

A Framework for Cross-Domain Clinical Temporal Information Extraction

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There is a significant amount of unstructured data in the form of clinical notes.

"The intraoperative as well as the postoperative course were uncomplicated. The patient was transferred from the Vision Care Center to the Laurel Beach Medical Center where she was observed overnight."

This information could be the basis of a patient "timeline", which could be used to track disease status, monitor treatment outcomes, discover side-effects, etc.

Clinical TempEval



SemEval-2017 Task 12

Follows Clinical TempEval 2015 and 2016
and the i2b2 2013 temporal challenge

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Focus: Domain adaptation

Source domain: colon cancer notes

Target domain: brain cancer notes

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Unsupervised and supervised domain
adaptation

1,200 notes, two domains, three types
(clinical, pathology, radiology)

Annotations of events (`EVENT`), and time
expressions (`TIMEX3`)

Relations between them (`TLINK`)

TIMEX3
CLASS=DATE
April 23, 2014

The patient did not have any

TIMEX3
CLASS=PREPOSTEXP
postoperative

CONTAINS

EVENT
TYPE=N/A
DEGREE=N/A
POLARITY=NEG
MODALITY=ACTUAL
DOCTIMEREL=BEFORE
bleeding

so we'll

EVENT
TYPE=ASPECTUAL
DEGREE=N/A
POLARITY=POS
MODALITY=ACTUAL
DOCTIMEREL=AFTER
resume

EVENT
TYPE=N/A
DEGREE=N/A
POLARITY=POS
MODALITY=ACTUAL
DOCTIMEREL=AFTER
chemotherapy

with a larger

EVENT
TYPE=N/A
DEGREE=N/A
POLARITY=POS
MODALITY=ACTUAL
DOCTIMEREL=AFTER
bolus

on

TIMEX3
CLASS=DATE
Friday

even if there is slight

CONTAINS

CONTAINS

EVENT
TYPE=N/A
DEGREE=LITTLE
POLARITY=POS
MODALITY=HYPOTHETICAL
DOCTIMEREL=AFTER
nausea

.

**Document
Time**

? **OVERLAP** ?
"The patient has a fever"

? **BEFORE** |
|
←
"The surgery was successful"

AFTER →
|
| ?
"She will see Dr. X in Cardiology..."

BEFORE/OVERLAP →
←
"The patient has been feeling nauseous for the last two days"

Overall Patient Timeline



Method



1. Span extraction
2. Attribute extraction
3. Relation extraction

Span Extraction



Supervised learning: CRF

Classifying tokens in text

Labels: BIO

Features: lexical, syntactic, distributional, semantic, rule-based

Window: ± 1 token

Span Extraction » Labels



"Patient taken to the operating theater on January 13, 2006 where he underwent a left-sided suboccipital craniectomy for two hours."

Tokenization

Patient_taken_to_the_operating_theater_on_
January_13_,_2006_where_he_underwent_a_ left_-_
sided_suboccipital_craniectomy_for_two_hours._

B-I-O Labeling

Patient_O taken_O to_O the_O operating_O theater_O on_O
January_B 13_I ,_I 2006_I where_O he_O underwent_O a_O
left_O -_O sided_O suboccipital_O craniectomy_O for_O
two_B hours_I ._O

Span Extraction » Features



Lexical: word, shape, lemma, casing flags, etc.

Syntactic: part-of-speech

Distributional: Brown clusters, word embeddings

Semantic: UMLS semantic types (MetaMap)

Rules: Regex

Span Extraction » Features



Feature	Examples
Date	12/3/2010, 1965-01-21
Month	January, Aug
Day	1st, 31
Day-of-week	Monday, Wed
Season	summer, spring
Year	2013, 1990s
Time	8:42, a.m.
Time Unit	minute, sec
Number	4, seventeen
Temporal preposition	in, after
Temporal adverb	daily, lately
Temporal prefix	pre, post

Attribute Extraction



Same approach as span extraction

One model per attribute

Labels: attribute values

Extended window of ± 3 tokens

Relation Extraction



Supervised learning: Gradient-boosted trees

Classification of “candidate” relation pairs

(Intra-sentence heuristic)

Labels: relation type, or null

Features: entity features, relation features

Relation Extraction » Features

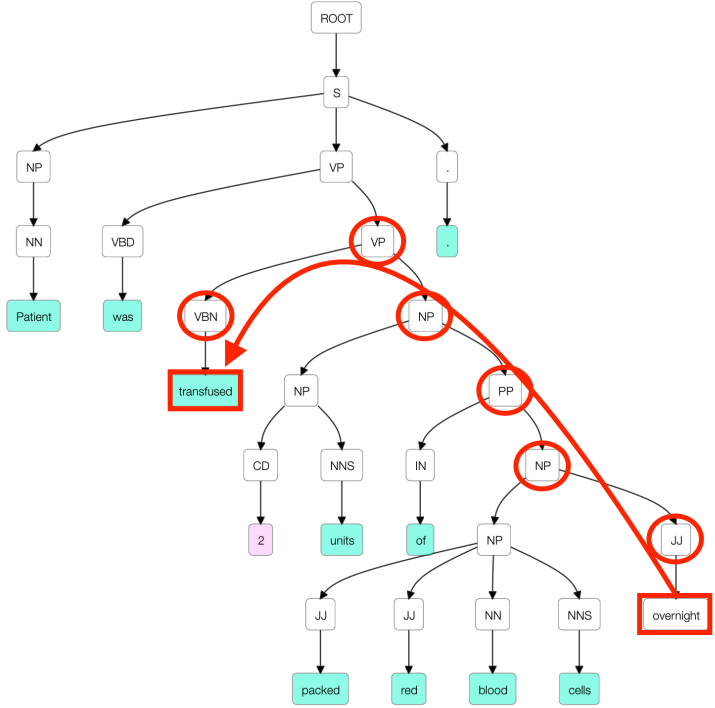


Token features described previously

Annotation attributes as features

Token distance, character distance, etc.

Constituent/dependency tree path, path distance, etc.



Results » Spans



	TIMEX3 Spans			EVENT Spans		
	P	R	F1	P	R	F1
Phase 1 (Unsupervised)						
Our System	0.61	0.53	† 0.57	0.64	‡ 0.80	‡ 0.71
Median	0.63	0.46	0.48	0.64	0.69	0.68
Phase 2 (Supervised)						
Our System	‡ 0.57	‡ 0.62	† 0.59	‡ 0.68	0.82	‡ 0.74
Median	0.53	0.52	0.54	0.67	0.76	0.71
MEMORIZE	0.64	0.22	0.33	0.61	0.51	0.56

† Top score; ‡ Second-best score

Results » Attributes



	TIMEX3 Class			EVENT Modality			EVENT Degree			EVENT Polarity			EVENT Type			
	P	R	F1	P	R	F1	P	R	F1	P	R	F1	P	R	F1	
Phase 1 (Unsupervised)																
Ours	0.55	0.47	‡0.51	0.50	0.64	0.56	0.61	‡0.77	‡0.68	0.59	‡0.74	0.65	0.61	‡0.76	‡0.68	
Median	0.56	0.45	0.46	0.55	0.63	0.59	0.62	0.71	0.68	0.60	0.70	0.66	0.61	0.70	0.66	
Phase 2 (Supervised)																
Ours	‡0.54	‡0.59	†0.56	0.60	0.72	‡0.66	‡0.67	0.80	‡0.73	0.54	0.64	0.58	‡0.66	0.79	‡0.72	
Median	0.49	0.48	0.48	0.57	0.68	0.63	0.66	0.77	0.71	0.62	0.70	0.66	0.65	0.76	0.70	
MEMORIZE	0.49	0.17	0.25	0.29	0.24	0.26	0.47	0.40	0.43	0.56	0.47	0.51	0.50	0.42	0.46	

† Top score; ‡ Second-best score

	CONTAINS			DOCTIMEREL		
	P	R	F1	P	R	F1
Phase 1 (Unsupervised)						
Our System	† 0.52	0.25	† 0.34	0.36	0.45	0.40
Median	0.33	0.25	0.32	0.39	0.45	0.41
Phase 2 (Supervised)						
Our System	† 0.59	0.16	0.25	0.45	0.55	0.50
Median	0.20	0.16	0.16	0.42	0.51	0.46
CLOSEST	0.33	0.08	0.12	-	-	-
MEMORIZE	-	-	-	0.22	0.18	0.20

† Top score; ‡ Second-best score

New date conventions (e.g. 12Jun2013)

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“Cancer” as EVENT in “Cancer Research Hospital”

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Domain-specific overfitting –
“temozolomide”

Questions?