

Incremental Tree Substitution Grammar for Parsing and Sentence Prediction

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Introduction

Incrementality and Sentence Prediction

- ► Human sentence processing is incremental (Demberg and Keller, 2008). Predicts upcoming words (Grabski and Scheffer, 2004).
- Assumes strongly lexicalized parsing model.

Incremental TSG Model (ITSG)

Incremental TSG Generative Process ► Arbitrarily large fragments as in TAG/TSG (Schabes, 1990; Bod et al., 2003). ► Fully connected incremental structure.

Experimental Setup

Corpus Setup

- ▶ Penn WSJ Treebank (Marcus et al., 1993).
- ► Removing traces and functional tags.
- ► Apply right binarization (Klein and Manning, 2003), with no horizontal and vertical conditioning (H0V1).
- ▶ Replace words appearing < 5 times in the train with lexical features (Petrov, 2009).

Grammar Extraction

- ► Use *FragmentSeeker* (Sangati et al., 2010).
- ► Remove all non valid frags (e.g., no lexical items).
- ► Add all one-word fragments (minSet).
- ► Count frag frequency in the training corpus.



► Left-Right generative process (each step must extend the prefix). More constraints on the fragments (lexical anchor in first or second position).



Fragment Types

3 fragment types:

- Initial: lexical anchor in the first position (sentence initial).
- Lex-First: lexical anchor in first position (non sentence-initial).
- ► Sub-First: lexical anchor in second position, and a substitution site in first.

- Parsing
- ► MPD: Maximum Probable (partial) Derivation.
- ► MPP: Maximum Probable Parse (approximated).
- ► MRP: Minimum Risk Parse (Goodman, 1996).

Evaluation

Metrics

- Standard Parsing Evaluation (full sentences).
- ► Incremental Parsing Evaluation: for each prefix of the input sentence we compute the parsing accuracy on the minimal structure spanning that prefix.
- Sentence Prediction (for every prefix).
- ▶ Word prediction PRD(m): whether the *m* predicted words are correct.
- ▶ Word presence PRS(m): whether the *m* predicted words are present in the same order.
- ► Longest common subsequence LCS: computes the LCS between the sequence of predicted words and the words following the prefix in the original sentence.

Other Models

- ► Demberg et al. (2014) [Standard Parsing]
- ► Schuler et al. (2010) [Standard Parsing]
- ► Roark (2001); Roark et al. (2009) [Standard Parsing, Incremental Parsing]
- ► 3-gram model using SRILM (Stolcke, 2002) [Sentence Prediction]

Results

Standard Parsing



Combining Operations

- 4 operations (+ START, STOP)
- ► Scan
- Backward Substitution
- Forward Substitution
- ► Complete

Probabilistic Chart Parser

Earley based (Earley, 1970; Stolcke, 1995). No cycles.





Prefix Length

Conclusions

First incremental TSG parser.

- Competitive results on both full sentence and sentence prefix F-score.
- Outperforms standard n-gram LM in predicting more than one upcoming word.

Sentence Prediction

	ITSG			3-gram LM (SRILM)		
	Correct	R	Р	Correct	R	Р
PRD(1)	4,637	8.7	12.5	11,430	21.5	21.6
PRD(2)	864	1.7	13.9	$2,\!686$	5.3	5.7
PRD(3)	414	0.9	20.9	911	1.9	2.1
PRD(4)	236	0.5	23.4	387	0.8	1.0
PRS(1)	34,831	65.4	93.9	21,954	41.2	41.5
PRS(2)	4,062	8.0	65.3	5,726	11.3	12.2
PRS(3)	$1,\!066$	2.2	53.7	$1,\!636$	3.4	3.8
PRS(4)	541	1.2	53.7	654	1.4	1.7
LCS	44,454	5.9	89.4	92,587	12.2	18.4

Prefix	Shares of UAL, the parent	PRD(3)	PRS(3)
ITSG	company of United Airlines,	_	_
SRILM	company, which is the – –		
Goldstd	of United Airlines, were extremely active all day		
	Friday .		
Prefix	PSE said it expects to report earnings of \$ 1.3		
	million to \$ 1.7 million, or 14		
ITSG	cents a share,		+
SRILM	% to \$ UNK	_	—
Goldstd	cents to 18 cents a share.		

 $i:_k X \triangleleft \lambda \bullet \mu \quad [\alpha, \gamma, \beta, \leftarrow]$ Prob. State

> forward probability $P(\ell_0^{i-1}, i: {}_kX \triangleleft \lambda \bullet \mu)$ lpha : $P(\ell_k^{i-1}, i: {}_kX \triangleleft \lambda \bullet \mu)$ inner probability $P(\ell_0^{k-1}, \ell_i^N, i: {}_kX \triangleleft \lambda \bullet \mu)$ outer probability

 \leftarrow : viterbi best previous state

Probability Frag. Type Symbol Fragment $f(\pi_{init})$ Initial π_{init} $f(\pi_{lex(X)})$ Lex-First $\pi_{lex(X)}$ $\overline{\sum_{\pi'_{lex(X)}} f(\pi'_{lex(X)})}$ $\frac{f(\pi_{sub(Y)})}{\sum_{\pi'_{sub(Y)}} f(\pi'_{sub(Y)})}$ Sub-First

Chart Algorithm
forwards (α) and inners (γ)

Start:	Propagating forwards (α) as	nd inners (γ)			
	$X \triangleleft \ell_0 \nu$		$\alpha = P_{init}(X \triangleleft x)$ $\gamma = P_{init}(X \triangleleft x)$	$\ell_0 \nu$)	
	$0: {}_0X \triangleleft \bullet \ell_0 \nu \ [\alpha, \gamma, \mu]$	3]	$\beta = \beta(1:_0X < \beta)$	$(1: {}_{0}X \triangleleft \ell_{0} \bullet \nu)$	
Scan:					
	$i:_k X \triangleleft \lambda \bullet \ell_i \mu \left[\alpha, \gamma, \right]$	β]	$\alpha' = \alpha$ $\gamma' = \gamma$		
	$i+1:_k X \triangleleft \lambda \ell_i \bullet \mu \left[\alpha', \gamma \right]$	$\gamma', \beta']$	$\beta = \beta'$		
Backwa	ard Substitution:				
	$i:_{k}X \triangleleft \lambda \bullet Y\mu \left[\alpha, \gamma, \beta\right]$	$Y \triangleleft \ell_i \nu$	$\alpha' + = c$ $\gamma' = P_{le}$	$ \frac{P_{lex}(Y \triangleleft \ell_i \nu)}{P_{rx}(Y \triangleleft \ell_i \nu)} $	
	$i: {}_{i}Y \triangleleft \bullet \ell_{i}\nu \left[\alpha', \gamma \right]$	$\gamma', \beta']$	β not i	updated (done with back-completion	
Forwar	d Substitution:				
	$i: {}_0Y \triangleleft \nu \bullet \ [\alpha, \gamma, \beta]$	$X \triangleleft Y \ell_i \mu$	$\alpha' + = \alpha$ $\gamma' + = \gamma$		
	$i: {}_{0}X \triangleleft Y \bullet \ell_{i}\mu \left[\alpha'\right]$	$[\gamma,\gamma',\beta']$	$\beta + = \beta'$	$\beta + = \beta' \cdot P_{sub}(X \triangleleft Y \ell_i \mu)$	
Comple	etion:				
	$i:_{j}Y \triangleleft \ell_{j}\nu \bullet \ [\alpha, \gamma, \beta]$	$j:_kX \triangleleft \lambda$	• $Y\mu \left[\alpha', \gamma', \beta' \right]$	$\alpha'' + = \alpha' \cdot \gamma$ $\gamma'' + = \gamma' \cdot \gamma$	
	$i:{}_kX \triangleleft \lambda Y$	• $\mu \left[\alpha'', \gamma'', \beta \right]$	"]	$\beta' + = \beta'' \cdot \gamma'$ $\beta' + = \beta'' \cdot \gamma''$	
Stop:					
	$N:_0 Y \triangleleft \nu \bullet \ \big[\alpha = \gamma, \beta \big]$	$\ell_N = \text{STOF}$	$\alpha' = \gamma'$ $\beta' = 1$	$= \alpha \cdot P_{sub}(STOP)$	
	TERMINATE $(Y \triangleleft \nu)$	$[\alpha',\gamma',\beta']$	$\beta = P_{su}$	$_{bb}(STOP)$	

References

Rens Bod, Khalil Sima'an, and Remko Scha. 2003. Data-Oriented Parsing. University of Chicago Press, Chicago, IL, USA. Ciprian Chelba and Frederick Jelinek. 2000. Structured language modeling. Computer Speech and Language, 14:283-332. Vera Demberg and Frank Keller. 2008. Data from eye-tracking corpora as evidence for theories of syntactic processing complexity. *Cognition*, 101(2):193–210. Vera Demberg, Frank Keller, and Alexander Koller. 2014. Parsing with psycholinguistically motivated tree-adjoining grammar. Computational Linguistics, 40(1). Jay Earley. 1970. An efficient context-free parsing algorithm. *Commun. ACM*, 13(2):94–102. Joshua Goodman. 1996. Parsing algorithms and metrics. In Proceedings of the 34th annual meeting on Association for Computational Linguistics, pages 177–183. Association for Computational Linguistics, Morristown, NJ, USA. Korinna Grabski and Tobias Scheffer. 2004. Sentence completion. In Proceedings of the 27th annual international ACM SIGIR conference on Research and development in information retrieval, pages 433-439. Dan Klein and Christopher D. Manning. 2003. Accurate unlexicalized parsing. In ACL '03: Proceedings of the 41st Annual Meeting on Association for Computational Linguistics, pages 423–430. Association for Computational Linguistics, Morristown, NJ, USA. Mitchell P. Marcus, Mary Ann Marcinkiewicz, and Beatrice Santorini. 1993. Building a Large Annotated Corpus of English: The Penn Treebank. Computational Linguistics, 19(2):313–330. Slav Petrov. 2009. Coarse-to-Fine Natural Language Processing. Ph.D. thesis, University of California at Bekeley, Berkeley, CA, USA. Brian Roark. 2001. Probabilistic top-down parsing and language modeling. Comput. Linguist., 27:249–276. Brian Roark, Asaf Bachrach, Carlos Cardenas, and Christophe Pallier. 2009. Deriving lexical and syntactic expectation-based measures for psycholinguistic modeling via incremental top-down parsing. In Proceedings of the EMNLP 2009, pages 324–333. Association for Computat. Linguist., Stroudsburg, PA, USA. Federico Sangati, Willem Zuidema, and Rens Bod. 2010. Efficiently Extract Recurring Tree Fragments from Large Treebanks. In Proceedings of the Seventh conference on International Language Resources and Evaluation (LREC'10). European Language Resources Association (ELRA), Valletta, Malta. Yves Schabes. 1990. Mathematical and Computational Aspects of Lexicalized Grammars. Ph.D. thesis, University of Pennsylvania, Philadelphia, PA. William Schuler, Samir AbdelRahman, Tim Miller, and Lane Schwartz. 2010. Broad-coverage parsing using human-like memory constraints. Comput. Linguist., 36(1):1-30.

Andreas Stolcke. 1995. An efficient probabilistic context-free parsing algorithm that computes prefix probabilities. Comput. Linguist., 21(2):165–201. Andreas Stolcke. 2002. SRILM – An extensible language modeling toolkit. In Proc. International Conference on Spoken Language Processing, pages 257–286.

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