



Natural Language Processing

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CS 365 Artificial Intelligence

Two views of Grammar

The question "How can you construct a grammar with no appeal to meaning" is wrongly put, since the implication that obviously one can construct a grammar with appeal to meaning is totally unsupported.

- Chomsky, Syntactic Structures 1957, p.93

[Cognitive grammar] takes the radical position that grammar reduces to the structuring and symbolization of conceptual content and thus has no autonomous existence at all.

- Langacker, Grammar and Conceptualization, 2000 ,p.3

Autonomy of Syntax

1. Colorless green ideas sleep furiously.
2. Furiously sleep ideas green colorless.

Both are meaningless yet we can judge 1 as grammatical and 2 as ungrammatical.

Hence syntax is independent of meaning.

Probabilistic Grammar

$S \rightarrow NP VP [1.00]$

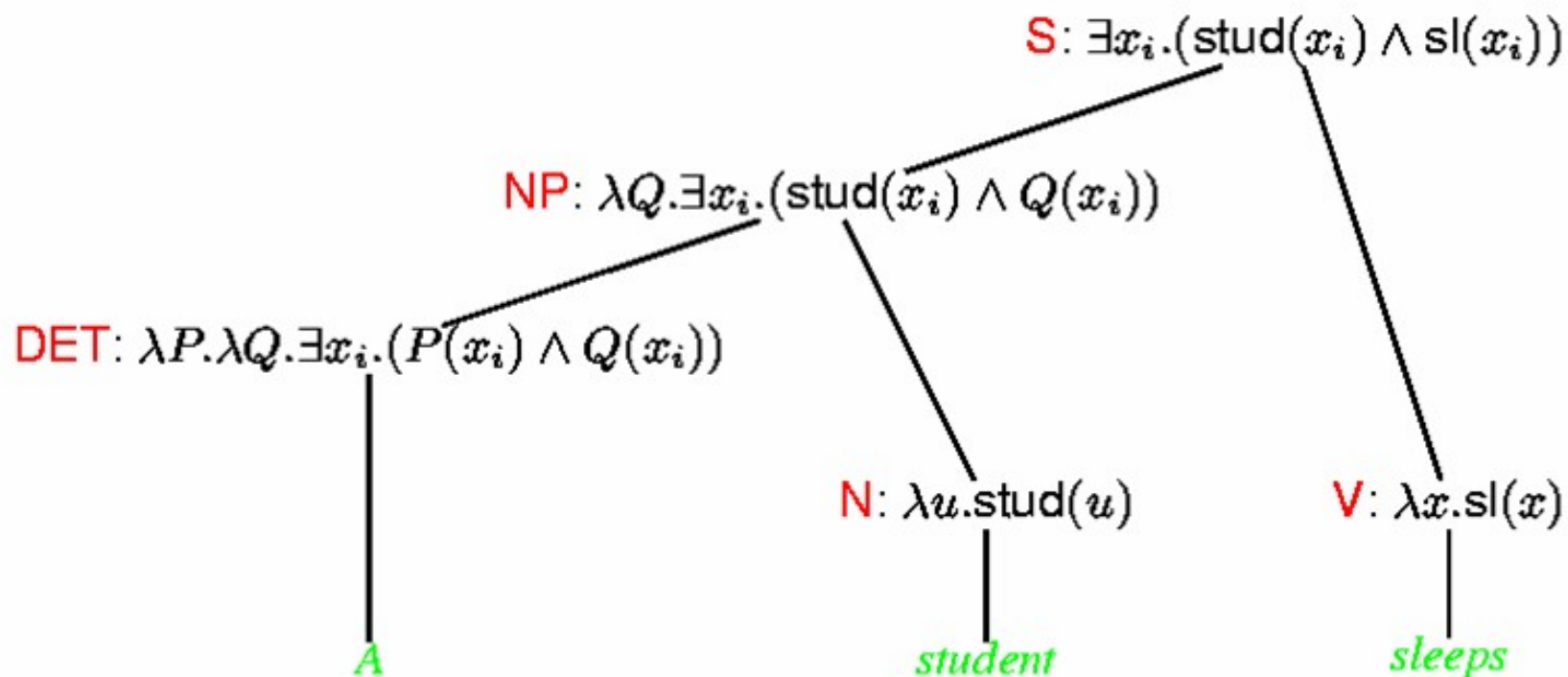
$NP \rightarrow$
| $Pronoun [0.10]$
| $Name [0.10]$
| $Noun [0.20]$
| $Article Noun [0.50]$
| $NP PP [0.10]$

$VP \rightarrow$
| $Verb [0.60]$
| $VP NP [0.20]$
| $VP PP [0.20]$

$PP \rightarrow Preposition NP [1.00]$

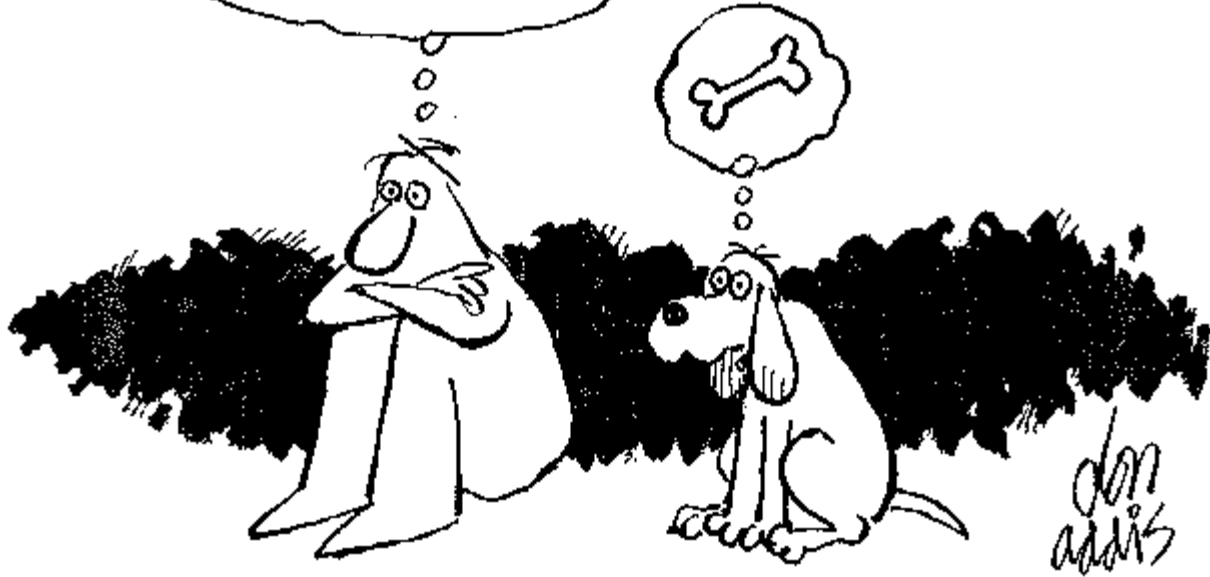
Semantics

Montagovian Semantics [1973]

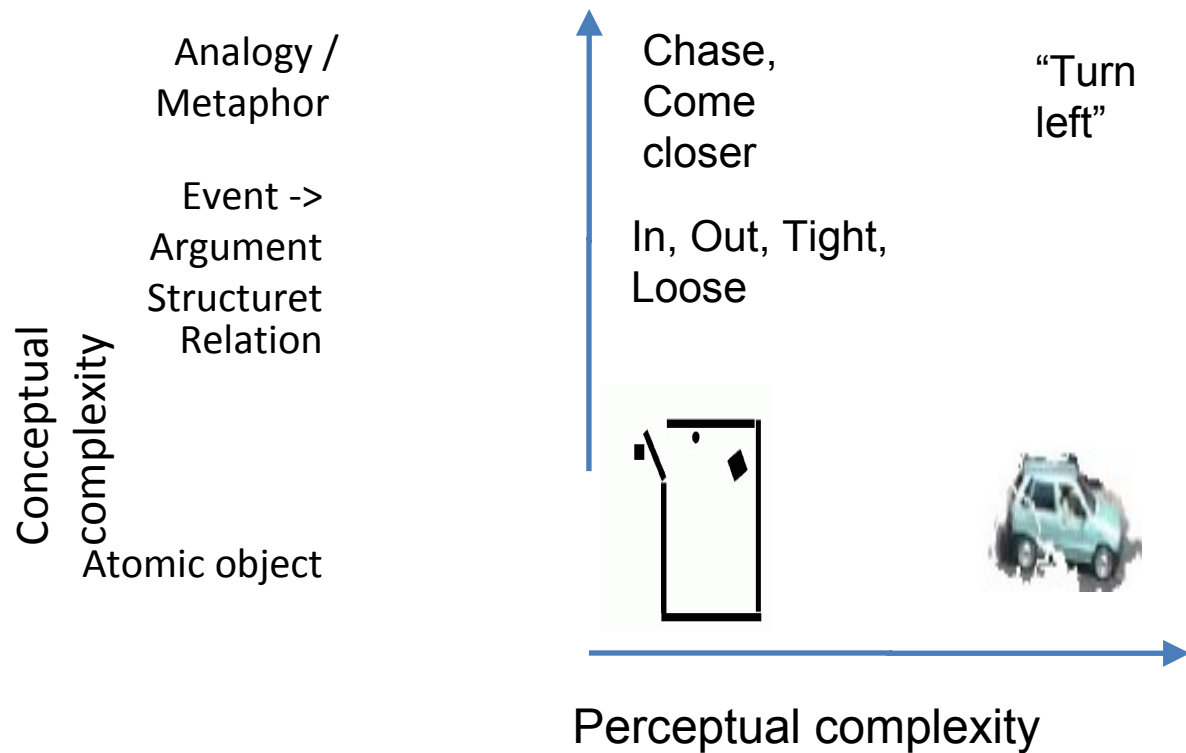


From [Kohlhase]

WHO? WHAT?
WHERE? WHEN? HOW?
WHY? WHICH? HOW MUCH?
HOW MANY? HOW LONG? HOW FAR?
WHAT FOR? WHAT NEXT? THEN
WHAT? WHY **ME?**



Semantics First: A pathway to Cognition



Instead of trying to produce a programme to simulate the adult mind, why not rather try to produce one which simulates the child's?

If this were then subjected to an appropriate course of education one would obtain the adult brain.

- Alan Turing

FrameNet: Semantic Roles

Familiar notion in NLP

■ Restaurant Frame:

□ “John ate chicken tandoori with his fingers.”

■ FrameNet = Comprehensive Lexicon of Frames

- Roger C. Schank. 1972. Conceptual dependency: A theory of natural language processing. *Cognitive Psychology*,
- Robert F. Simmons. 1973. Semantic networks: Their computation and use for understanding English sentences.

Semantic Roles

- The underlying relationship that a constituent has with the target word in a clause.
- Eg : *John hit Bill.*
Agent : John
Victim : Bill
- Apt for capturing semantic information -:
 - systematic method for capturing the event structure
 - the value that a role takes is independent of the syntactic structure of the sentence



FrameNet

The Frame is the basic lexical structure that links:

- individual word senses,
- relationships between the senses of polysemous words,
- relationships among semantically related words

Example

Frame : Ingestion

Frame Elements:

Core: Eater Eaten

Peripheral: Place Implement Manner Time

John [EATER]

ate [lexical unit]

chicken tandoori [EATEN]

at the Indian Restaurant [PLACE]

with his fingers [IMPLEMENT]

Participant semantics

The locals (Ingestor) EAT mainly fish and fruits (Ingestibles).

As the house doesn't have a dining room **the family(Ingestor) eats in the large kitchen(Place).**

She(Ingestor) took the ice-cream(ingestible) out of the fridge (source) and ate it.

Degree Ingestibles Ingestor Instrument Manner Means Place Source Time

Frame : Ingestion

Lexical
Units for
Ingestion

<u>English</u>	<u>Hindi</u>	<u>Bangla</u>
breakfast.v	नाश्ता	prAtarAsh v
Consume.v	भोग करना	bhog k.v
drink.v	पी	khA.v
eat.v	खा	khA.v
feast.v	भोज करना	bhoj k .v
feed.v	खिला	khAoyA.v
gulp.v	निगल	gelA.v
have.v	ले	Neo.v
munch.v	चवा	chebA.v
nibble.v	कुतर	ThokrA.v
sip.n	घूँट	chumuk.n
sip.v	घूँट लेना	Chumuk de.v

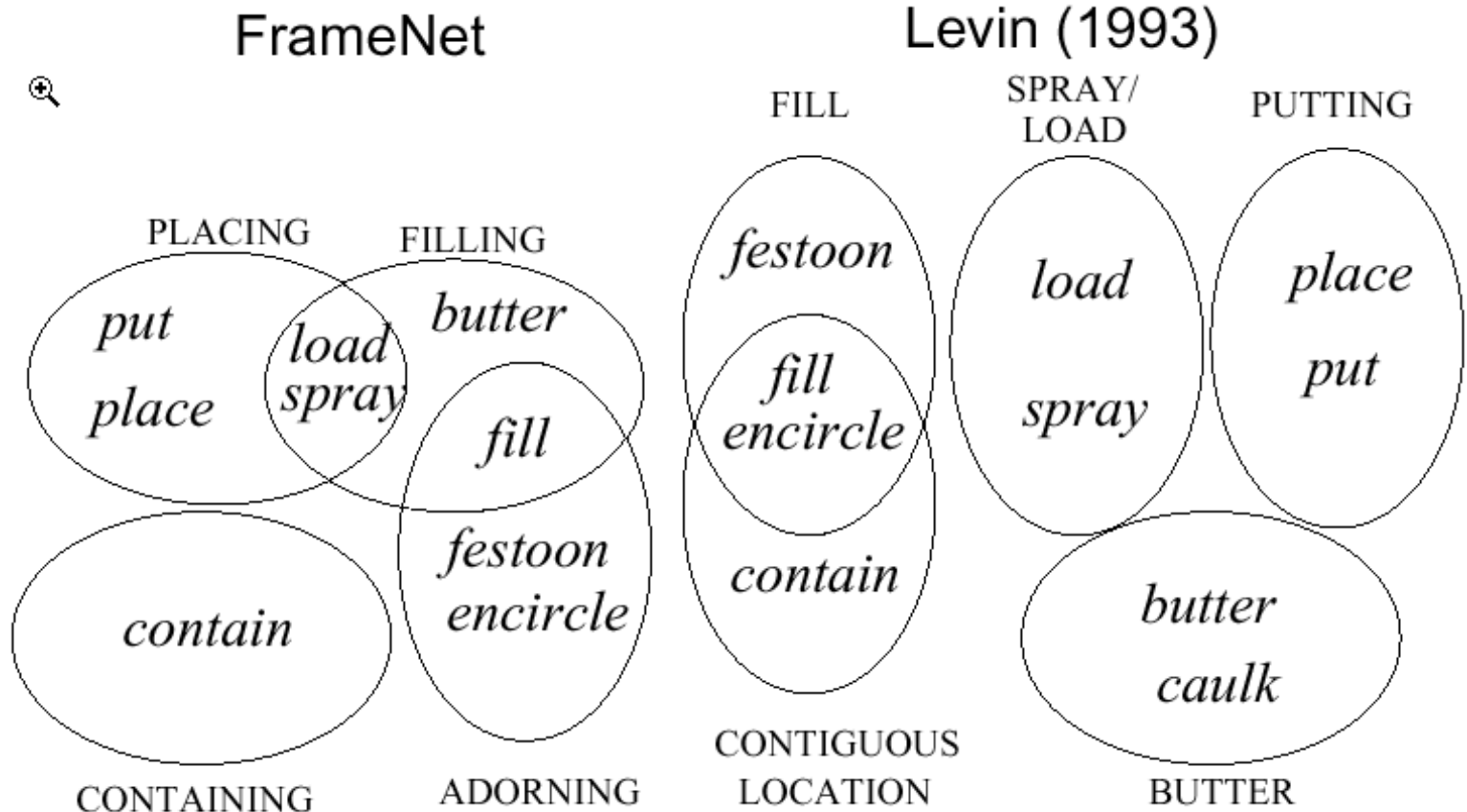
Parallel Sentence Analysis

As the house doesn't have a dining room, *the family* [EATER] *eat* [Lexical Unit] *in the large kitchen* [PLACE].

क्योंकि घर में भोजन कक्ष नहीं है , परिवार के लोग
[EATER] बड़े रसोयी [PLACE] में खाते हैं [LU]

bARite bhojan kakSha nei tAi *paribArer sabAi*
[EATER] *baRa rAnnAghare* [PLACE] *khAy* [LU].

Other Semantic Categorization Schemes :



Communication Verbs

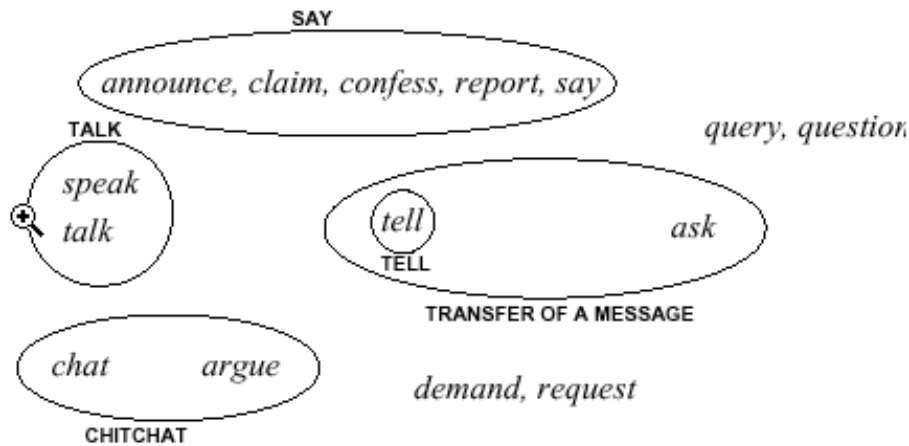


Figure 2 Communication verbs in Levin

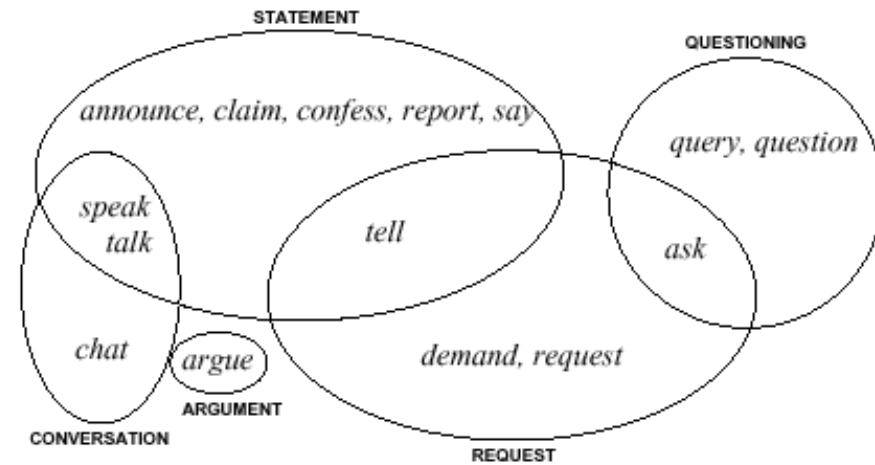


Figure 3 Communication verbs in FrameNet

PropBank / VerbNet

eat-39.1

Members

- [drink(1 2), eat(1 2 3)]

Thematic Roles

- Agent[+animate]
- Patient[+comestible]
- Instrument[+concrete]

Frames

Basic Transitive ()

"Cynthia ate the peach"

Agent V Patient

Unspecified Object Alternation ()

"Cynthia ate"

Agent V

Conative ()

"Cynthia ate at the peach"

Agent V Prep(at) Patient

Resultative ()

"Cynthia ate herself sick"

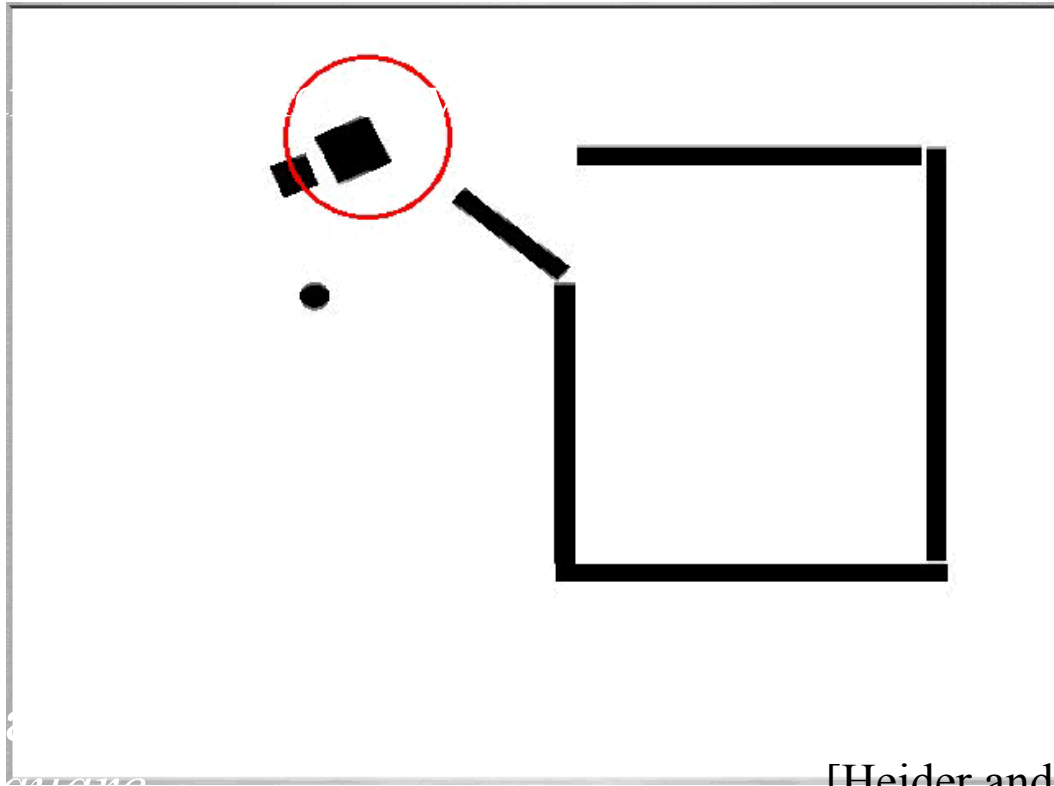
Agent V Oblique Adj

Semantic Tagging

- Probabilistic Role Assignment based on FrameNet Corpus [Gildea, 2002]
- **Linking Theory** : *“There is a unique relationship between the syntactic and semantic structure of a sentence”*
- Based on features extracted from parse tree, and probability (A statistical approach)

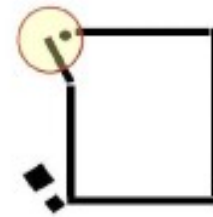
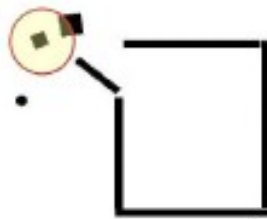
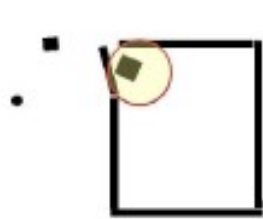
Grounded Language Learning

Heider/Simmel video



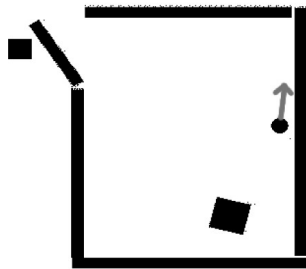
[Heider and Simmel 1944]
video recreated by Bridgette Hard
at Barbara Tversky lab, Stanford U

Visual attention model

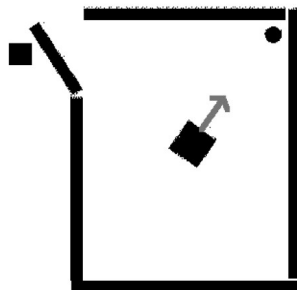


Maji, Singh and Mukerjee 2005

Narratives: "Chase" Video



00:00:56



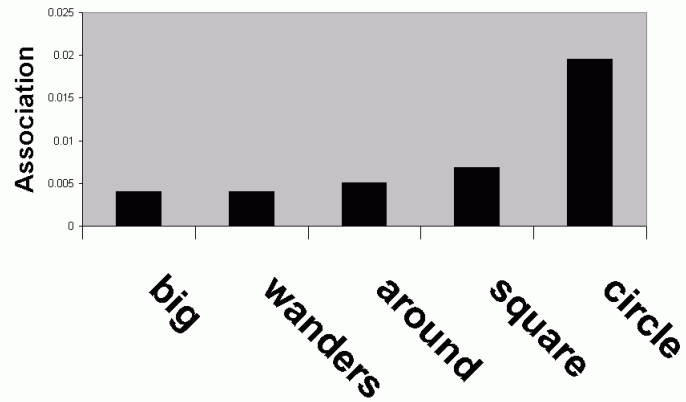
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Wide variation in Narratives :

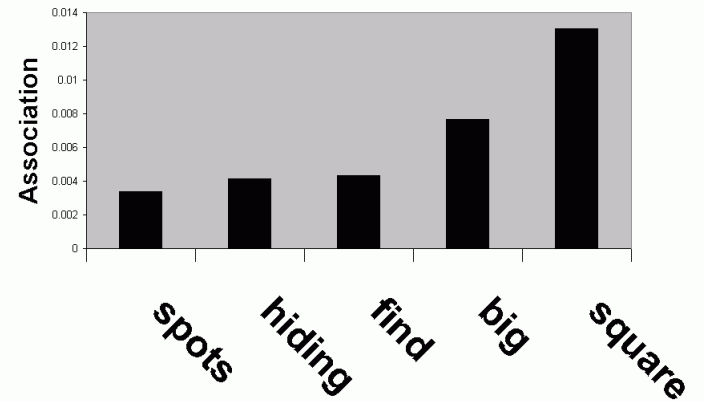
1. Large square corners the little circle
2. Big square approaches little circle
3. Little square is moving away from the big square; and objects inside are moving closer together
4. Big block tries to go after little circle

Noun Learning

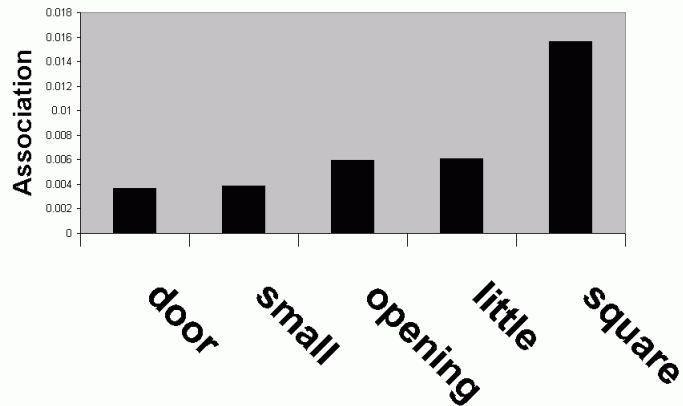
Circle



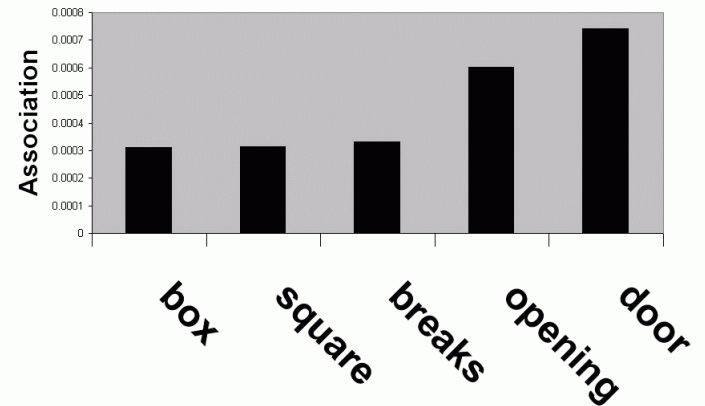
Big Square



Little Square

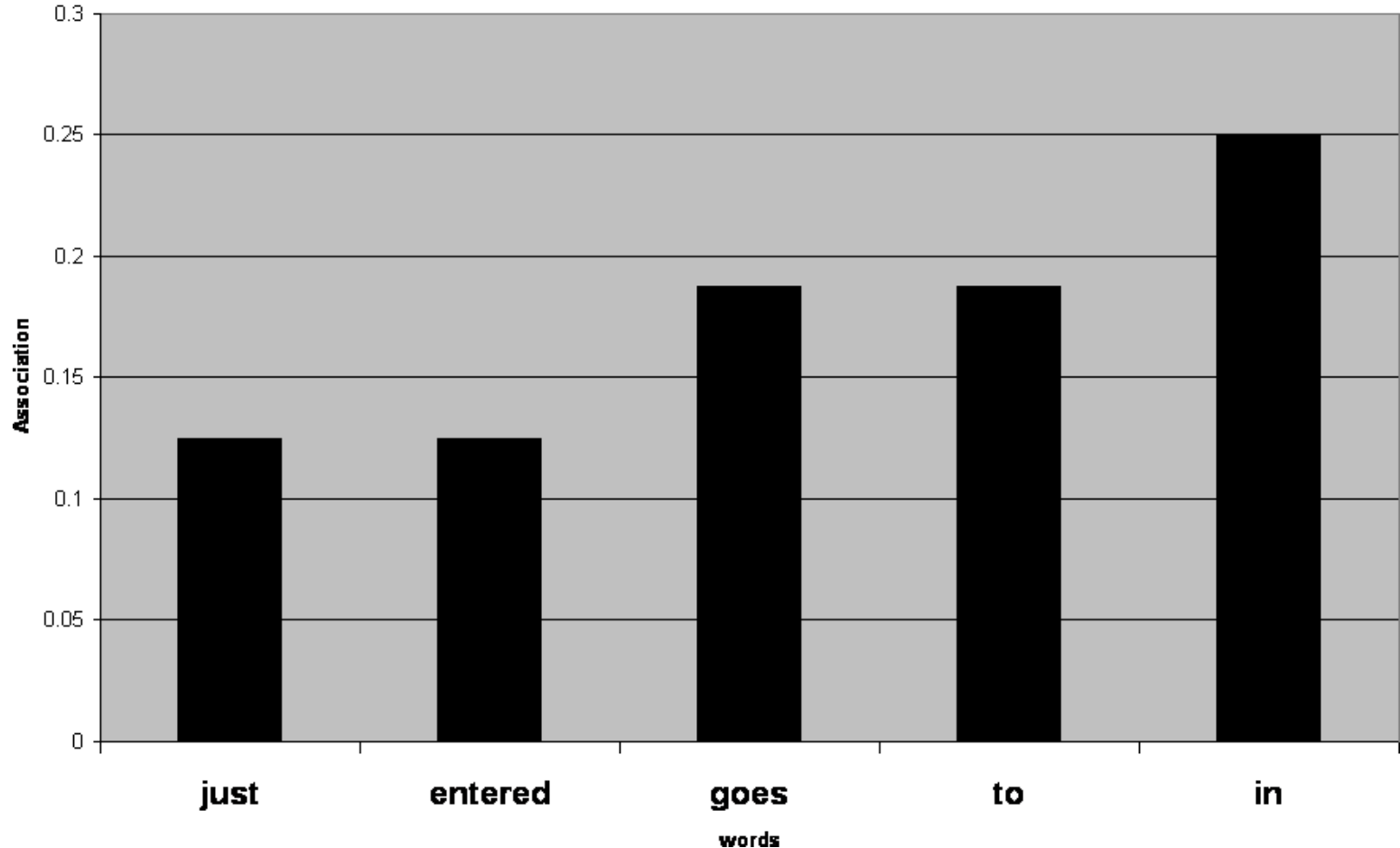


Door



Trajectories ending inside (“in”)

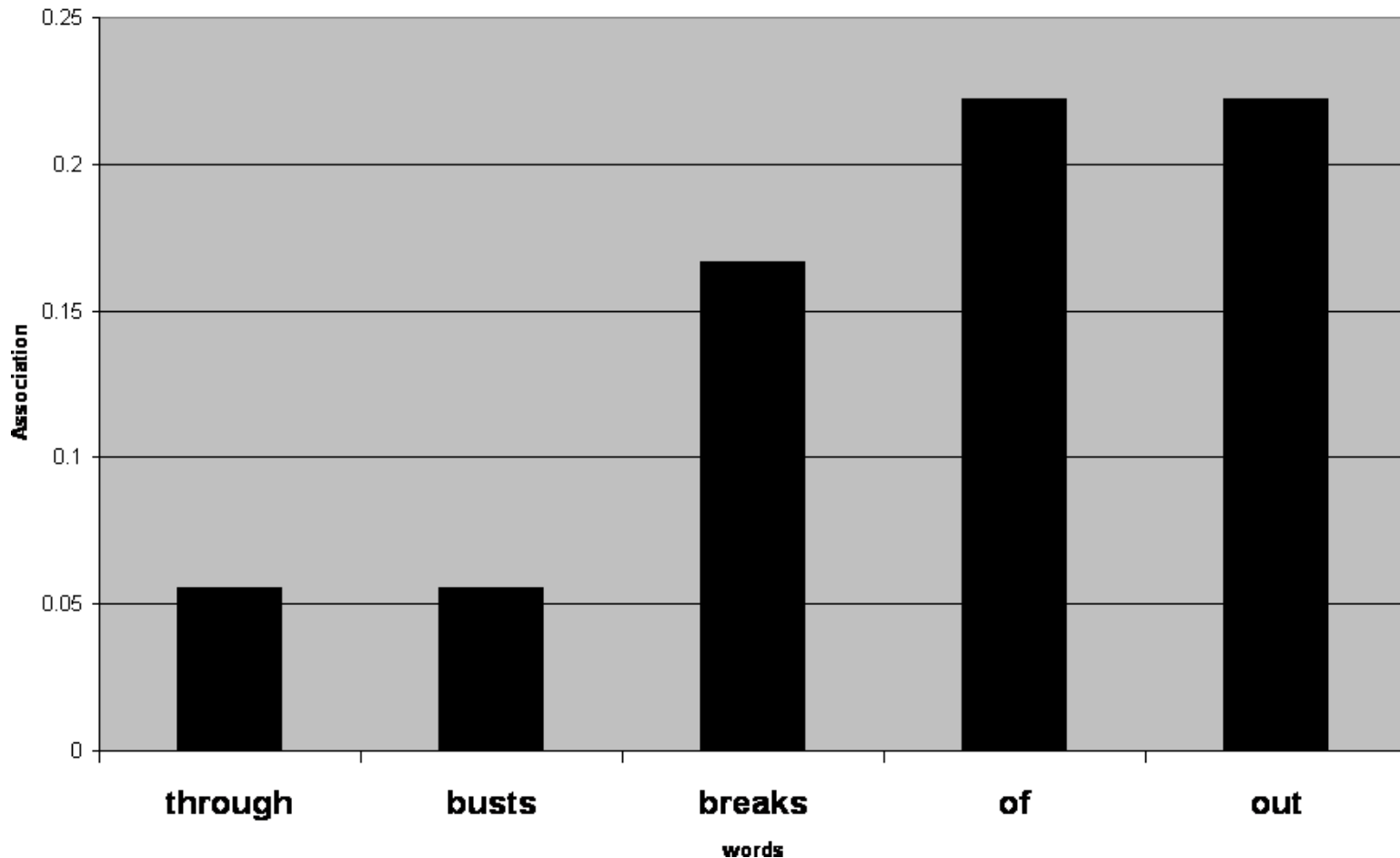
Out to In with verbs



Based on intervals where the attended agent is ending “in” the box

Trajectories ending Outside

In to Out with verbs



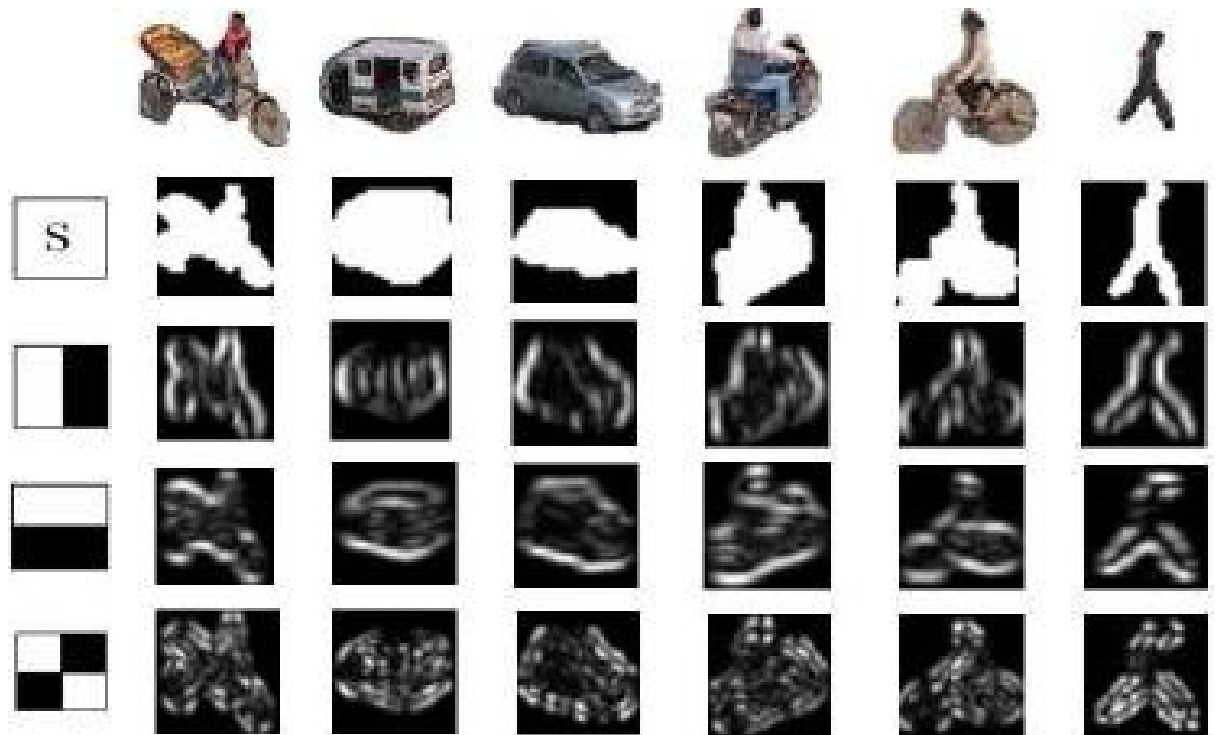
Learning Containment Spatial Descriptors

Recognition from real video

Learning Agent Appearances



Shape + Haar clusters



Guha and Nandi 09 model

PHOW clusters

C16



C19



C21



C23



C26



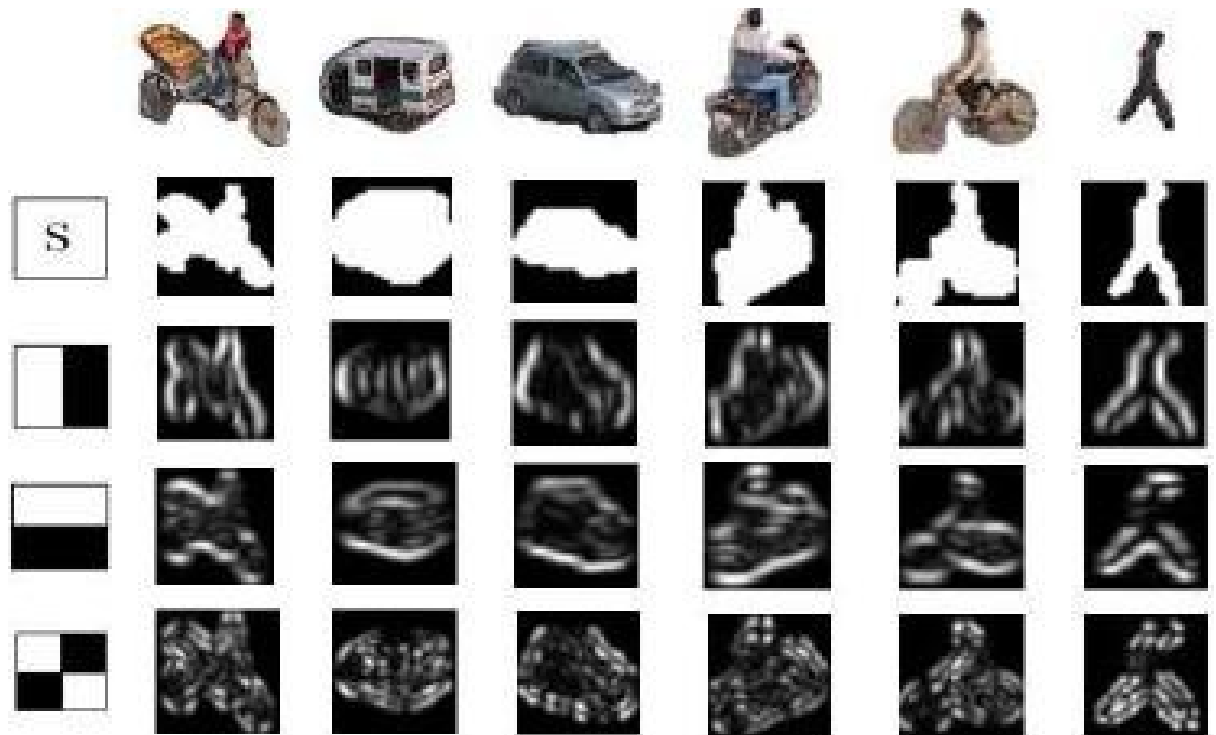
C27



C29



Shape + Haar clusters



Guha and Nandi 09 model

Unsupervised clustering results

H:52

C1	12/13	12H,1X
C2	7/8	7H,1N
C4	8/9	8H,1C
C10	7/9	7H,2N
C11	4/6	4H,1X,1N
C13	5/8	5H,1X,1B,1N
C14	2/4	2H,1T,1M
C21	6/6	6H
C3	3/3	3M
C8	8/9	8M,1X
C9	11/15	11M,2T,1X,1B
C22	6/6	6M
C23	3/5	3M,2B
C24	2/2	2M
C26	2/8	2M,2B,1X,1T,1R,1C

M:36

B:32

C5	5/5	5B
C6	1/2	1B,1X
C7	2/3	2B,1X
C15	2/2	2B
C20	5/5	5B
C28	7/8	7B,1T
C0	8/16	8T,4X,2C,1L,1R
C16	1/1	1T
C17	1/2	1T,1B
C18	1/1	1T
C25	4/7	4T,2C,1N
C12	4/5	4L,1C
C29	7/8	7L,1T
C19	9/10	9C,1X
C27	2/4	2N,1B,1H

T:21

L:12

C:16

N:8

Guha and Nandi 09 model

Sample Commentaries

	Sentence	Interval
S1	एक बाइक गयी अभी	1158 - 1224
	ek bAik gayI abhI	
	One bike go+past now.	
	A bike went just now.	
S2	साइड में एक साइकल रिक्शा पे एक आदमी चढ़ा	1216 - 1382
	sAiD meN sAikal rikshA pe ek ADamI chaDhA	
	Side [on] one cycle rickshaw [on] one man climb+past	
	A man climbed on a cycle rickshaw on the side (of the scene).	
S3	*साइकल बाइक आये जा रहे हैं	1239 - 1354
	sAikal bAik Aye jA rahe haiN.	
	Bicycles bikes come+pp go+pp are.	
	Bicycles, bikes are coming and going.	

Word-Object Associations

Hindi phrase(l)	Trans literation	Gloss	$P(l \gamma)$
γ =HUMAN			
जा टेम्पो सड़क	jA Tempo saDak	going tempo road	0.0391 0.0204 0.0187
γ =MOTORCYCLE			
जा लोग फिर	jA log phir	going people again	0.0295 0.0202 0.0191
γ =BICYCLE			
साइकल जा लोग	sAikal jA log	bicycle going people	0.0699 0.0553 0.0444
γ =TEMPO			
जा लोग ट्रक	jA log Trak	going people truck	0.0425 0.0208 0.0182
γ =TRUCK (Lorry)			
ट्रक लोग फिर	Trak log phir	truck people again	0.0374 0.0228 0.0216
γ =CAR			
कार जा ट्रक	kAr jA Trak	car going truck	0.0402 0.0342 0.0254