Time Expression Recognition Using a Constituent-based Tagging Scheme

Xiaoshi Zhong and Erik Cambria

Nanyang Technological University
{xszhong, cambria}@ntu.edu.sg

Outline

- Time expression analysis
  - Datasets: TimeBank, Gigaword, WikiWars, Tweets
  - Findings: loose structure, differentiable
- Time expression recognition
  - TOMN: a constituent-based tagging scheme
  - Baselines: HeidelTime, SUTime, SynTime, ClearTK, UWTime
  - Datasets: TE-3, WikiWars, Tweets

Time Expression - Examples

Today
Friday
September
Last week
2 years ago
September 2006
2006 September
January 30, 1998
1 September 2006
the third quarter of 1984

Time Expression - Constituents

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- Time token
  - Explicitly express time information
- Modifier
  - Modifier time tokens
- Numeral
  - Numbers and ordinals (except year)

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Time Expression Analysis

- Datasets
  - TimeBank
  - Gigaword
  - WikiWars
  - Tweets
- Findings
  - Loose structure
  - Differentiable

Time Expression Analysis - Datasets

- Datasets
  - TimeBank: a benchmark dataset used in TempEval evaluations
  - Gigaword: a large dataset with automatically generated labels
  - WikiWars: a war domain dataset collected from Wikipedia
  - Tweets: a tweet dataset collected from Twitter
- Dataset statistics

Dataset	#Docs	#Words	#Timex
TimeBank	183	61,418	1,243
Gigaword	2,452	666,309	12,739
WikiWars	22	119,468	2,671
Tweets	942	18,199	1,127

The datasets differ in size, source, domain, and text type, but their time expressions demonstrate similar characteristics

Time Expression Analysis - Finding 1

- **Loose structure**: time expressions are formed by loose structure

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  - Loose collocation

    September
    September 2006
    1 September 2006
  - Exchangeable order

    September 2006
    2006 September

Time Expression Analysis - Finding 1

- **Loose structure**: time expressions are formed by loose structure
  - Loose collocation

    September/**B**
    September/**B** 2006/I
    1/B September/**I** 2006/I
  - Exchangeable order

    September/**B** 2006/I
    2006/B September/**I**

In perspective of position within time expressions, under BIO scheme, 'September' may appear as (i) **B**eginning or (ii) **I**nside word of time expressions

Time Expression Analysis - Finding 1

- **Loose structure**: time expressions are formed by loose structure
  - Loose collocation

    September/**U**
    September/**B** 2006/L
    1/B September/**I** 2006/L
  - Exchangeable order

    September/**B** 2006/L
    2006/B September/**L**

Under BILOU scheme, 'September' may appear as (1) **U**nit-word time expressions, (2) **B**eginning, (3) **I**nside, or (4) **L**ast word of time expressions

Time Expression Analysis - Finding 1

- **Loose structure**: time expressions are formed by loose structure

Percentage of distinct time tokens and modifiers that appear in *different positions* within time expressions

Dataset	**BIO Scheme**		**BILOU Scheme**	
	Time Token	Modifier	Time Token	Modifier
TimeBank	58.18	33.33	63.64	33.33
Gigaword	61.29	45.83	77.05	46.00
WikiWars	53.57	26.19	61.40	29.55
Tweets	67.21	27.59	72.58	27.59

Time Expression Analysis - Finding 2

- **Differentiable**: time tokens can differentiate time expressions from common text

Time Expression Analysis - Finding 2

- **Differentiable**: time tokens can differentiate time expressions from common text

  Percentage of time expression's constituents that appear in time expressions (*Ptimex*) and in common text (*Ptext*)

Dataset	**Time Token**		**Modifier**		**Numeral**	
	Ptimex	*Ptext*	*Ptimex*	*Ptext*	*Ptimex*	*Ptext*
TimeBank	94.61	0.34	47.39	22.56	22.61	3.16
Gigaword	96.44	0.65	28.05	22.82	20.24	2.03
WikiWars	91.81	0.14	31.64	26.14	38.01	9.82
Tweets	96.01	0.50	21.38	13.03	18.81	0.128

$$Ptimex(T) = \frac{\#\,timex\ \ that\ \ contain\ \ T}{\#\,total\ \ timex} \qquad Ptext(T) = \frac{\#\,tokens\ \ that\ \ are\ \ T}{\#\,total\ \ tokens}$$

Fundamental Problem - Inconsistent Tag Assignment

- Position-based tagging scheme
  - BIO scheme: **B**eginning or **I**nside word of time expressions, **O**utside time expressions
  - BILOU scheme: **U**nit-word time expressions, **B**eginning, **I**nside, **L**ast word of multi-word time expressions, **O**utside time expressions
- Inconsistent tag assignment
  - During training, a word is assigned with different tags simply because the word appears in different positions within labeled chunks

Inconsistent Tag Assignment

- Position-based tagging scheme
  - BILOU scheme: **U**nit-word time expressions, **B**eginning, **I**nside, **L**ast word of multi-word time expressions, **O**utside time expressions

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1) September/U
2) September/B 2006/L
3) 2006/B September/L
4) 1/B September/I 2006/L

Inconsistent Tag Assignment

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1) September/U
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3) 2006/B September/L
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1) (…, w=September, …, U)
2) (…, w=September, …, B)
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Inconsistent tag assignment reduces the predictive power of ‘September’

This contradicts finding 2 that time tokens can differentiate time expressions from common text

Review

- Two findings
  - Finding 1: time expressions are formed by loose structure
  - Finding 2: time tokens can differentiate time expressions from common text
- Finding 1 leads BILOU scheme to inconsistent tag assignment
  - Reduce the predictive power of time tokens

1) September/U 2) September/B 2006/L

3) 2006/B September/L 4) 1/B September/I 2006/L

  - Under BILOU scheme, Finding 1 contradicts Finding 2

Overcome Inconsistent Tag Assignment

- Constituent-based tagging scheme
  - TOMN scheme: **T**ime token, **M**odifier, **N**umeral, **O**utside time expressions
  - TOMN scheme assigns a word with a tag according to its constituent role

Overcome Inconsistent Tag Assignment

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1) September/T
2) September/T 2006/T
3) 2006/T September/T
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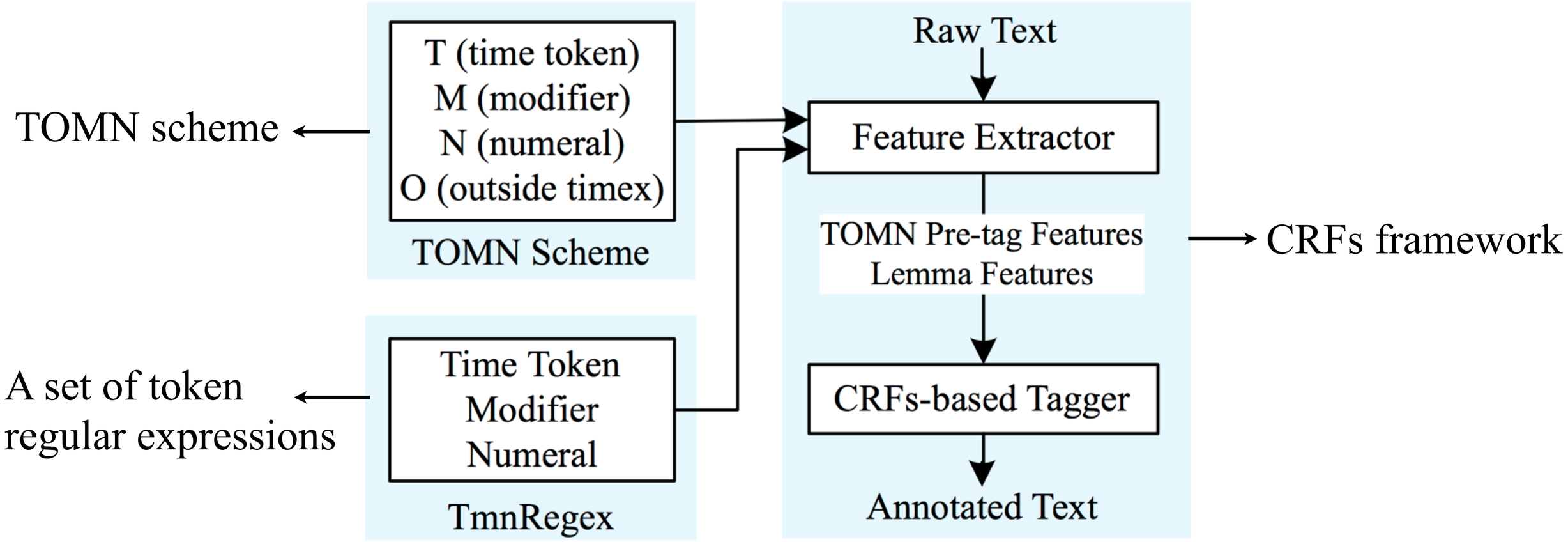
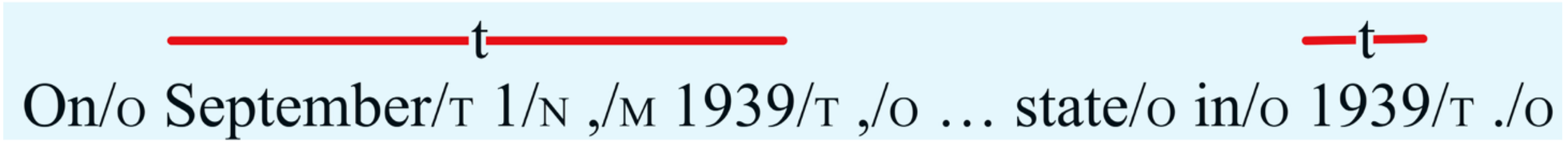
Consistent tag assignment protects ‘September’s predictive power

Time Expression Recognition - TOMN



Time Expression Recognition - Examples

- Non-O words that appear together form a time expression



Time Expression Recognition - Experiments

- Our method: TOMN
- Baselines
  - HeidelTime: rule-based
  - SUTime: rule-based
  - SynTime: type-based
  - ClearTK: learning-based
  - UWTime: learning-based
- Datasets
  - TE-3, WikiWars, Tweets

Performance of TOMN and baselines. **Best results** in boldface and second best underlined

Dataset	Method	Strict Match			Relexed Match		
		Pr.	*Re.*	*F1*	*Pr.*	*Re.*	*F1*
TE-3	HeidelTime (Strotgen et al., 2013)	83.85	78.99	81.34	93.08	87.68	90.30
	SUTime (Chang and Manning, 2013)	78.72	80.43	79.57	89.36	91.30	90.32
	SynTime (Zhong et al., 2017)	91.43	**92.75**	**92.09**	94.29	**95.65**	**94.96**
	ClearTK (Bethard, 2013)	85.90	79.70	82.70	93.75	86.96	90.23
	UWTime (Lee et al., 2014)	86.10	80.40	83.10	94.60	88.40	91.40
	TOMN	**92.59**	90.58	91.58	**95.56**	93.48	94.51
WikiWars	HeidelTime (Strotgen et al., 2013)	**88.20**	78.50	83.10	95.80	85.40	90.30
	SUTime	78.61	76.69	76.64	95.74	89.57	92.55
	SynTime (Zhong et al., 2017)	80.00	80.22	80.11	92.16	**92.41**	92.29
	ClearTK	87.69	80.28	**83.82**	96.80	90.54	93.54
	UWTime (Lee et al., 2014)	87.70	78.80	83.00	**97.60**	87.60	92.30
	TOMN	84.57	**80.48**	82.47	96.23	92.35	**94.25**
Tweets	HeidelTime	**91.67**	74.26	82.05	96.88	78.48	86.71
	SUTime	77.69	79.32	78.50	88.84	90.72	89.77
	SynTime (Zhong et al., 2017)	89.52	94.07	91.74	93.55	**98.31**	**95.87**
	ClearTK	86.83	75.11	80.54	96.59	83.54	89.59
	UWTime	88.36	70.76	78.59	**97.88**	78.39	87.06
	TOMN	90.69	**94.51**	**92.56**	93.52	97.47	95.45

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Time Expression Recognition - Experiments

- Cross-dataset experiments
  - Train on one dataset's training set
  - Test on other datasets' test sets
- Learning-based baselines
  - ClearTK
  - UWTime
- Datasets
  - TE-3
  - WikiWars
  - Tweets

Cross-dataset performance on **test set of TE-3**

(Color background indicates single-dataset results)

Training Set	**Method**	**Strict Match**			**Relexed Match**		
		Pr.	*Re.*	*F1*	*Pr.*	*Re.*	*F1*
TE-3	ClearTK	85.90	79.70	82.70	93.75	86.96	90.23
	UWTime	86.10	80.40	83.10	94.60	88.40	91.40
	TOMN	**92.59**	**90.58**	**91.58**	**95.56**	**93.48**	**94.51**
WikiWars	ClearTK	65.67	63.77	64.71	87.31	84.78	86.03
	UWTime	76.92	72.46	74.63	88.46	83.33	85.82
	TOMN	**84.06**	**84.06**	**84.06**	**93.48**	**93.48**	**93.48**
Tweets	ClearTK	72.59	71.01	71.79	**93.33**	91.30	92.31
	UWTime	80.00	72.46	76.05	92.80	84.06	88.21
	TOMN	**85.42**	**89.13**	**87.23**	91.67	**95.65**	**93.62**

Cross-dataset performance on **test set of WikiWars**

Training Set	**Method**	**Strict Match**			**Relexed Match**		
		Pr.	*Re.*	*F1*	*Pr.*	*Re.*	*F1*
TE-3	ClearTK	74.38	60.76	66.89	**97.54**	79.68	87.71
	UWTime	**87.01**	**79.34**	**83.00**	96.07	87.60	91.64
	TOMN	82.18	75.65	79.07	96.26	**87.93**	**91.90**
WikiWars	ClearTK	87.69	80.28	**83.82**	96.80	90.54	93.54
	UWTime	**87.70**	78.80	83.00	**97.60**	87.60	92.30
	TOMN	84.57	**80.48**	82.47	96.23	**92.35**	**94.25**
Tweets	ClearTK	57.75	54.73	56.20	91.93	87.12	**89.46**
	UWTime	**80.28**	62.81	**70.48**	**94.37**	73.83	82.84
	TOMN	60.29	**66.00**	63.02	84.74	**92.76**	88.57

Cross-dataset performance on **test set of Tweets**

Training Set	**Method**	**Strict Match**			**Relexed Match**		
		Pr.	*Re.*	*F1*	*Pr.*	*Re.*	*F1*
TE-3	ClearTK	81.16	47.26	59.73	97.10	56.54	71.47
	UWTime	89.66	65.82	75.91	94.83	69.62	80.29
	TOMN	**92.92**	**88.61**	**90.71**	**96.90**	**92.41**	**94.60**
WikiWars	ClearTK	72.48	45.57	55.96	95.30	59.92	73.58
	UWTime	**87.43**	61.60	72.28	**95.81**	67.61	79.21
	TOMN	85.00	**86.08**	**85.53**	93.75	**94.94**	**94.34**
Tweets	ClearTK	86.83	75.11	80.54	96.59	83.54	89.59
	UWTime	88.36	70.76	78.59	**97.88**	78.39	87.06
	TOMN	**90.69**	**94.51**	**92.56**	93.52	**97.47**	**95.45**

Time Expression Recognition - Efficiency

- TOMN is more efficient

Runtime of going through a whole process (unit: seconds)

Method	**TE-3**	**WikiWars**	**Tweets**
ClearTK	152	223	86
UWTime	864	1,050	160
TOMN	36	48	42

Summary

- Have two findings
  - Loose structure
  - Differentiable
- Reveal a fundamental problem in position-based tagging scheme
  - Inconsistent tag assignment
- Define a constituent-based tagging scheme
  - Good results
  - Less time