# Developing Probabilistic Models for Identifying Semantic Patterns in Texts

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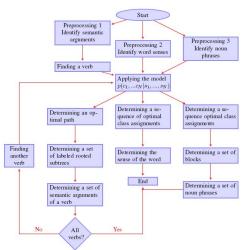
#### Introduction

#### Three text patterns capturing semantics of a sentence

- Semantic arguments of a verb
  - Semantic arguments of a verb can be used to answer the questions of who, what, when, where, and why.
- The meaning of a word
  - The sense of a polysemous word can be used to understand the meaning of the word.
- Noun phrases
  - Noun phrases of a sentence combining with verbs can be used to find the abstraction of the sentence.

## The Algorithm

• The key of the algorithm is a probabilistic graphical model.



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## The Probabilistic Graphical Model

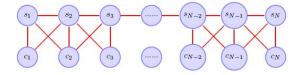


Figure: The conditional independence graph defining our graphical model.

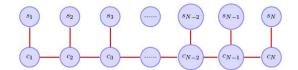
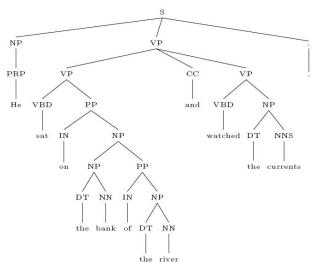
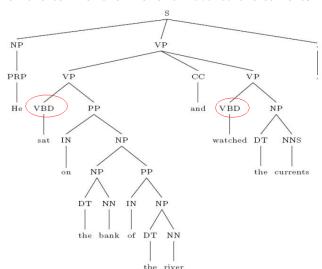
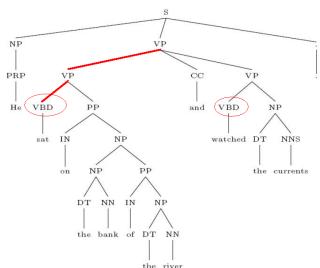


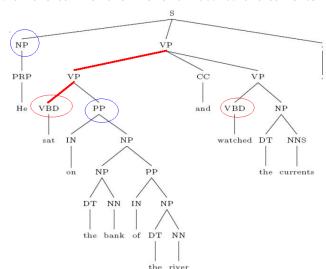
Figure: The usual conditional independence graph for Markov dependencies among the classes.

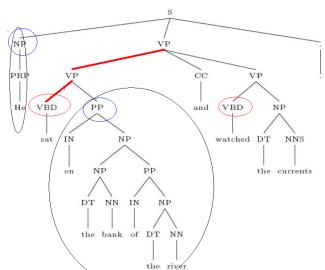


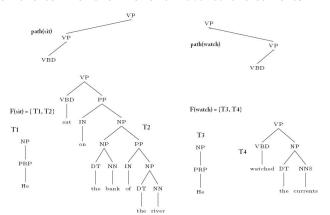












- Semantic arguments:
  - sit: he; on the bank of the river
  - watch: he: the currents

# **Empirical Results**

- Results for identifying semantic arguments of a verb in a sentence on WSJ data from Penn Treebank and PropBank
- About 600 verbs associating with about 2000 semantic arguments
- 10 fold cross validation technique

Files	Precision	Recall	F-Measure
20, 37, 49, 89	%	%	%
Average	92.335	94.1675	93.2512
Standard Deviation	0.6195	0.5174	0.4605

## **Empirical Results**

- Results for identifying the meaning of a word in a sentence on line data
- Results are better than those published by other researchers
- 10 fold cross validation technique

	Accuracy	Accuracy	# of Context words in Training Set	Base Line
	3 senses	6 senses	6 senses	6 senses
	%	%	k	%
LSA [11]	75			
Bayesian [9]	76	71	8.9	16.67
Context Vector [9]	73	72	8.9	16.67
Neural Network [9]	79	76	8.9	16.67
This Method	85.25	81.12	2.45	19.09

Outline Introduction The Algorithm The Model An Example **Empirical Results** 

## **Empirical Results**

- Results for identifying noun phrases in a sentence on CoNLL — 2000 data
- Results are better than The-Context-Independent-Bayes model
- Results are better than those published by other researchers

Method	Recall	Precision	F-measure
	%	%	%
Role Based Learning [12]	92.03	91.05	91.54
HMM [1]	93.52	93.43	93.48
Naive Bayes			93.69
MEMM [13]	_	_	93.70
Voted perceptrons [14]	93.29	94.19	93.74
CRF [13]	_	_	94.38
SVM [15]	94.38	94.52	94.45
our method [16]	95.31	96.36	95.74

