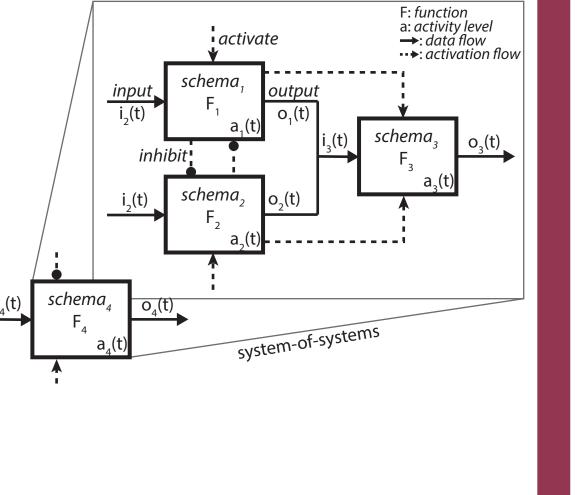
Cooperative computation

Schemas compete and cooperate to form schema assemblages. Cooperating schemas reinforce each-other's activation levels, while competing schemas inhibit each-other. These assemblages form flexible and distributed control structures that adaptively organize how information is processed by the organism.

From schema theory to neural schema theory A schema model becomes part of neural schema theory when it makes contact with brain data such as:

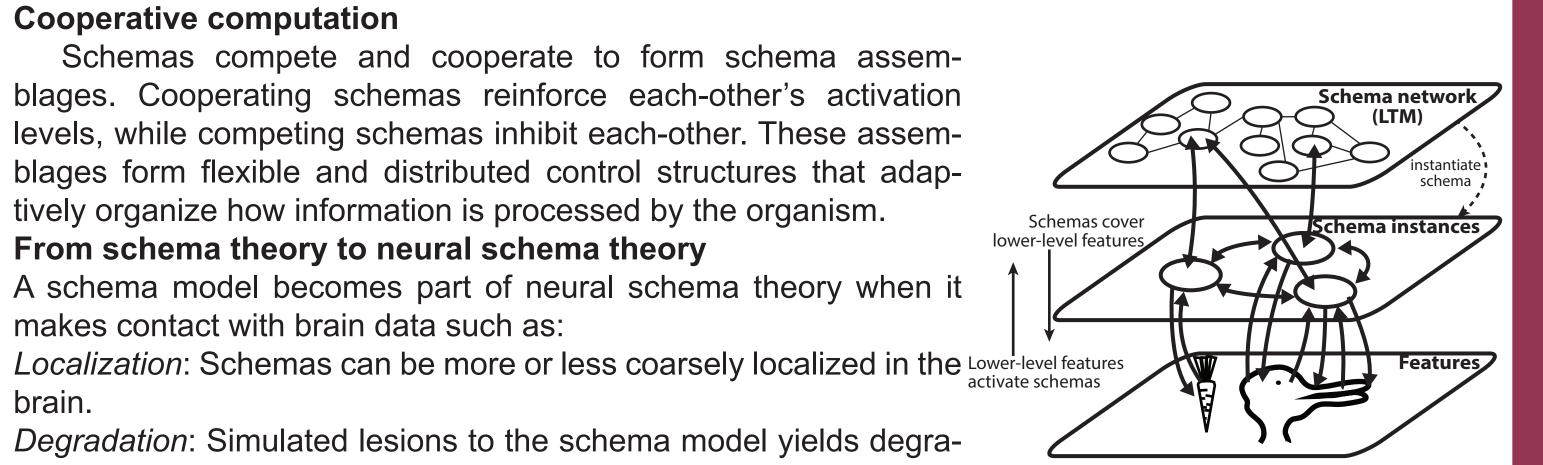
Degradation: Simulated lesions to the schema model yields degra-

dations patterns matching neuropsychological data.



TOP-DOWN

Algorithmic model



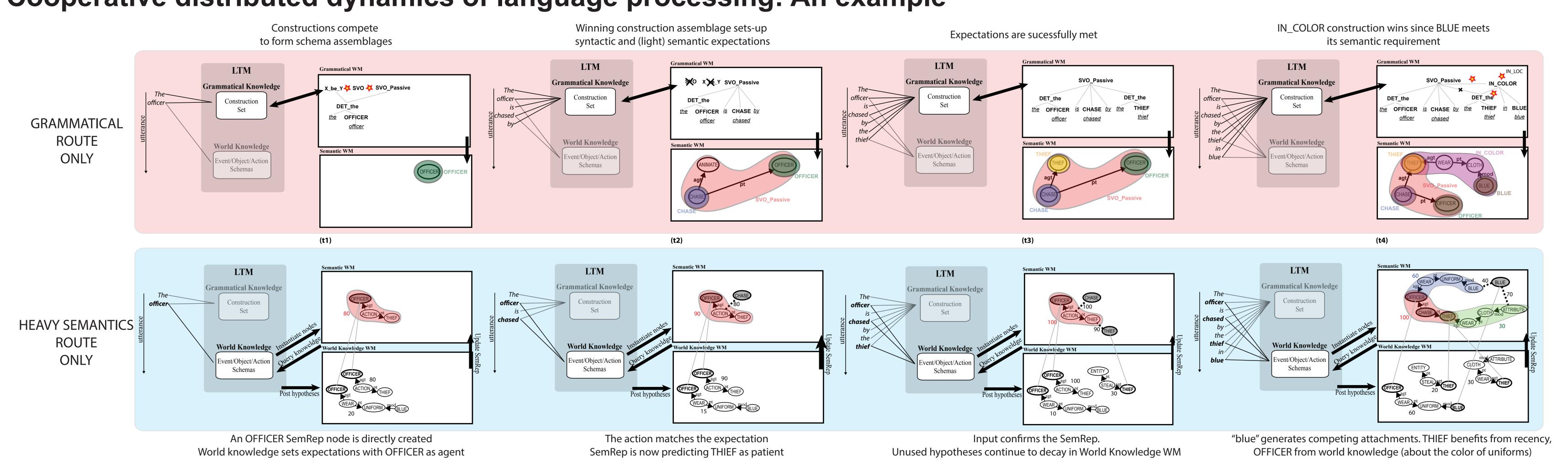
Cooperative distributed dynamics of language processing: An example

SemRep

Visual WM

(Scene Interpretation Network)

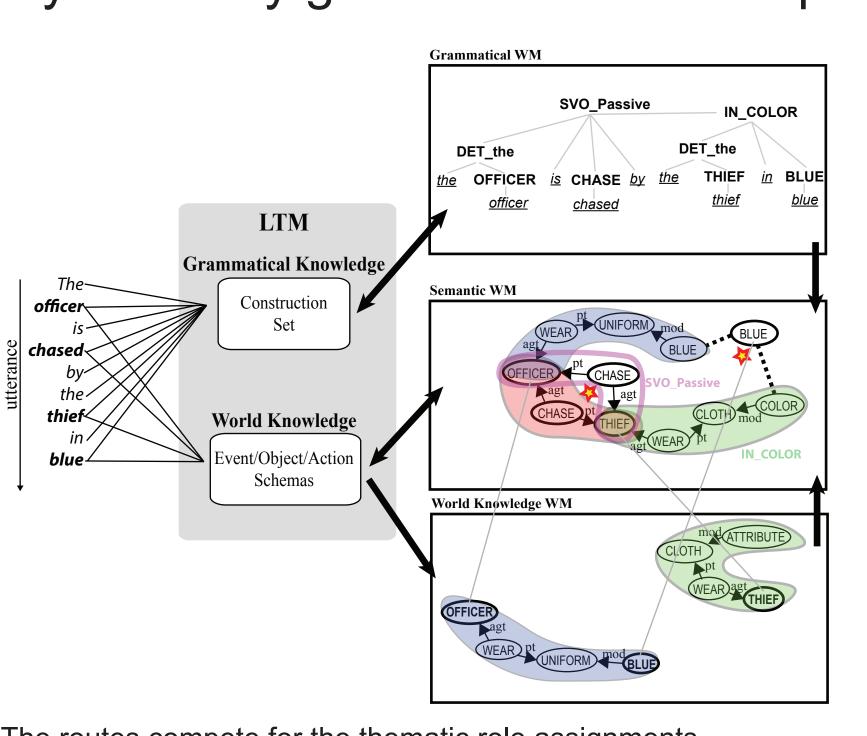
Schema-labeled regions



Grammatical WM

World Knoweldge WM

At each time step, the two routes compete and cooperate to dynamically generate the SemRep



The routes compete for the thematic role assignments of OFFICER and THIEF.

The grammatical route disambiguates the semantic interpretations of BLUE generated by the heavy-semantics route.