

Joint Learning for Event Coreference Resolution

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Nelson Mandela has died at age 95.

The world has lost a great man, former Prime Minister Julia Gillard said.





Trigger Detection: Determine trigger word and subtype

```
Nelson Mandela has died at age 95.

The world has lost a great man, former Prime Minister Julia Gillard said.

Life.Die Personnel.End-Position Contact.Broadcast
```





Anaphoricity Detection: Determine if an event mention is anaphoric

Not Anaphoric

Nelson Mandela has died at age 95.
The world has lost a great man, former Prime Minister Julia Gillard said.

Anaphoric

Not Anaphoric

Not Anaphoric



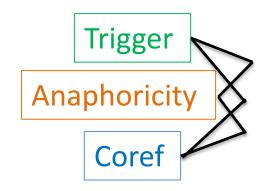


 Coreference: Determine which event mentions refer to the same realworld event

Nelson Mandela has died at age 95.
The world has lost a great man, former Prime Minister Julia Gillard said.







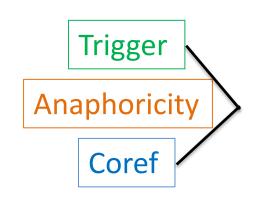
These three event subtasks are inter-dependent





Trigger-Coreference Dependencies

Nelson Mandela has died aged 95 surrounded



The world has lost a great man...

Life.Die Personnel.End-Position Life.Injury NULL

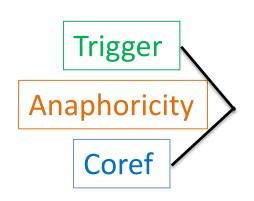




Trigger-Coreference Dependencies

Life.Die

Nelson Mandela has died aged 95 surrounded



The world has lost a great man...

Life.Die Personnel.End-Position Life.Injury NULL

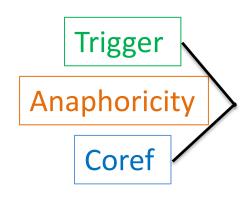




Trigger-Coreference Dependencies



Nelson Mandela has died aged 95 surrounded



The world has lost a great man...

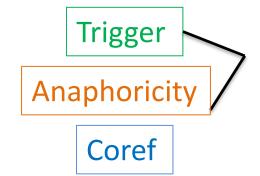
Life.Die Personnel.End-Position Life.Injury NULL

Coreferent event mentions must have same subtype





Trigger-Anaphoricity Dependencies



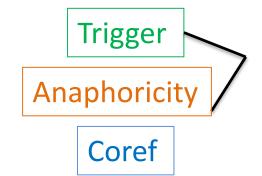
The world has lost a great man...

Life.Die Personnel.End-Position Life.Injury NULL





Trigger-Anaphoricity Dependencies



The world has lost a great man...

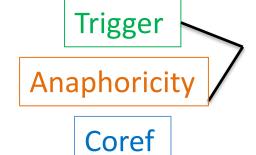
Life.Die Personnel.End-Position Life.Injury NULL

Anaphoric





Trigger-Anaphoricity Dependencies



The world has lost a great man...

Life.Die Personnel.End-Position Life.Injury

.

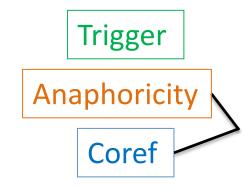
Anaphoric

Anaphoric event mentions cannot have subtype NULL





Coref-Anaphoricity Dependencies



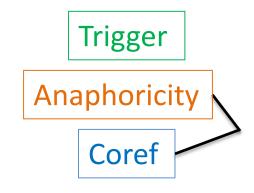
The world has lost a great man, former...

Non-anaphoric



HO SYLINGON OF SE SO WOOM

Coref-Anaphoricity Dependencies



The world has lost a great man, former...

Non-anaphoric

Non-anaphoric event mentions cannot have antecedents





Top Event Coreference Resolvers in KBP'16

Largely capture inter-dependencies with a pipeline architecture



Weakness: error propagation





Goal

- Improve event coreference performance by jointly learning three subtasks: trigger detection, anaphoricity detection, and coreference
 - Motivated by Durrett and Klein's (2014) model and its success in jointly modeling the key tasks in the entity stack
 - Event anaphoricity detection introduced as an auxiliary task to mediate the interactions between trigger detection and coreference





Plan for the Talk

- Models
 - Independent model (unary, per-task factors)

Trigger

Anaphoricity

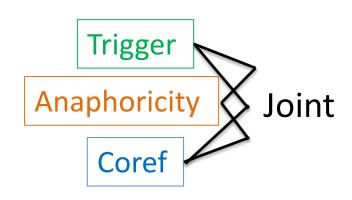
Coref





Plan for the Talk

- Models
 - Independent model (unary, per-task factors)
 - Joint model (higher-order, cross-task factors)

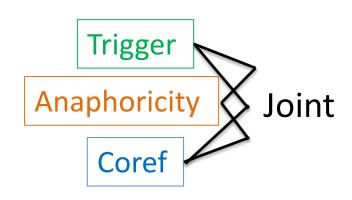






Plan for the Talk

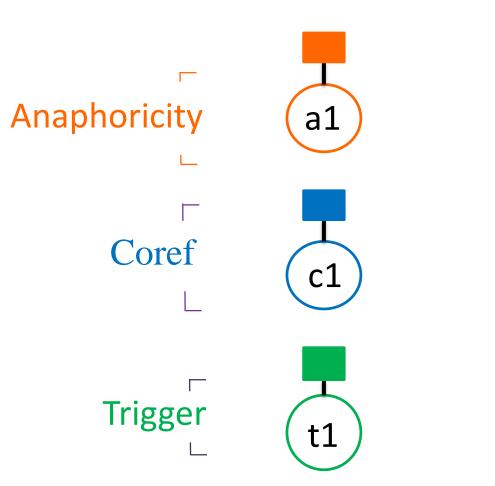
- Models
 - Independent model (unary, per-task factors)
 - Joint model (higher-order, cross-task factors)
- Evaluation

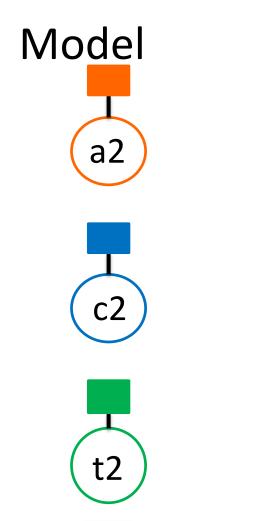


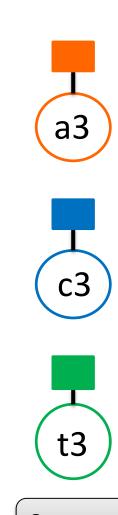


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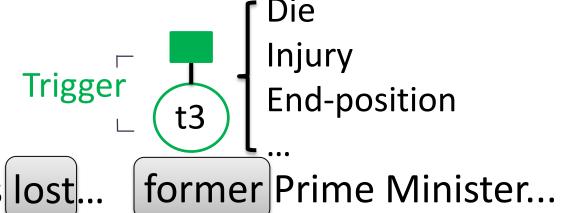


Mandela has died...The world has lost...

former Prime...



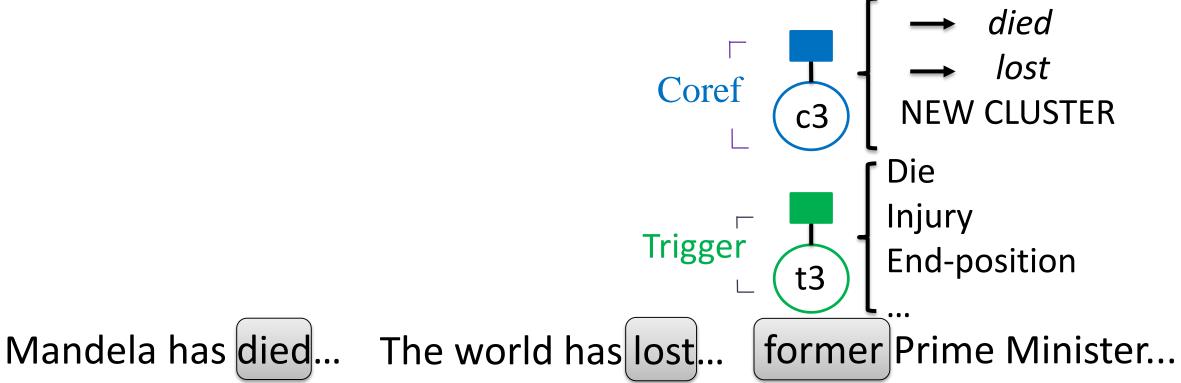




Mandela has died... The world has lost...

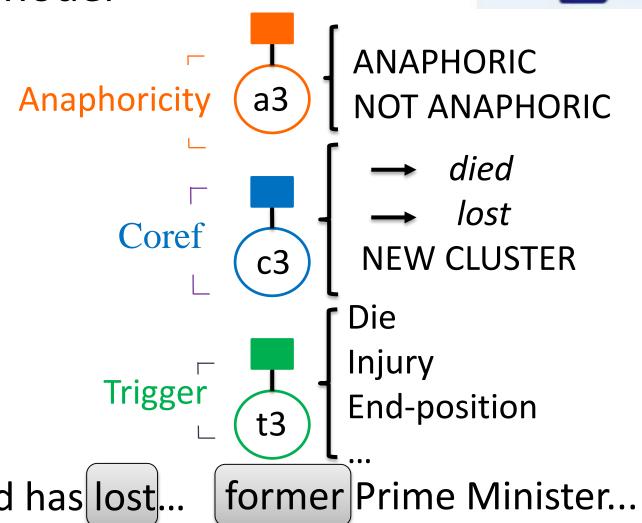












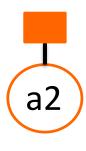
Mandela has died...



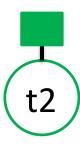


Log-linear probability distribution over the variables

$$p(t, a, c | x; \Theta) \propto \exp(\sum_{i} \theta_{i} f_{i}(t, a, c, x))$$



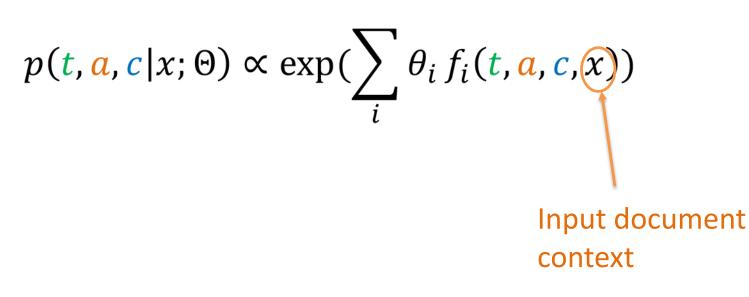


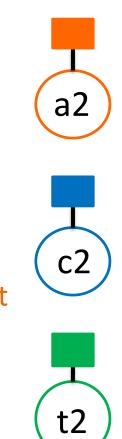






Log-linear probability distribution over the variables

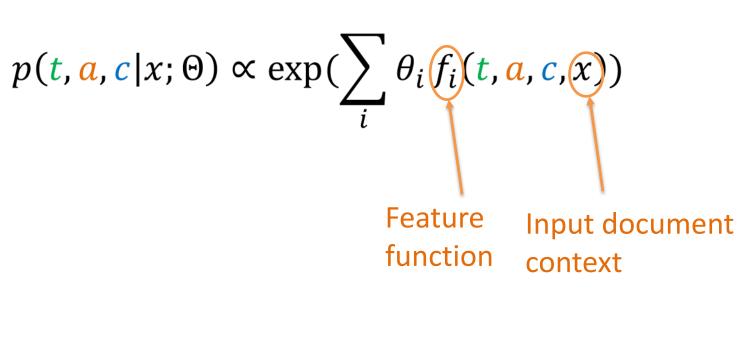


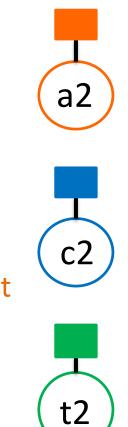






Log-linear probability distribution over the variables







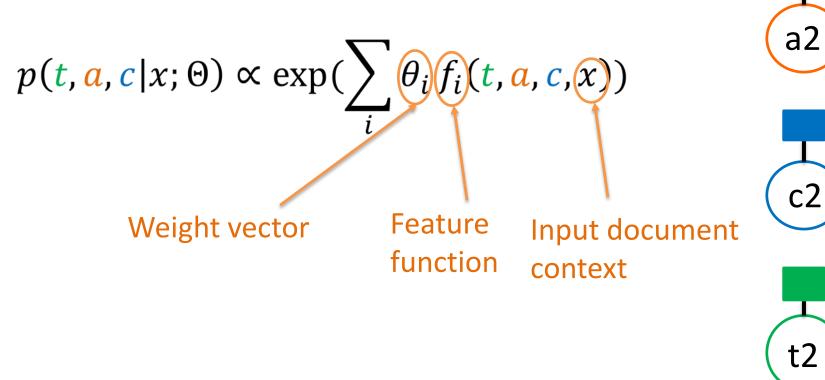


c2

t2

Model

Log-linear probability distribution over the variables



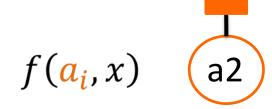


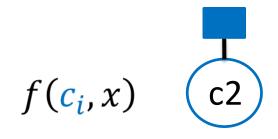


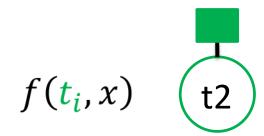
Independent Model

$$p(t, \boldsymbol{a}, \boldsymbol{c} | \boldsymbol{x}; \boldsymbol{\Theta}) \propto \exp(\boldsymbol{\Theta}^T \boldsymbol{F}(t, \boldsymbol{a}, \boldsymbol{c}, \boldsymbol{x}))$$

$$\mathbf{F}_{Indep} = \sum_{i=1}^{n} f(\mathbf{a}_i, x) + f(\mathbf{c}_i, x) + f(\mathbf{t}_i, x)$$













The world has lost a great man, former Prime Minister...







Trigger word	lost







Trigger word Trigger Lemma	lost lose







Trigger word	lost
Trigger Lemma	lose
Bigram	has_lost, lost_a







The world has lost a great man, former Prime Minister...

Trigger word	lost
Trigger Lemma	lose
Bigram	has_lost, lost_a
Nearest Entity (Left)	world
Nearest Entity (Right)	man







The world has lost a great man, former Prime Minister...

Trigger word	lost
Trigger Lemma	lose
Bigram	has_lost, lost_a
Nearest Entity (Left)	world
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Dependencies	World, man





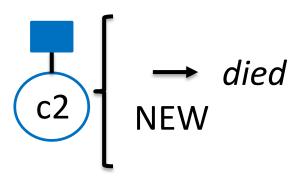


Trigger word	lost
Trigger Lemma	lose
Bigram	has_lost, lost_a
Nearest Entity (Left)	world
Nearest Entity (Right)	man
Dependencies	World, man





Event Coreference

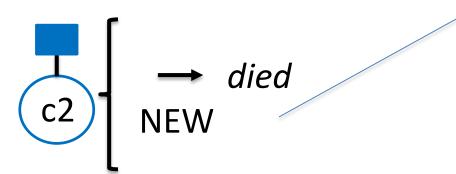






Event Coreference Features

Features representing the current mention



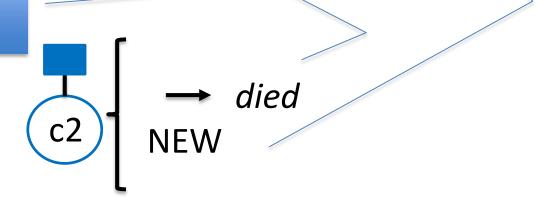




Event Coreference Features

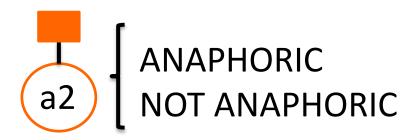
Features representing the current mention

Features representing a preceding mention and its relationship with the current mention





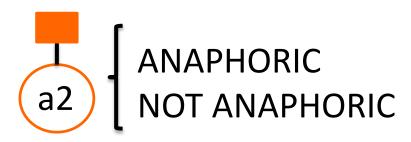




The world has lost a great man, former Prime Minister...





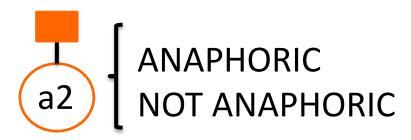


The world has lost a great man, former Prime Minister...

Trigger words of preceding mentions died





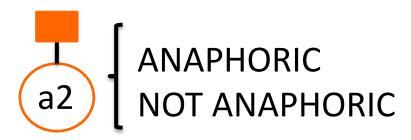


The world has lost a great man, former Prime Minister...

Trigger words of preceding mentions died Same lemma No







The world has lost a great man, former Prime Minister...

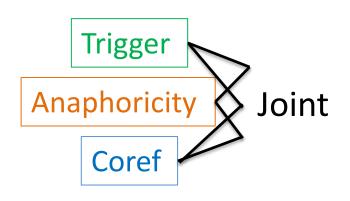
Trigger words of preceding mentions died Same lemma No Anaphoricity probability 0.75





Plan for the Talk

- Models
 - Independent model (unary, per-task factors)
 - Joint model (higher-order, cross-task factors)
- Evaluation



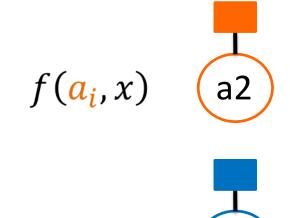


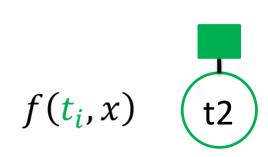


Joint Model

$$p(t, \boldsymbol{a}, \boldsymbol{c} | \boldsymbol{x}; \boldsymbol{\Theta}) \propto \exp(\boldsymbol{\Theta}^T \boldsymbol{F}(t, \boldsymbol{a}, \boldsymbol{c}, \boldsymbol{x}))$$

$$\mathbf{F}_{Indep} = \sum_{i=1}^{n} f(\mathbf{a}_i, x) + f(\mathbf{c}_i, x) + f(\mathbf{t}_i, x)$$







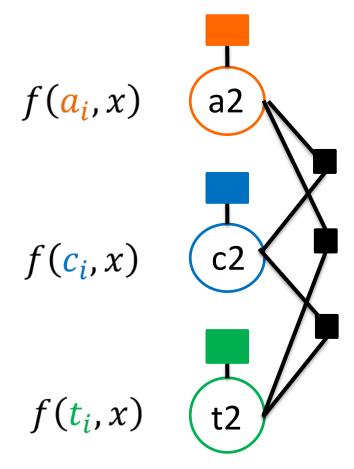


Joint Model

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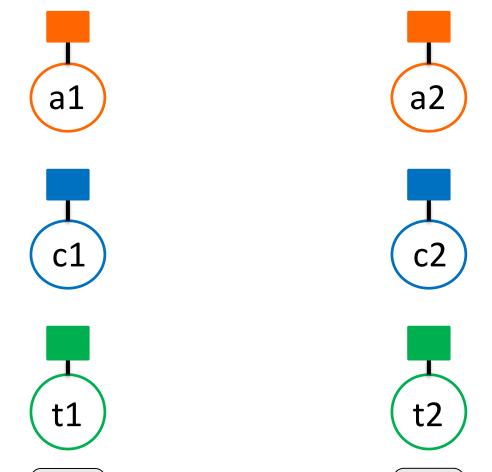
$$\mathbf{F}_{Indep} = \sum_{i=1}^{n} f(\mathbf{a}_i, x) + f(\mathbf{c}_i, x) + f(\mathbf{t}_i, x)$$

$$F_{Joint} = F_{Indep} + \sum_{i} joint factors$$



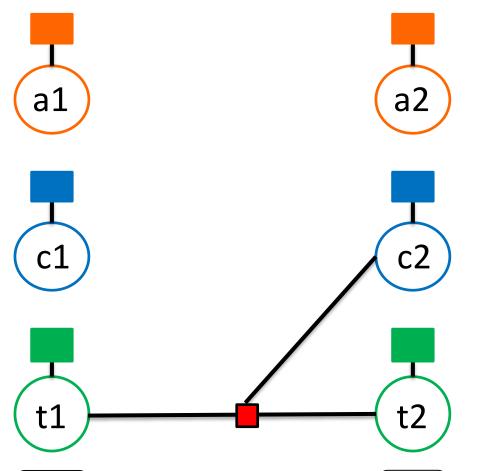






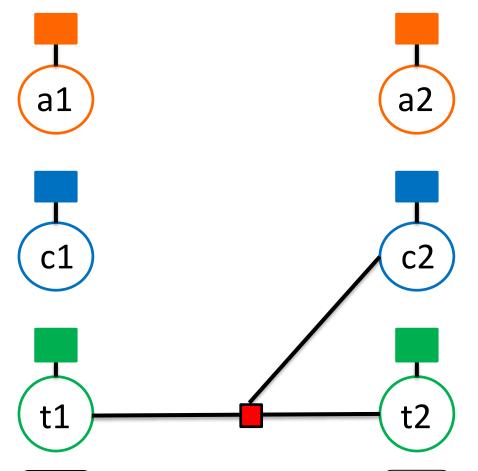








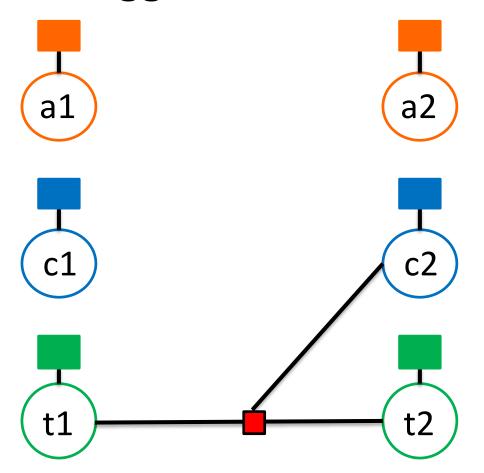




Coreferent mentions should have the same subtype





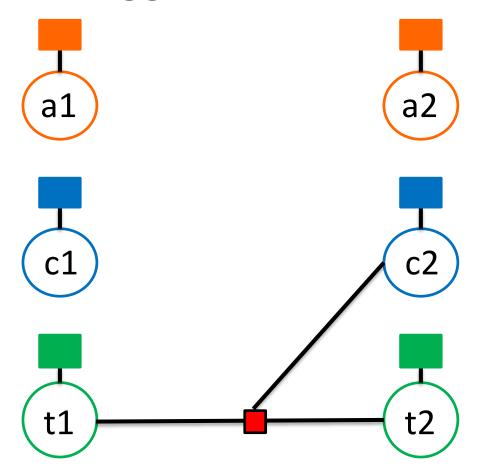


Coreferent mentions should have the same subtype

Subtypes=Die-Die







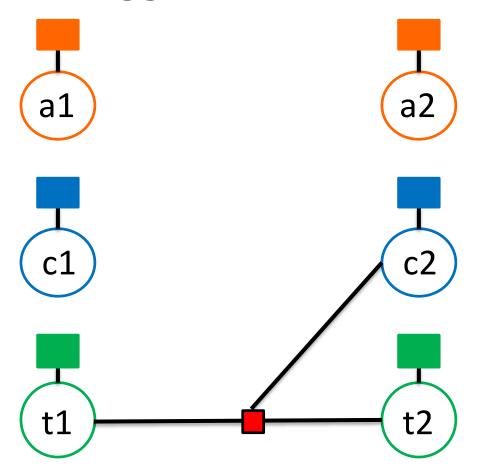
Coreferent mentions should have the same subtype

Subtypes=Die-Die CurrType=Die, Prev = died

Mandela has died... The world has lost...







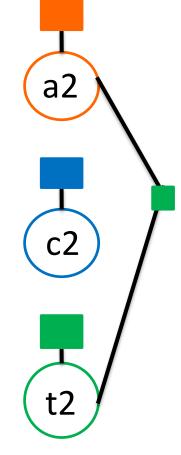
Coreferent mentions should have the same subtype

Subtypes=Die-Die CurrType=Die, Prev = died PrevType=Die, Curr = lost





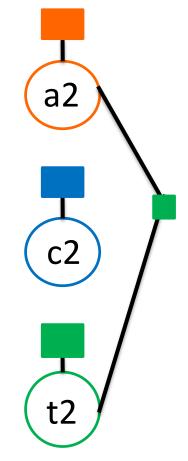
Trigger - Anaphoricity Factors







Trigger - Anaphoricity Factors

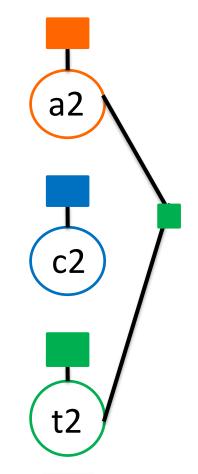


Anaphoric event mention cannot have subtype NULL





Trigger - Anaphoricity Factors



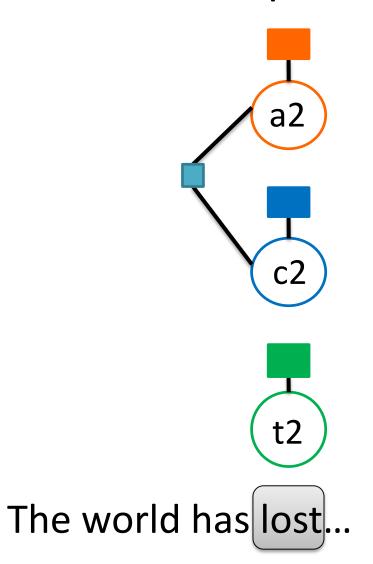
Anaphoric event mention cannot have subtype NULL

Subtypes=Die, Lemma=lose, Anaphoricity = Yes





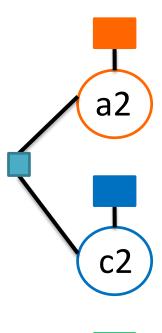
Coref - Anaphoricity Factors



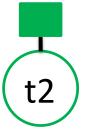




Coref - Anaphoricity Factors



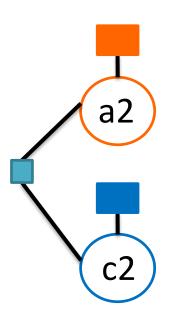
Anaphoric event mentions should have antecedents





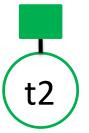


Coref - Anaphoricity Factors



Anaphoric event mentions should have antecedents

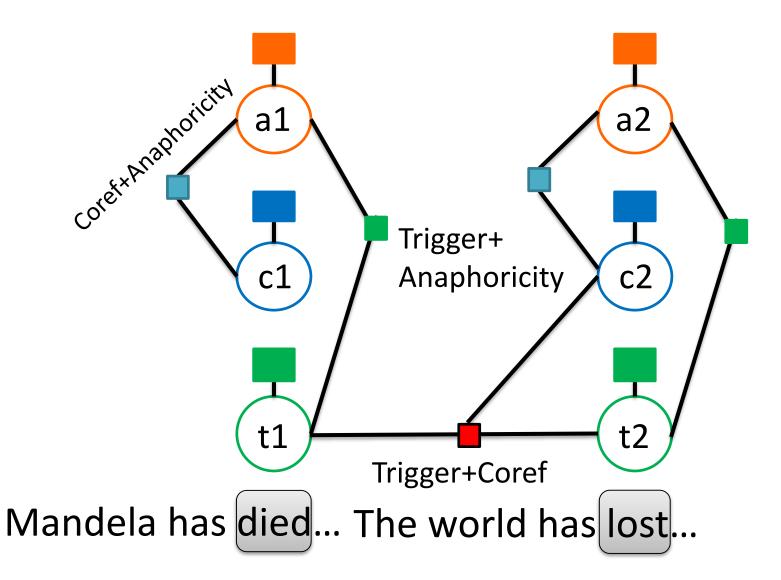
Anaphoricity = Yes, antecedent = Yes







Joint Model







Training

Log-linear probability distribution over the variables

$$p(t, a, c | x; \Theta) \propto \exp(\sum_{i} \theta_{i} f_{i}(t, a, c, x))$$

Conditional log-likelihood of the training data with L1 regularization

$$L(\Theta) = \sum_{i=1}^{d} log \sum_{\substack{\text{latent} \\ \text{stuff}}} p'(t_i^*, a_i^*, c_i^* | x_i; \Theta) + \lambda \|\Theta\|_1$$





Training

Log-linear probability distribution over the variables

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$$p'(t, a, c|x_i; \theta) \propto p(t, a, c|x_i; \theta) \exp[\alpha_t l_t(t, t^*) + \alpha_a l_a(a, a^*) + \alpha_c l_c(c, C^*)]$$





Training

Log-linear probability distribution over the variables

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Conditional log-likelihood of the training data with L1 regularization

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Inference: Belief Propagation

- computes the marginals for a variable or a set of variables to which a factor connects
- encourages compatibility among component models
- To improve scalability
 - Restrict the domains of the coreference variables
 - Restrict the domains of the subtype variables





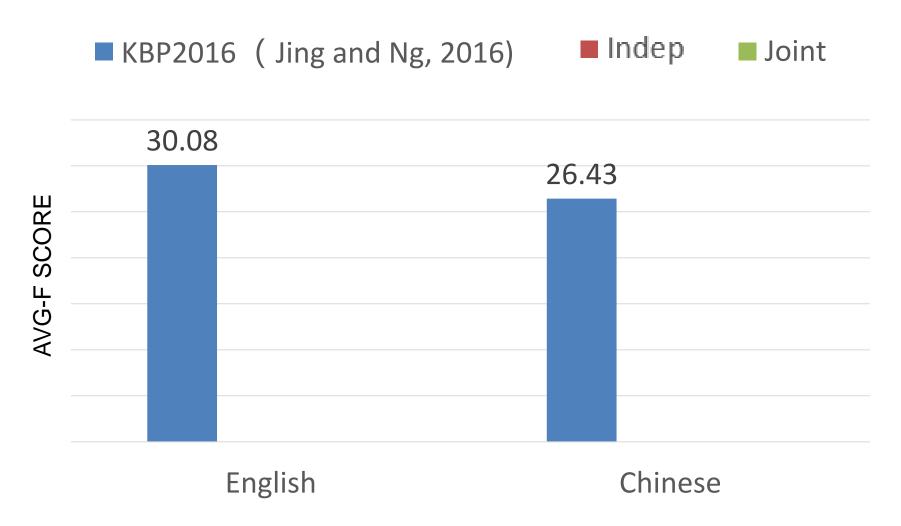
Evaluation

- Dataset:
 - KBP 2016 English and Chinese corpora
- Evaluation metrics:
 - Event Coreference: MUC, B₃, CEAF_e, BLANC, and average of their F-scores
 - Trigger detection: F-score
 - Anaphoricity detection: F-score





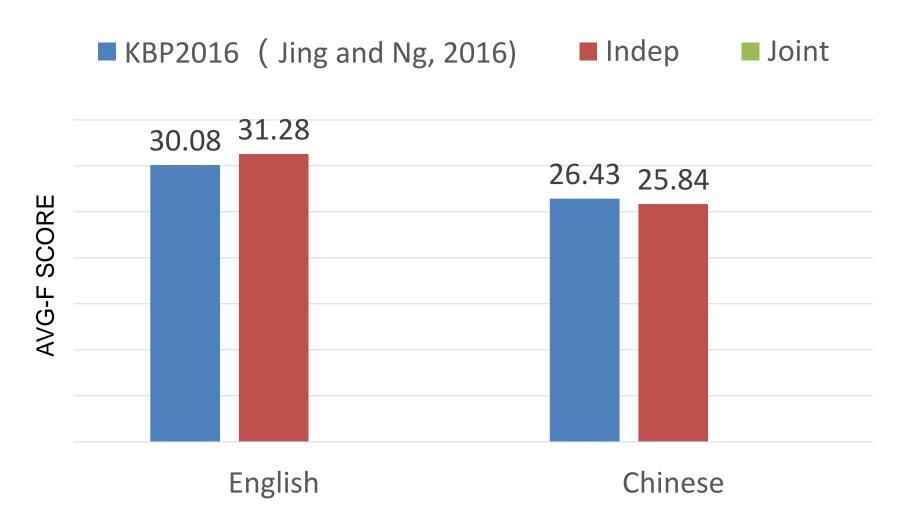
Event Coreference Results







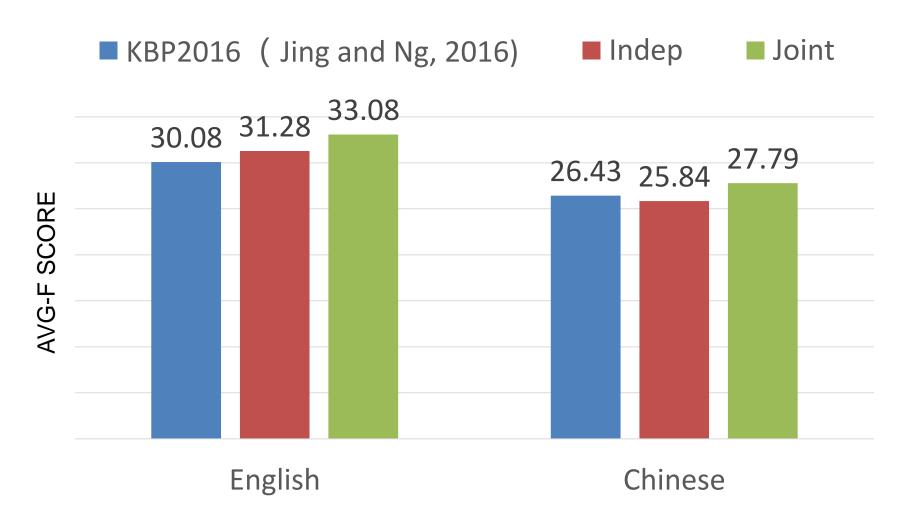
Event Coreference Results







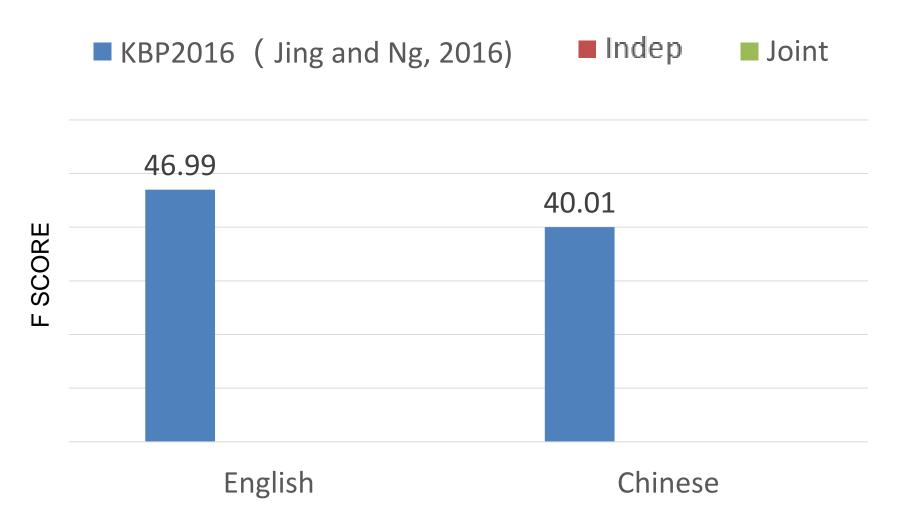
Event Coreference Results







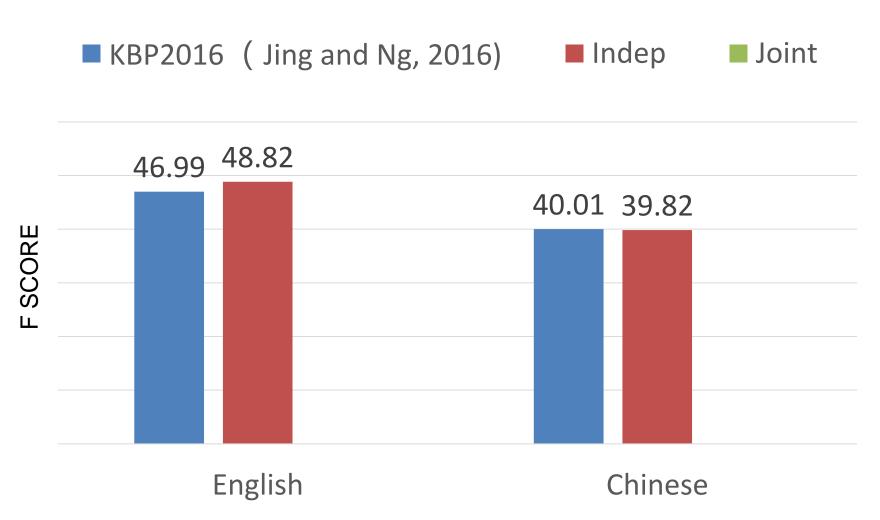
Trigger Detection Results







Trigger Detection Results

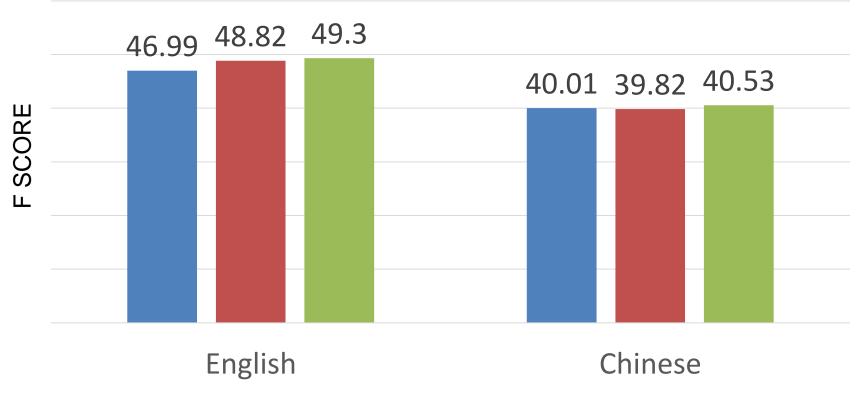






Trigger Detection Results

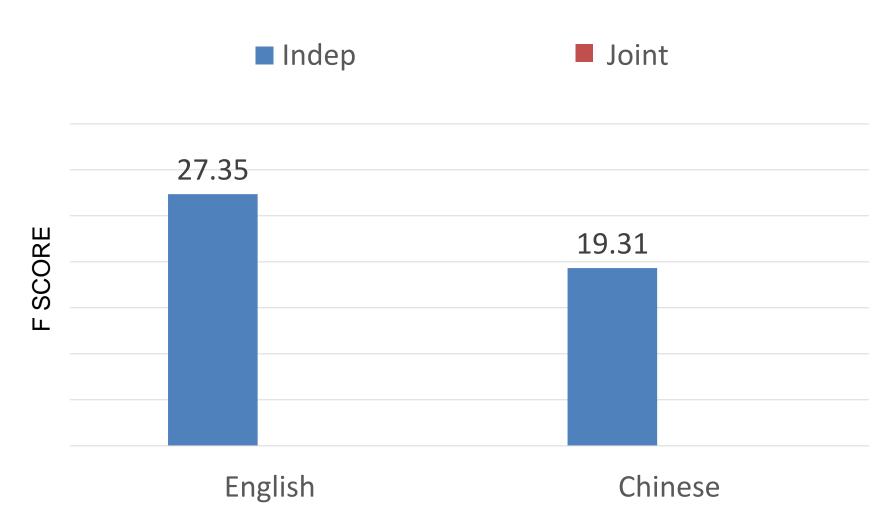








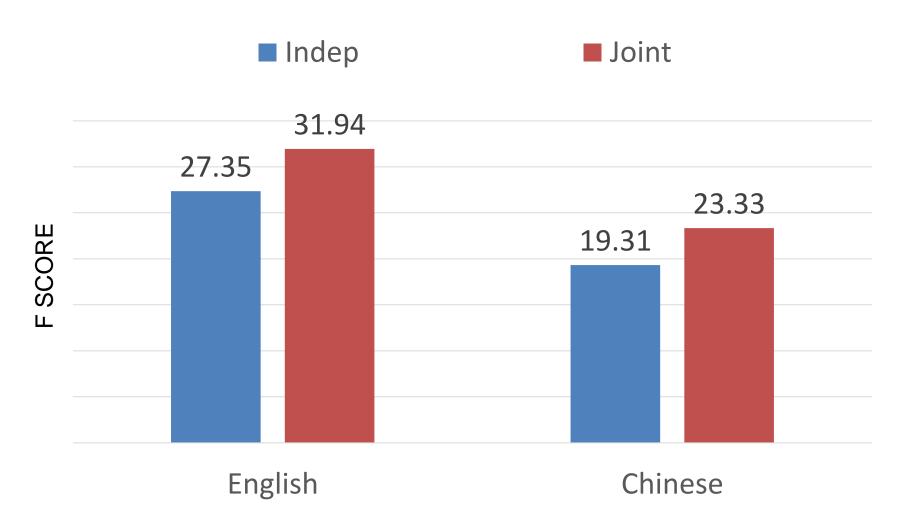
Anaphoricity Detection Results







Anaphoricity Detection Results







Conclusion

- Joint predictions on trigger detection, event coreference resolution, and event anaphoricity
- Joint models effectively improve the performance of event coreference resolution
 - achieves the best results to date on KBP 2016 English and Chinese corpora





Model Ablation

- evaluate the importance of each of the three types of joint factors
 - add each type of joint factors to the independent model
 - remove each type of joint factors from the full joint model





Model Ablation: Coref-Trigger interactions

	English			Chinese			
	Coref	Trigger	Anaph	Coref	Trigger	Anaph	
Indep.	31.28	48.82	27.35	25.84	39.82	19.31	
INDEP.+CorefTrigger	+0.39	+0.42	-0.05	+0.95	+0.56	-0.37	
INDEP.+CorefAnaph	+0.40	-0.08	+3.45	+0.37	+0.04	-0.11	
INDEP.+TriggerAnaph	+0.11	+0.38	+1.35	+0.14	+0.52	+0.02	
JOINT—CorefTrigger	+0.56	-0.06	+4.41	+0.19	+0.35	+3.34	
JOINT—CorefAnaph	+0.63	+0.66	+1.46	+1.50	+0.88	+0.28	
JOINT—TriggerAnaph	+1.89	+0.50	+4.01	+1.65	+0.50	+1.79	
JOINT	+1.80	+0.48	+4.59	+1.95	+0.71	+4.02	

Contributes the most to coreference performance, and trigger detection



Model Ablation: Coref-Anaphoricity interactions

	English			Chinese		
	Coref	Trigger	Anaph	Coref	Trigger	Anaph
INDEP.	31.28	48.82	27.35	25.84	39.82	19.31
INDEP.+CorefTrigger	+0.39	+0.42	-0.05	+0.95	+0.56	-0.37
INDEP.+CorefAnaph	+0.40	-0.08	+3.45	+0.37	+0.04	-0.11
INDEP.+TriggerAnaph	+0.11	+0.38	+1.35	+0.14	+0.52	+0.02
JOINT—CorefTrigger	+0.56	-0.06	+4.41	+0.19	+0.35	+3.34
JOINT—CorefAnaph	+0.63	+0.66	+1.46	+1.50	+0.88	+0.28
JOINT—TriggerAnaph	+1.89	+0.50	+4.01	+1.65	+0.50	+1.79
JOINT	+1.80	+0.48	+4.59	+1.95	+0.71	+4.02

In isolation: improve coreference performance

In combination: improves both tasks, particularly anaphoricity determination



Model Ablation: Trigger-Anaphoricity interactions

	English			Chinese		
	Coref	Trigger	Anaph	Coref	Trigger	Anaph
INDEP.	31.28	48.82	27.35	25.84	39.82	19.31
INDEP.+CorefTrigger	+0.39	+0.42	-0.05	+0.95	+0.56	-0.37
INDEP.+CorefAnaph	+0.40	-0.08	+3.45	+0.37	+0.04	-0.11
INDEP.+TriggerAnaph	+0.11	+0.38	+1.35	+0.14	+0.52	+0.02
JOINT—CorefTrigger	+0.56	-0.06	+4.41	+0.19	+0.35	+3.34
JOINT—CorefAnaph	+0.63	+0.66	+1.46	+1.50	+0.88	+0.28
JOINT—TriggerAnaph	+1.89	+0.50	+4.01	+1.65	+0.50	+1.79
JOINT	+1.80	+0.48	+4.59	+1.95	+0.71	+4.02

In isolation: improve both tasks

In combination: improves anaphoricity determination





Error Analysis

- Precision Error
 - Erroneous and mistyped triggers
 - assign the same subtype to event mentions triggered by the same word.
 - assigns the wrong subtype to triggers that possess different subtypes in different contexts
 - » Example: trigger "leave" sometimes has the subtype Movement. Transport-Person and sometimes has the subtype Personnel. End-Position.
 - words that are only sometimes used as triggers are often wrongly posited as triggers when they are not
 - Failure to extract arguments
 - we only exploit dependency relations for argument extraction
 - Sometimes arguments that do not appear in the same sentence as their trigger.





Error Analysis

Recall Error

- Missing triggers
 - fails to identify trigger words that are unseen or rarely occurring in the training data.
- Lack of entity coreference information.
 - the corresponding arguments of two coreferent event mentions are typically coreferent
 - This issue is particularly serious in discussion forum documents,
 - pronouns often serve as subjects and objects of event mentions
 - More serious in Chinses documents due to the zero pronouns
- Lack of contextual understanding
 - Additional contextual information present in neighboring sentences is not used
 - when describing the fact that Tim Cook will attend Apple's Istanbul store opening
 - "Cook is expected to return to Turkey for the store opening"
 - "Tim travels abroad YET AGAIN to be feted by the not-so-high-and-mighty".