SVMTool: A general POS tagger generator based on Support Vector Machines

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Outline

- Introduction
 - Part-of-speech Tagging
 - Idea and Motivation
 - Learning Framework
- SVMT tool
- Evaluation
- Conclusions

Introduction

Part-of-Speech Tagging

The_DT SVMTool_NNP is_VBZ now_RB being_VBG presented_VBN to_TO NLP_NNP researchers_NNS in_IN Lisbon_NNP ._.

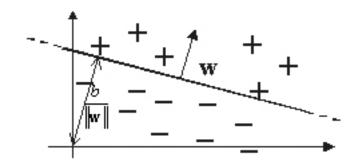
- Brill [Brill, 1995]
- TnT [Brants, 2000]

Idea and Motivation

- Accuracy [in the state-of-the-art]
- Efficiency [both learning and tagging]
- Flexibility [highly customizable]
- Portability [language independent]
- Robustness [against overfitting and on-line mistakes]
- Simplicity [easy to use]

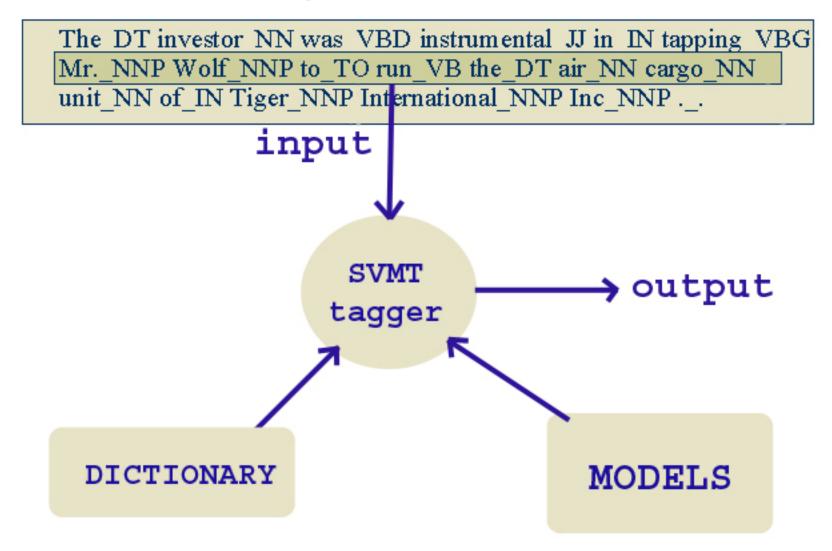
Learning FrameWork

Support Vector Machines



$$h(\mathbf{x}) = \text{sign } (\langle \mathbf{w} \cdot \mathbf{x} \rangle + b) = \begin{cases} +1 & if \langle \mathbf{w} \cdot \mathbf{x} \rangle + b > 0 \\ -1 & otherwise \end{cases}$$

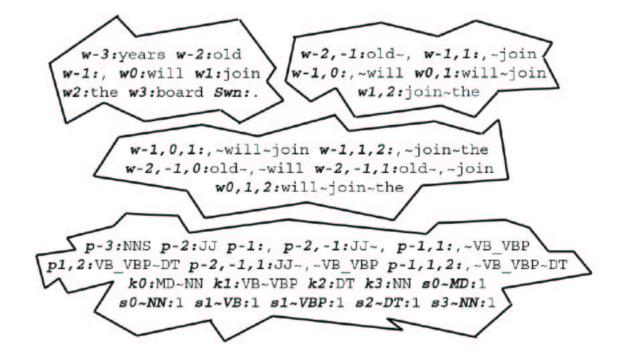
SVMT tool



Feature Patterns

word unigrams	$w_{-3}, w_{-2}, w_{-1}, w_0, w_{+1}, w_{+2}, w_{+3}$
word bigrams	$(oldsymbol{w}_{-2},oldsymbol{w}_{-1}), (oldsymbol{w}_{-1},oldsymbol{w}_{+1}), (oldsymbol{w}_{-1},oldsymbol{w}_{0}), (oldsymbol{w}_{0},oldsymbol{w}_{+1}), (oldsymbol{w}_{+1},oldsymbol{w}_{+2})$
word trigrams	$(oldsymbol{w}_{-3}, oldsymbol{w}_{-2}, oldsymbol{w}_{-1}), (oldsymbol{w}_{-2}, oldsymbol{w}_{-1}, oldsymbol{w}_{0}), (oldsymbol{w}_{-2}, oldsymbol{w}_{-1}, oldsymbol{w}_{+1}),$
	$(oldsymbol{w}_{-1},oldsymbol{w}_{0},oldsymbol{w}_{+1}), (oldsymbol{w}_{-1},oldsymbol{w}_{+1},oldsymbol{w}_{+1},oldsymbol{w}_{+1},oldsymbol{w}_{+1},oldsymbol{w}_{+2})$
POS unigrams	p_{-3}, p_{-2}, p_{-1}
POS bigrams	$(p_{-2},p_{-1}),(p_{-1},a_{+1}),(a_{+1},a_{+2})$
POS trigrams	$(p_{-3},p_{-2},p_{-1}),(p_{-2},p_{-1},a_{+1}),(p_{-1},a_{+1},a_{+2})$
ambiguity classes	$oxed{a_0,a_1,a_2,a_3}$
maybe's	$oxed{m_0,m_1,m_2,m_3}$
prefixes	s_1 , s_1s_2 , $s_1s_2s_3$, $s_1s_2s_3s_4$
suffixes	s_n , $s_{n-1}s_n$, $s_{n-2}s_{n-1}s_n$ $s_{n-3}s_{n-2}s_{n-1}s_n$
binary word-form	intial_Upper_Case, all_Upper_Case, no-initial_Capital_Letter(s),
features	all_Lower_Case, contains_(period/number/hyphen)
word length	integer
Sentence info	last_word ('.', '?', '!')

```
Pierre NNP Vinken NNP , , 61 CD years NNS old JJ , , will ?? join VB the DT board NN as IN a DT nonexecutive JJ director NN Nov. NNP 29 CD . .
```



Outline

- Introduction
- SVMT tool
 - SVMT-learner [Training of SVM classifiers]
 - SVMT-tagger [POS-tagging of a given input]
 - SVMT-evaluator [Study of tagging results]
 - SVMT API [Embedded usage of SVMT-tagger]
- Evaluation
- Conclusions

SVMT-learner

Options

- slinding window: length [def: 5] core position [def: 2]
- feature set [configurable]
- feature filtering [default: (2 / 100,000)]
- C parameter tuning (greedy) [default: disabled]
- SVM model compression [default: disabled]
- ambiguous/open-class POS lists may be provided if available. [automatically created by default]

SVMT-learner

- dictionary repairing
 - heuristic [by default]
 - using a list of corrections provided
 - < the 50975 6 CD 1 DT 50959 JJ 7 NN 1 NNP 6 VBP 1 >
 - < the 50975 1 DT 50959 >

SVMT-tagger

Options

- tagging scheme
 - * greedy [default]
 - * sentence-level
- tagging direction
 - * left-to-right [default]
 - * right-to-left
 - * both left-to-right and right-to-left
- number of tagging passes (1 or 2) [default: 1]
- backup lexicon

SVMT-evaluator

 $gold\ output\ +\ SVMT\ output\ =\ report$

- brief report
- known vs. unknown tokens
- level of ambiguity
- class of ambiguity
- part-of-speech study

SVMT API

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- Introduction
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- Evaluation
 - English on WSJ
 - Spanish on LEXESP
- Conclusions

Evaluation

learning time	1-20 cpu hour		
tagging speed	1500 words/second		

2Ghz Pentium-IV processor; 1Gb RAM
 Perl v5.005_03 (Benchmark package for timing)

Evaluation for English

- Wall Street Journal [Penn Treebank III]
- 1,17 million words
 [Training (912k), Validation (132k) and Test (130k)]
- Penn Treebank tagset -> 48 tags
 35 parts-of-speech present ambiguity
 17 are open-classes

	TnT	Collins 02	SVMT	Toutanova et al.
Accuracy	96.46%	97.11%	97.16%	97.24%

Evaluation for Spanish

- LEXESP
- 106k words [Training (86k) and Test (20k)]
- Parole tagset -> 183 tags -> 61 tags (reduced tagset)
 43 parts-of-speech present ambiguity
 12 are open-classes

	TnT	SVMT
Accuracy	96.50%	96.89%

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Conclusions

- highly accurate: 97.0 97.2% [English on WSJ]
- efficient: linear SVMs, primal formulation
- robust: soft margin SVMs, two-passes, LR + RL
- very flexible: rich feature set, tagging strategies
- portable: applied to English, Spanish and Catalan
- simple: ease to configure, tune and use

Ongoing Steps

- C++ version coming soon
- Study of more flexible and robust tagging schemes
- Better guessing of unknown words
- Unsupervised learning

Thanks

you may download SVMTool v 1.2 at

http://www.lsi.upc.es/~nlp/SVMTool

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