



SinoCoreferencer: An End-to-End Chinese Event Coreference Resolver

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Event Coreference Resolution

- Determine which event mentions in a text refer to the same real-world event

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Since there is little work on event coreference, our understanding of this task is fairly limited

Goal

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 - An event coreference system lies towards the end of the standard information extraction pipeline
 - To what extent is the noisy output of each of its upstream components limiting the performance?

Focus on Chinese event coreference resolution

Why **Chinese** Event Coreference?

- Provide publicly available **results**
 - related work mostly on English event coreference
 - Humphreys et al.(1997), Chen et al.(2009), Bejan and Harabagiu(2010), Chen et al.(2011), Lee et al.(2012), ...

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 - Humphreys et al.(1997), Chen et al.(2009), Bejan and Harabagiu(2010), Chen et al.(2011), Lee et al.(2012), ...
- Provide a publicly available **implementation**
 - **SinoCoreferencer** contains the major components of a typical information extraction pipeline
 - Each component can be run in a standalone manner
 - Complements the Stanford Chinese NLP tools

Plan for the Talk

- ACE Event Coreference
- System Architecture
- Evaluation

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- Three event mentions: [injured], [stabbed], [criminal]
 - [stabbed] and [criminal] are coreferent because they refer to the same real-world event

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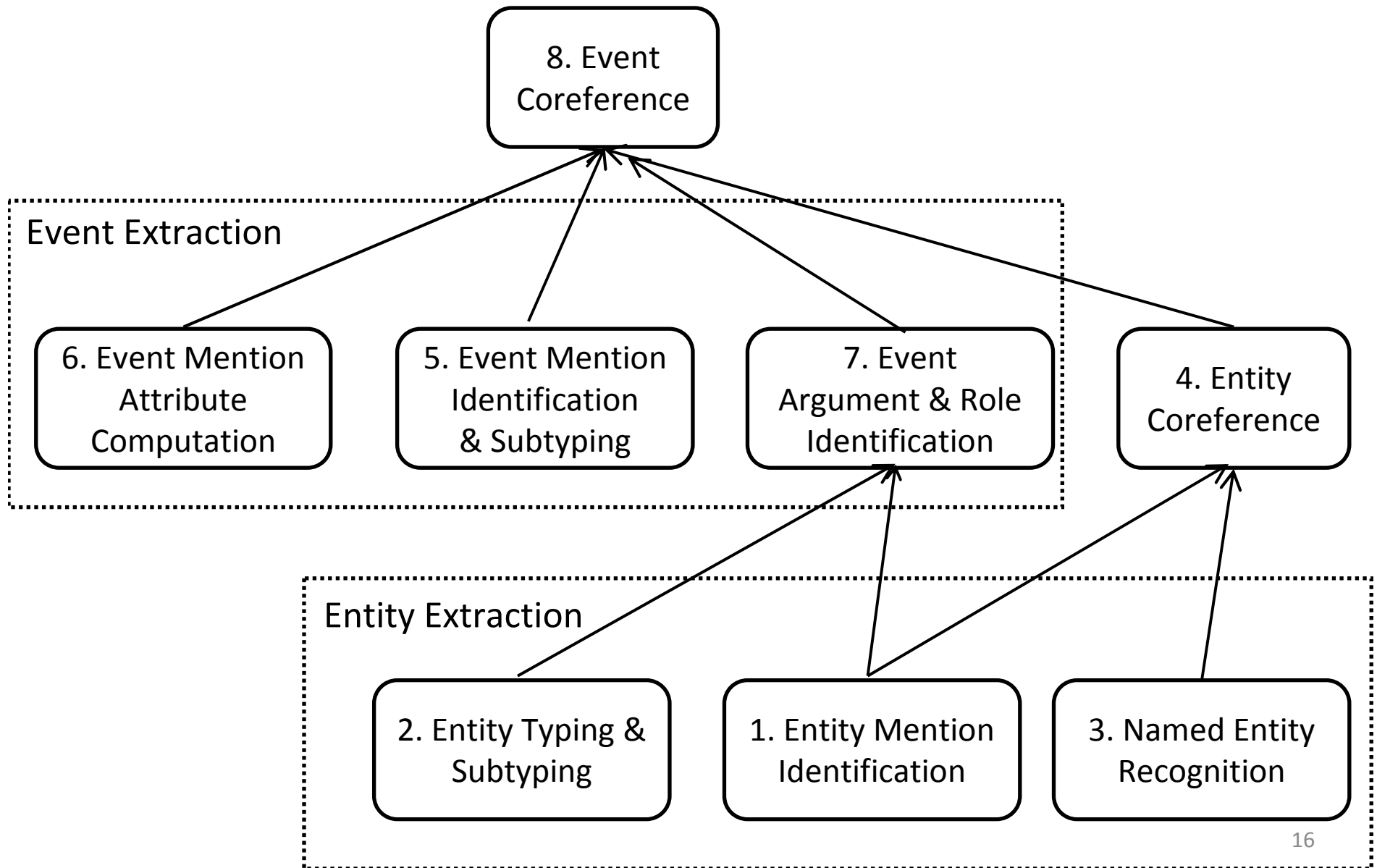
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- **Restricted** event coreference
 - Considers only event mentions belonging to certain types
 - ACE 2005: 7 event types and 33 event subtypes

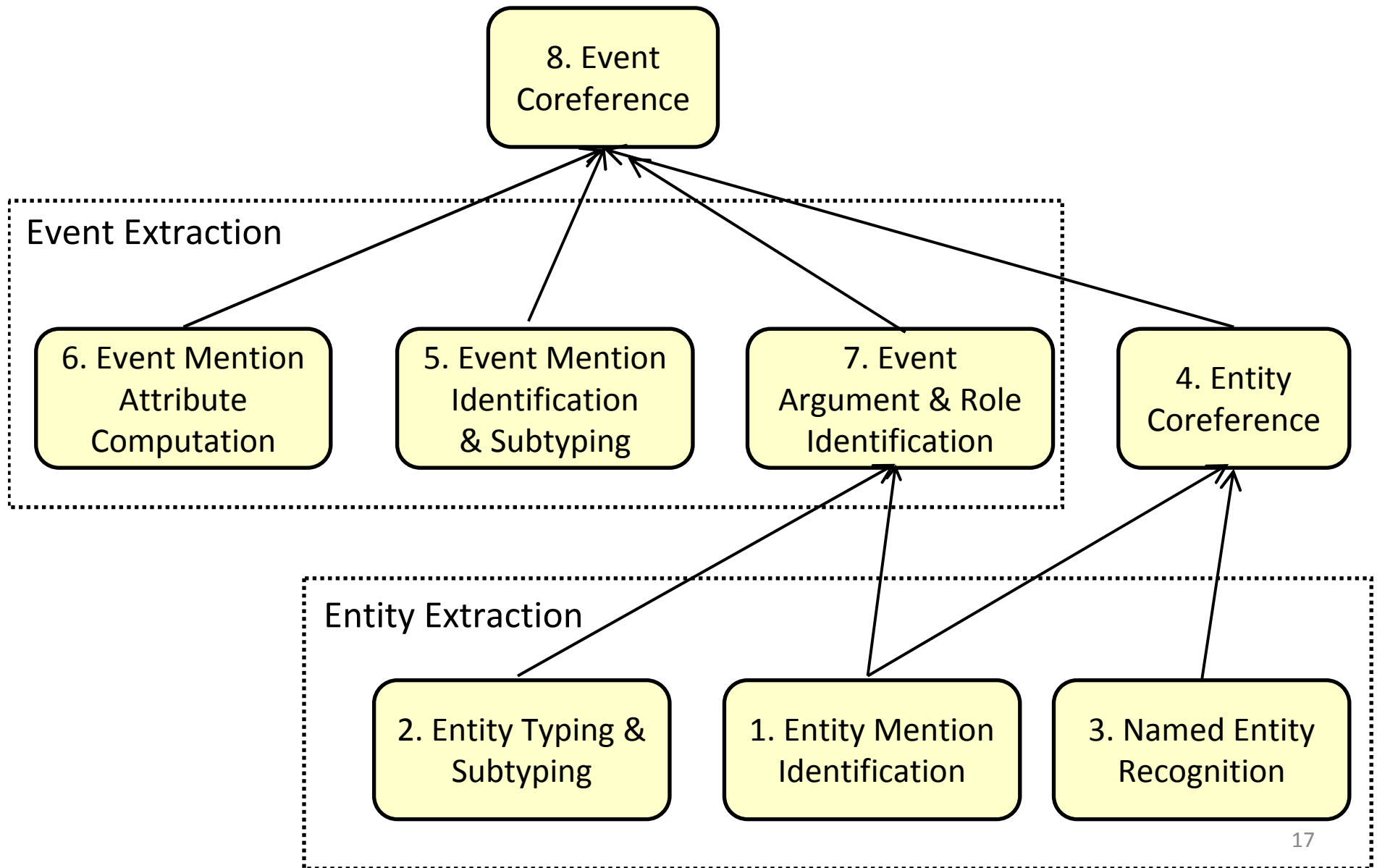
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- **System Architecture**
- Evaluation

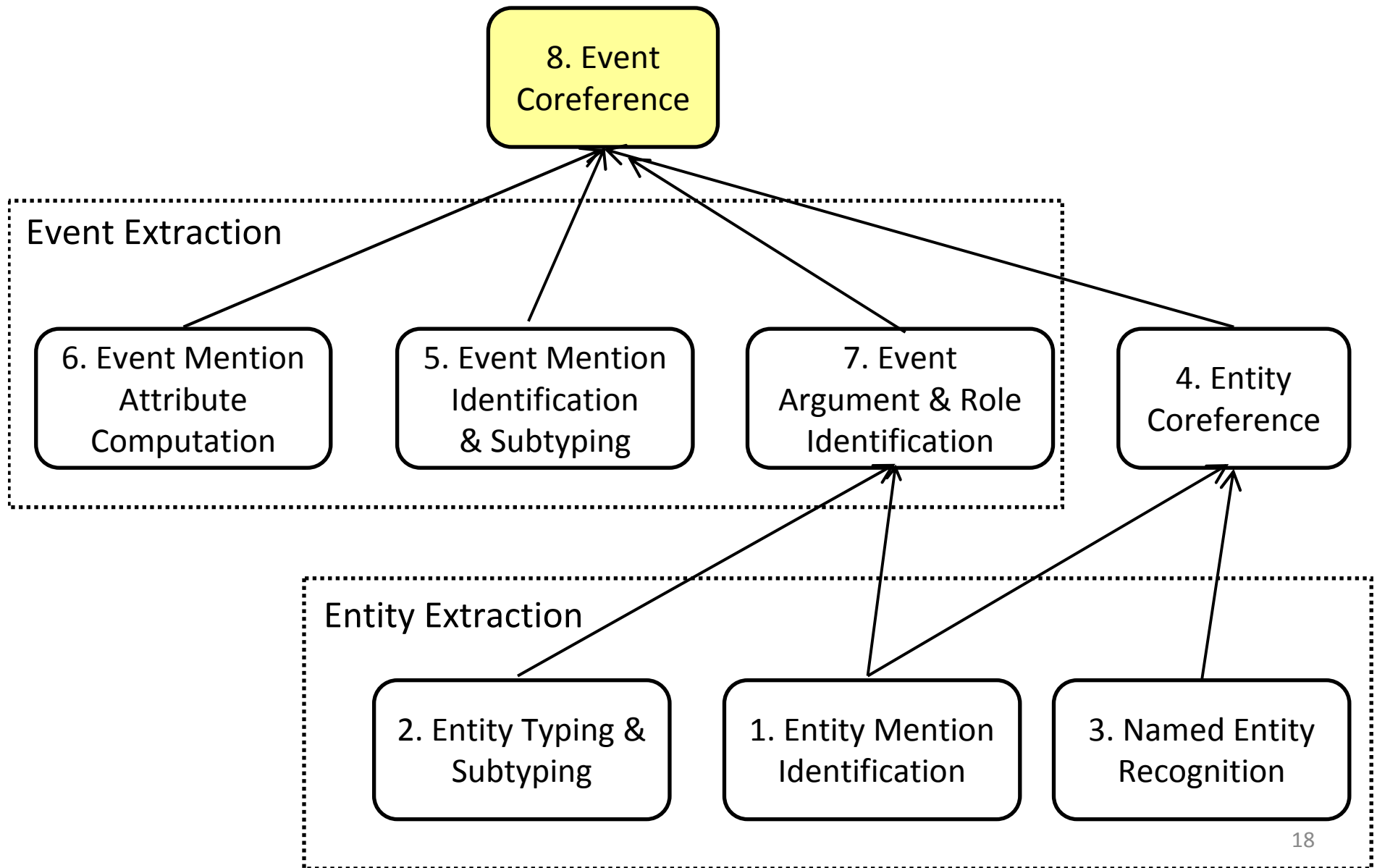
SinoCoreferencer: System Architecture



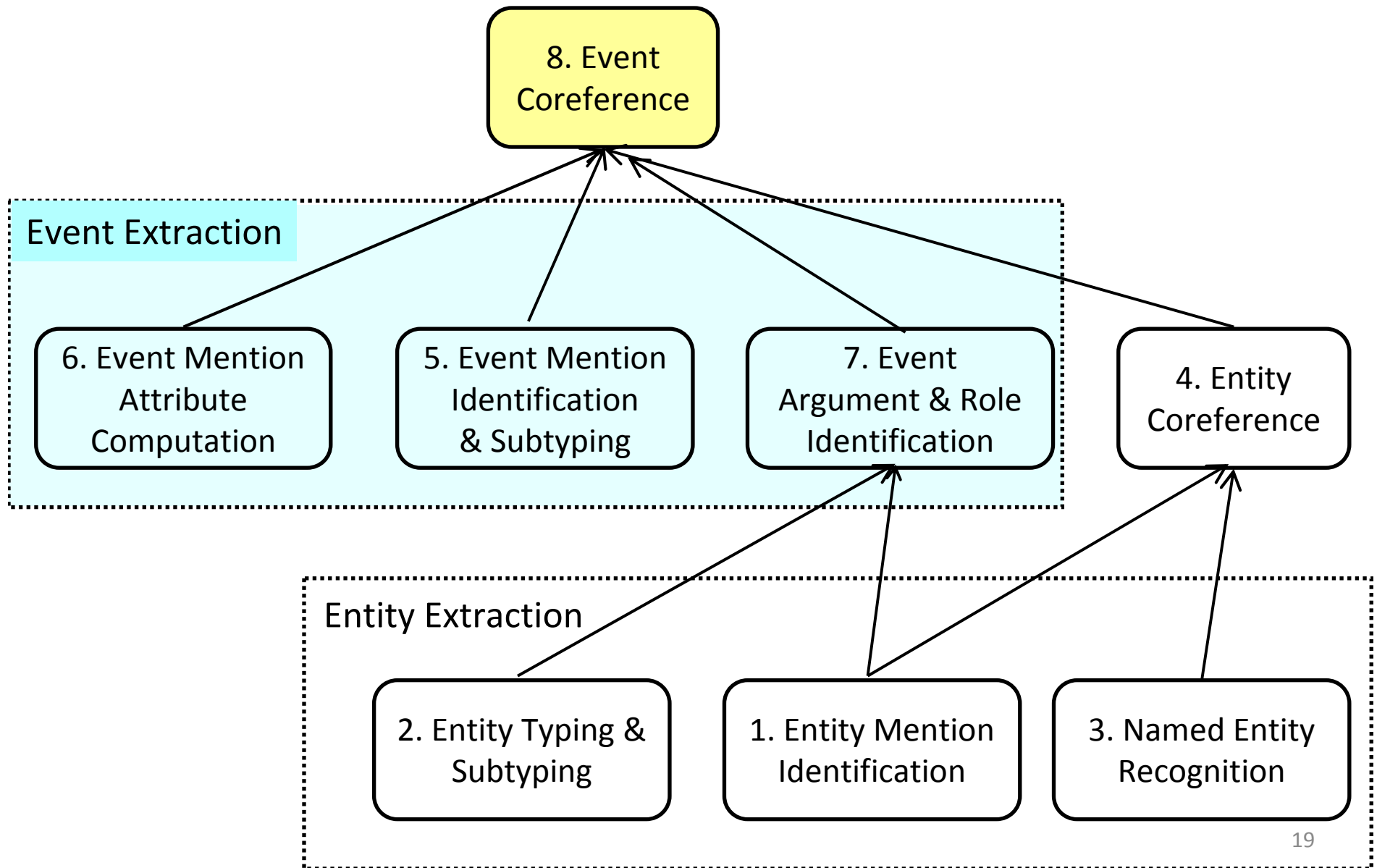
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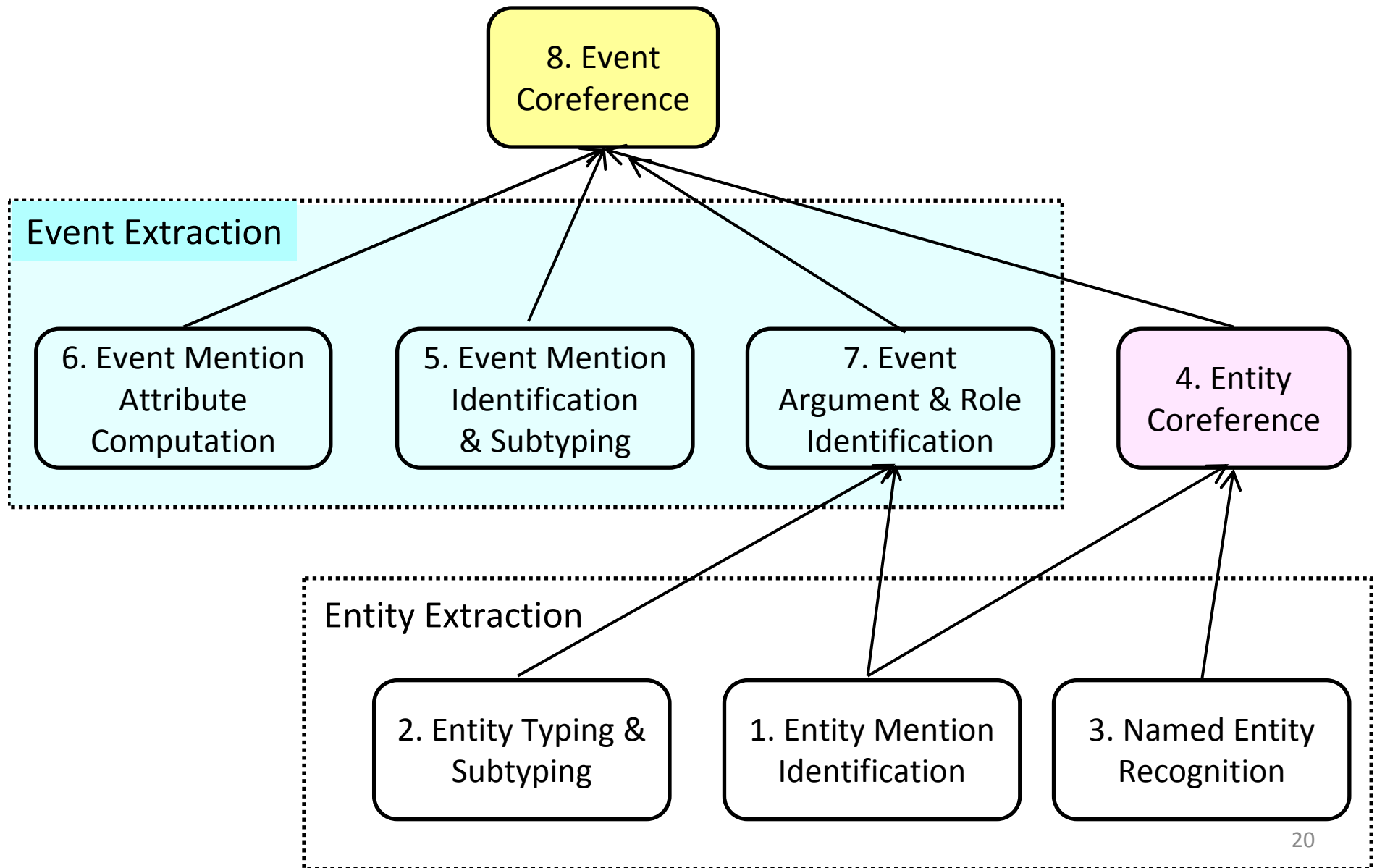
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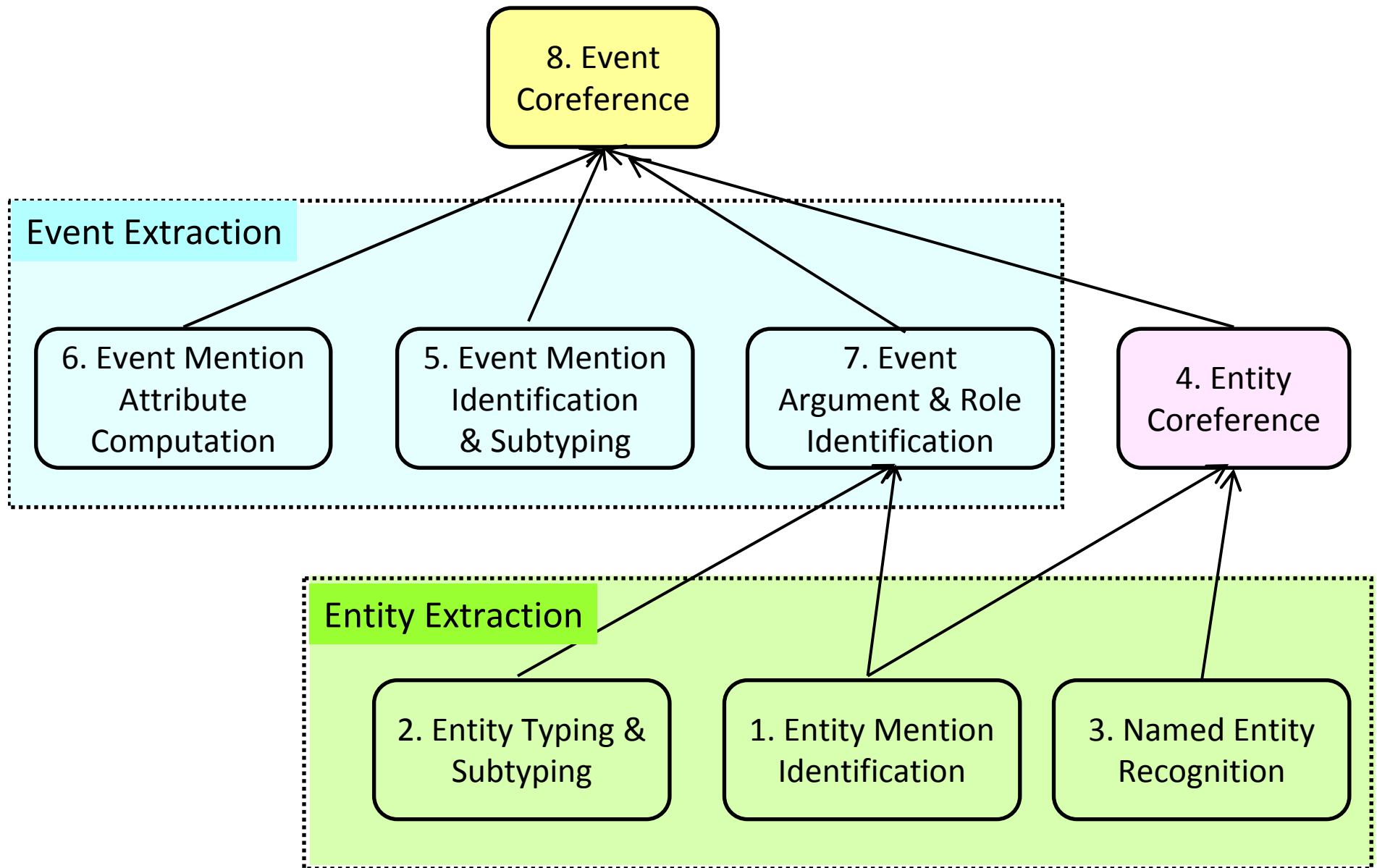
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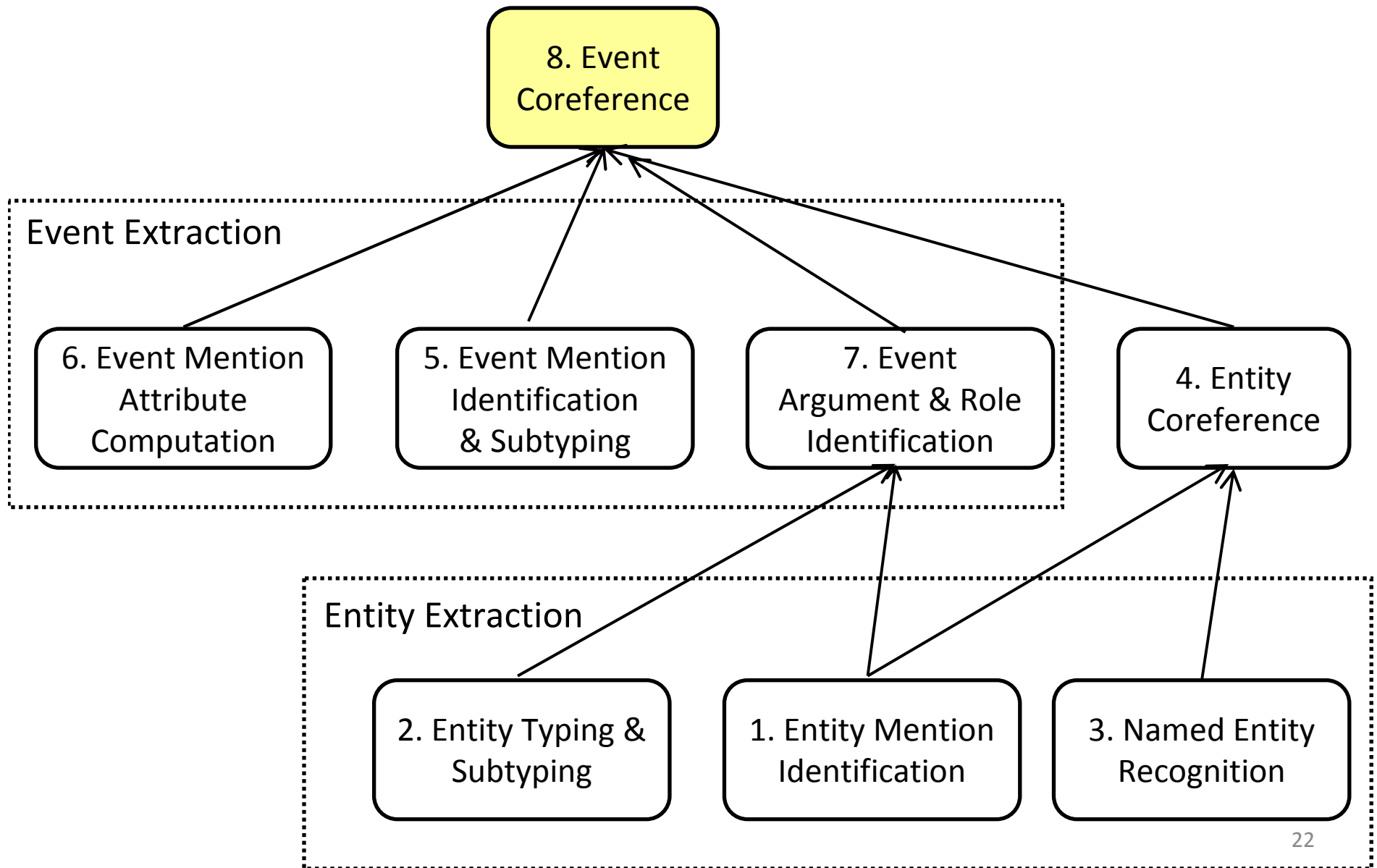
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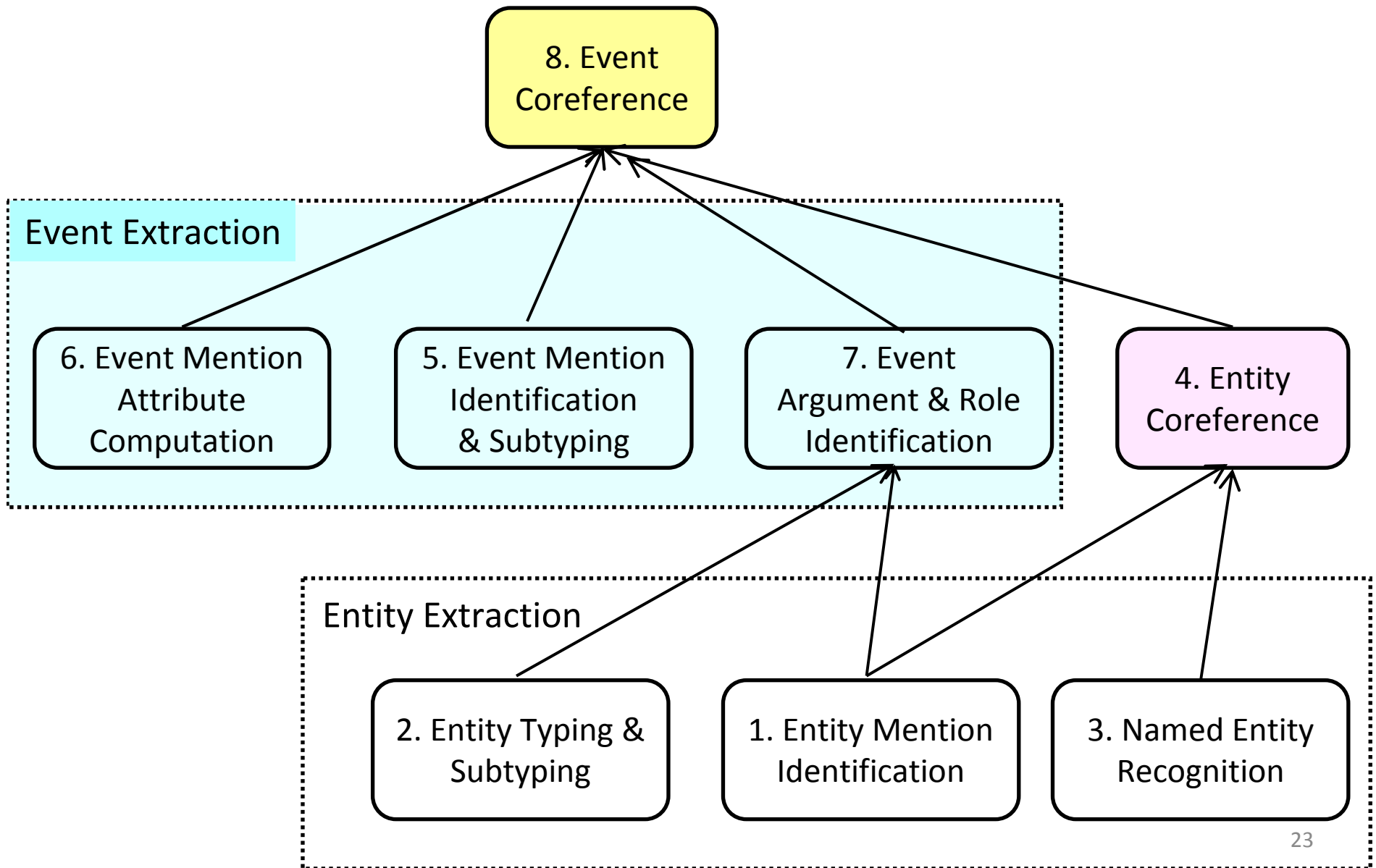
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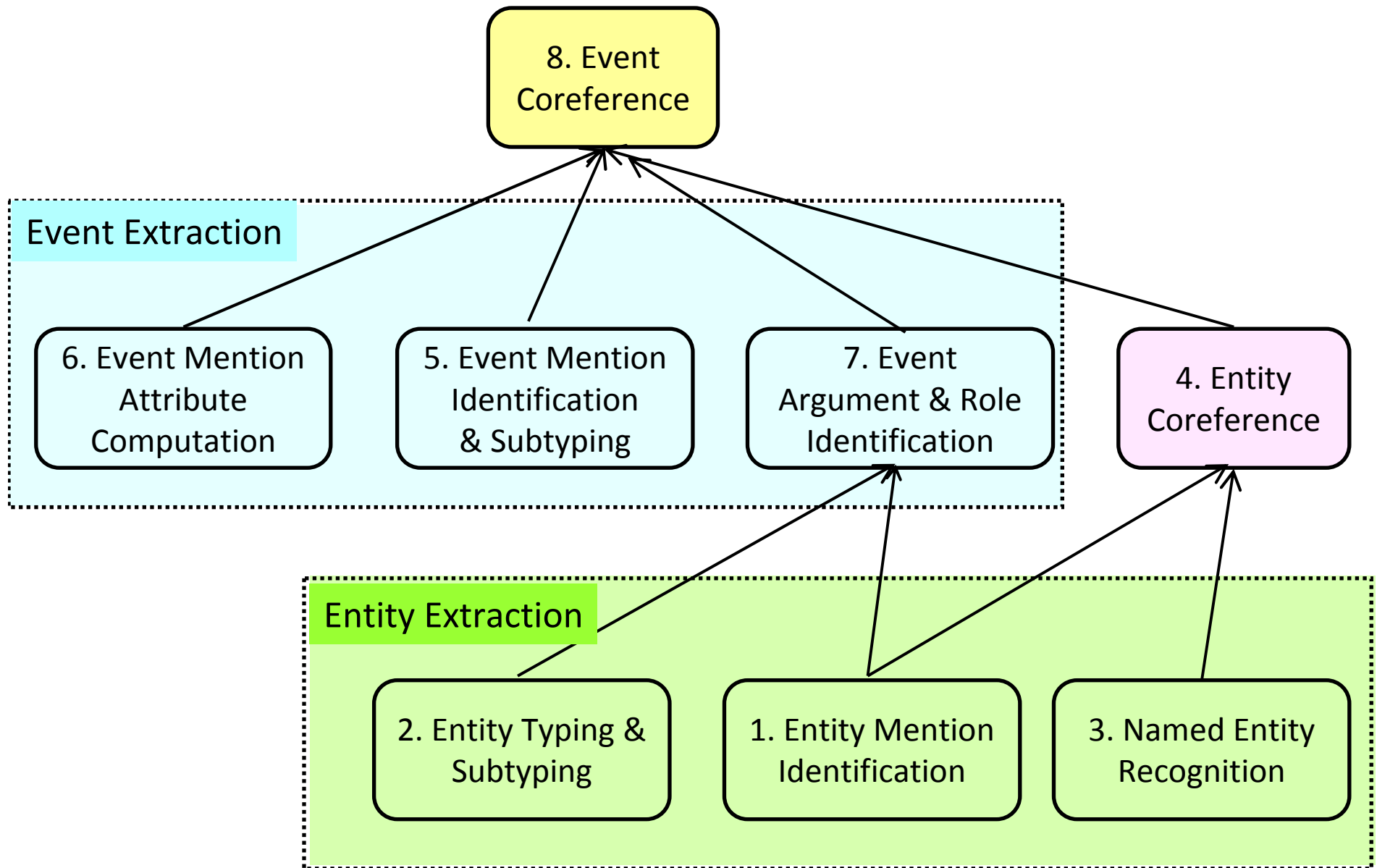
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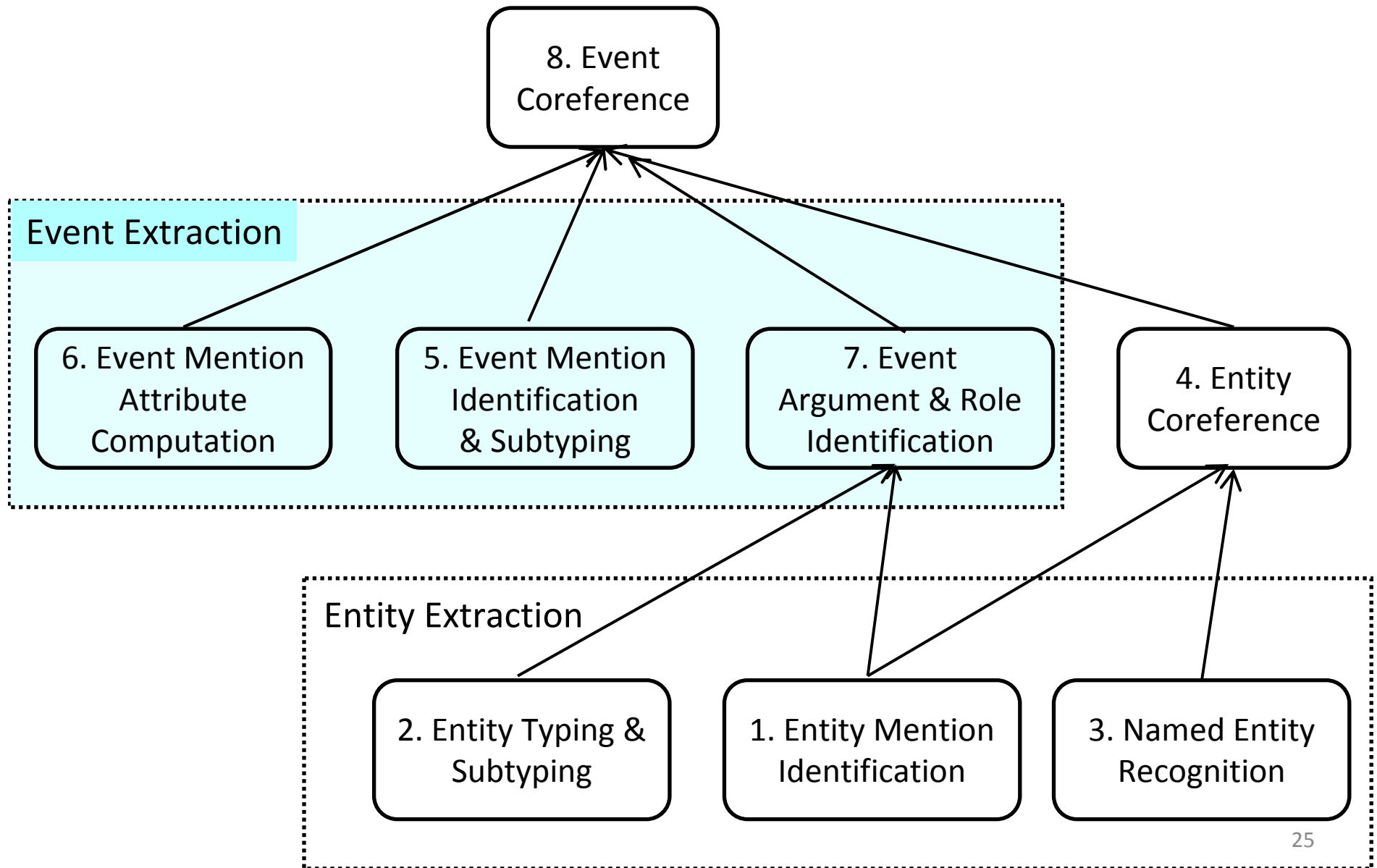
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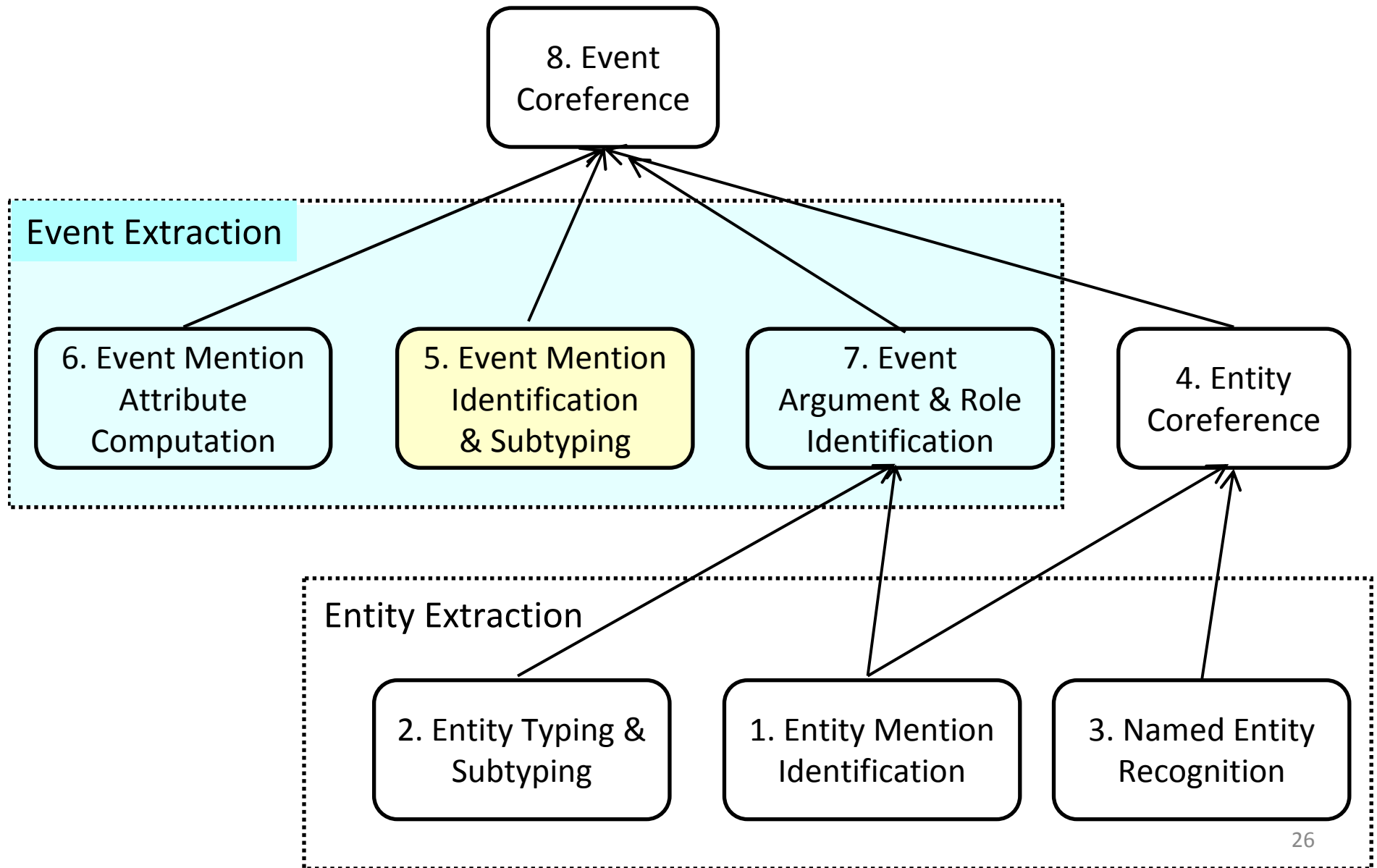
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Event Mention Identification & SubTyping

- Goals
 - Identify the event mentions
 - Label each event mention with its event subtype

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injured	INJURE
stabbed	ATTACK
criminal	ATTACK

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33 event subtypes
defined in ACE 2005

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Event Mention Identification & SubTyping

- Goals
 - Identify the event mentions
 - Label each event mention with its event subtype
- Why is this component useful for event coreference?
 - Provide the event mentions for event coreference
 - Subtyping information is useful for determining whether two event mentions are coreferent
 - » Two event mentions with different event subtypes cannot be coreferent

Event Mention Identification & SubTyping

- How to implement this component?
 - train a CRF (using CRF++) to jointly perform event mention identification and subtyping

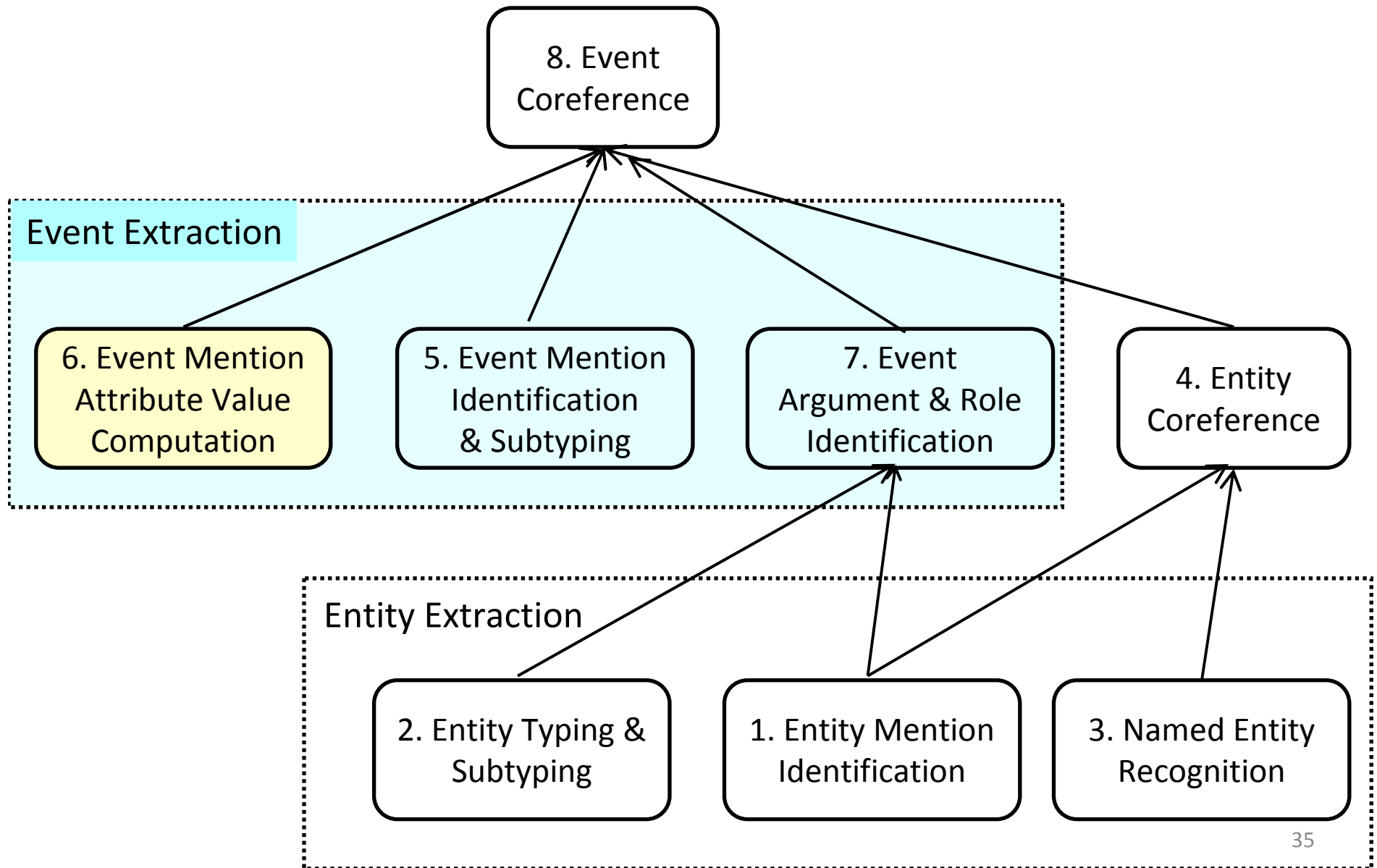
Event Mention Identification & SubTyping

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- **Results** (on ACE 2005)

Identification			SubTyping		
R	P	F	R	P	F
60.0	71.3	65.1	56.4	67.1	61.3

SinoCoreferencer: System Architecture



Event Mention

Attribute Value Computation

- Goal
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POLARITY, MODALITY, GENERICITY and TENSE

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Indicates whether the event happened or not

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- Why is this component useful for event coreference?
 - Two event mentions that differ in any of the four attributes cannot be coreferent

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- **Results** (in terms of F-score)
 - Perfect vs. predicted event mentions

	POLARITY	MODALITY	GENERICITY	TENSE
Input	F	F	F	F
Perfect	96.5	86.9	91.2	67.1
Predicted	62.9	56.2	59.4	36.7

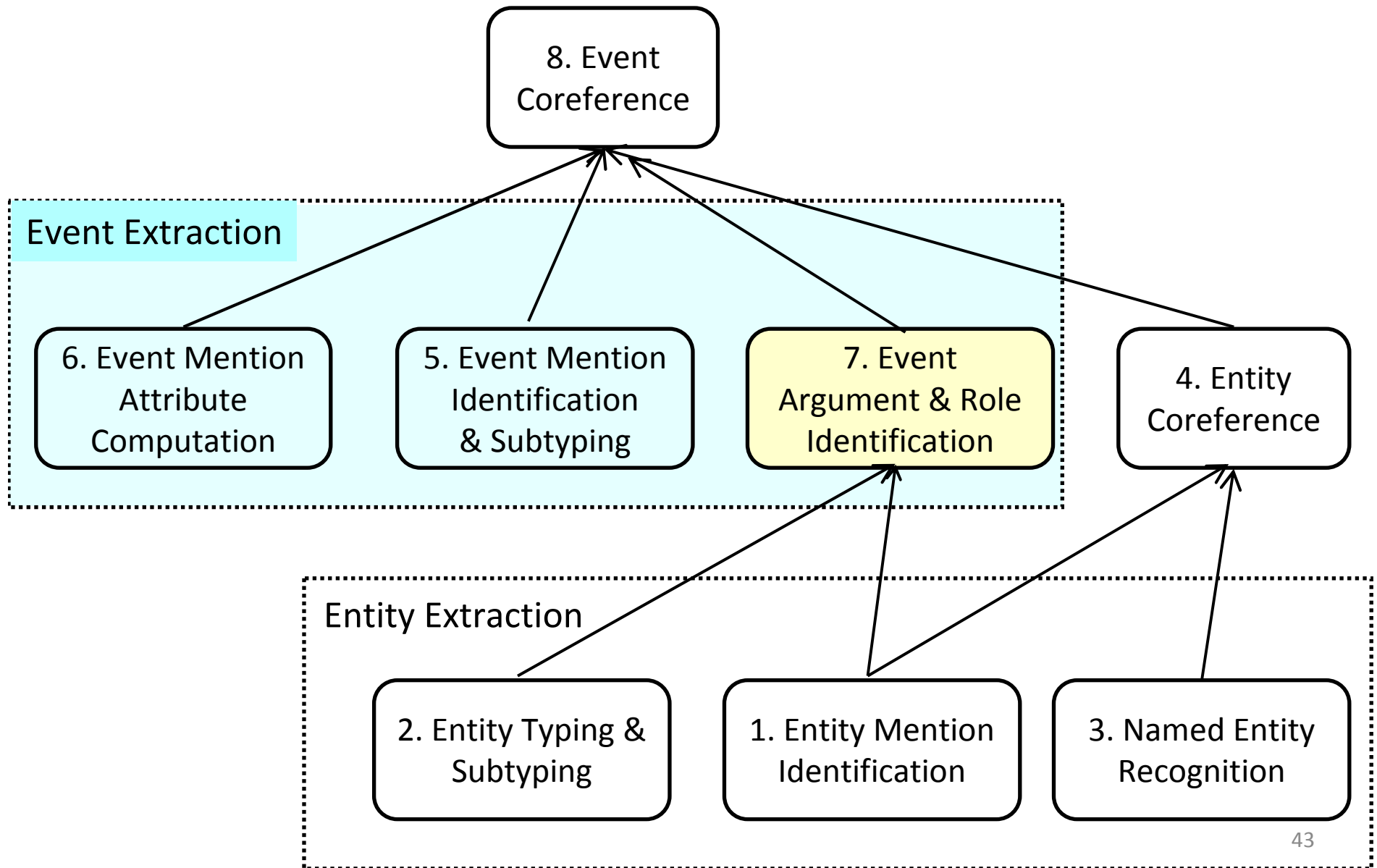
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- **Results** (in terms of R, P, F)
 - Perfect vs. predicted event mention boundary & subtype

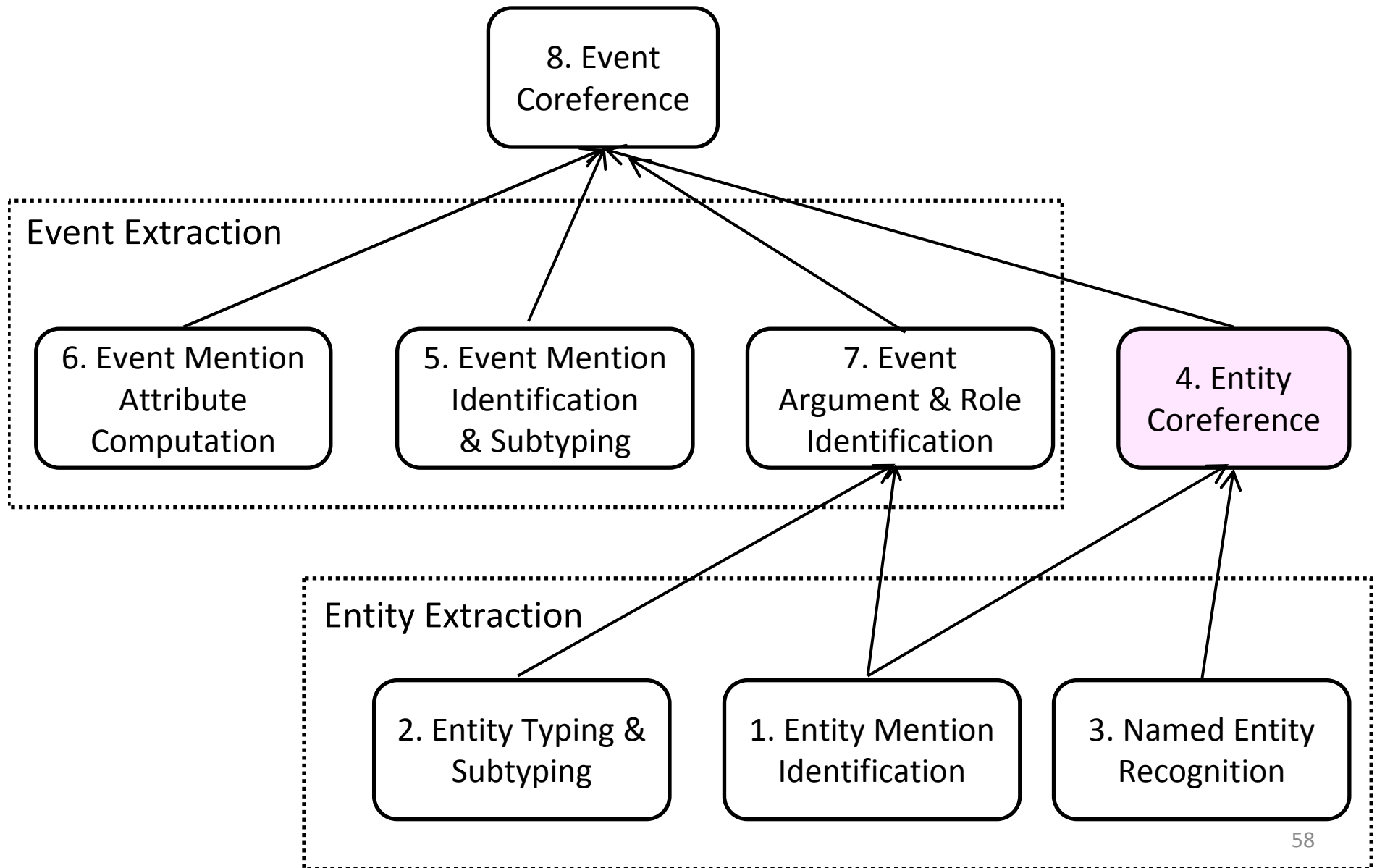
	Argument			Role		
Input	R	P	F	R	P	F
Perfect	68.9	87.1	76.9	61.1	77.2	68.2
Predicted	23.1	36.7	28.3	20.0	31.9	24.6

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SinoCoreferencer: System Architecture



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 - Two event mentions having coreferent arguments are likely to be coreferent

Entity Coreference Resolution

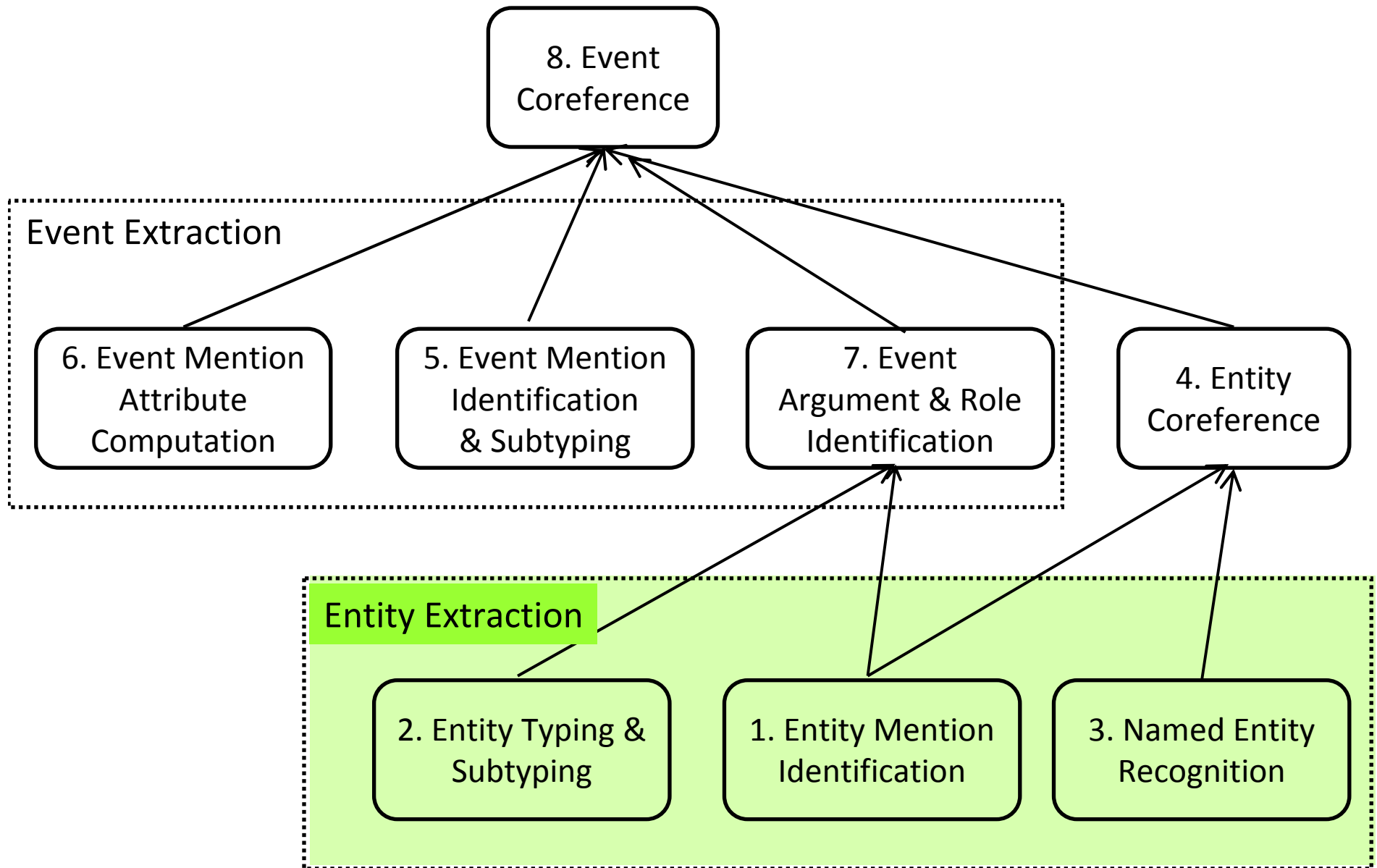
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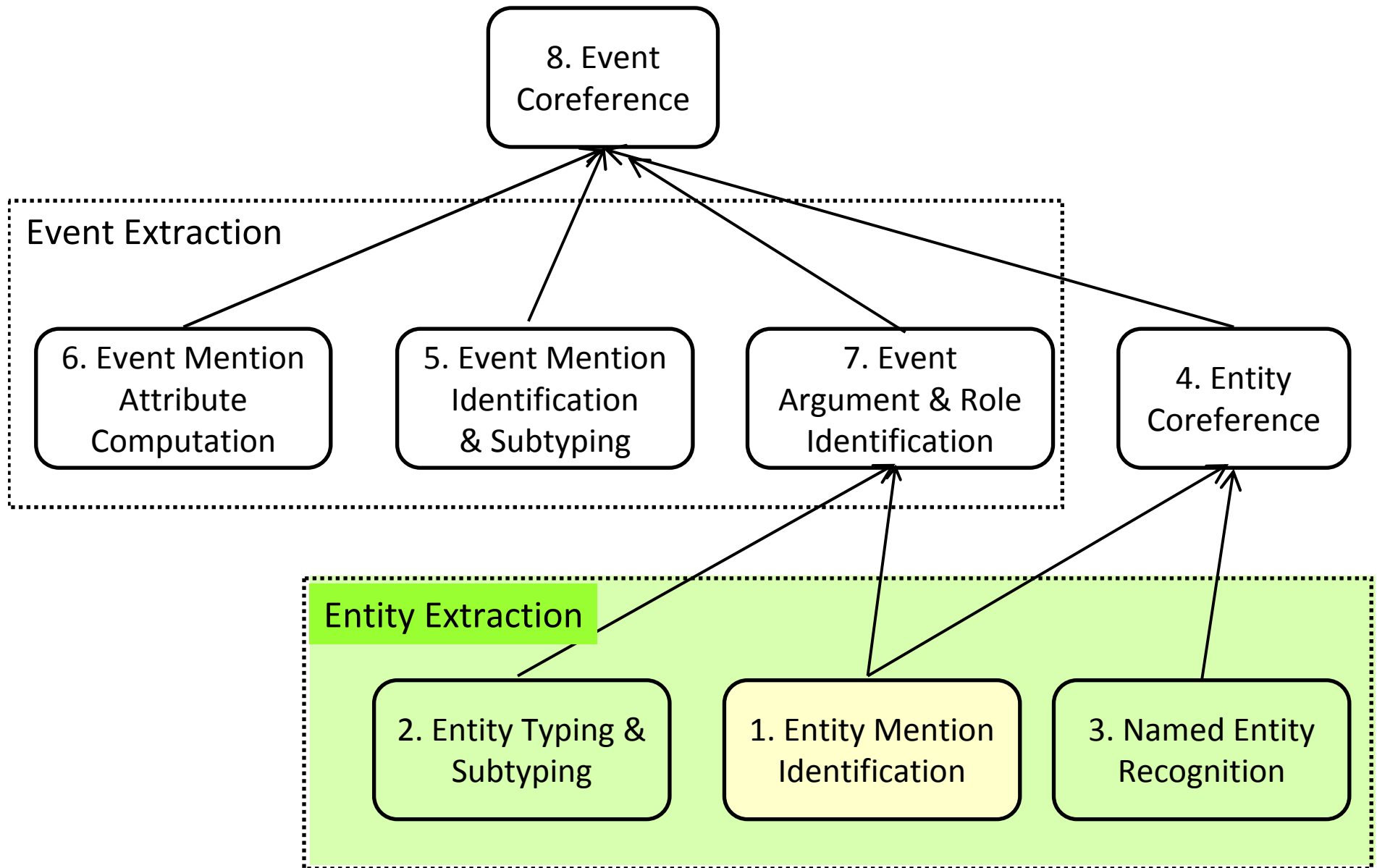
- How to implement this component?
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 - Hybrid rule-based and learning-based approach
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- **Results** (in terms of MUC, B³, and CEAF_e)
 - Perfect vs. predicted entity mentions

	MUC			B ³			CEAF _e			AvgF
Input	R	P	F	R	P	F	R	P	F	F
Perfect	71.5	85.8	78.0	67.4	88.0	76.4	69.4	48.8	57.3	70.6
Predicted	61.7	78.0	68.9	63.6	84.6	72.6	57.9	40.3	47.6	63.0

SinoCoreferencer: System Architecture



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Entity Mention Identification

- Goal
 - Provide the entity mentions needed by the downstream components
 - Candidate arguments of event mentions
 - Entity mentions needed for entity coreference

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 - Provide the entity mentions needed by the downstream components
 - Candidate arguments of event mentions
 - Entity mentions needed for entity coreference
 - Indirect influence on event coreference

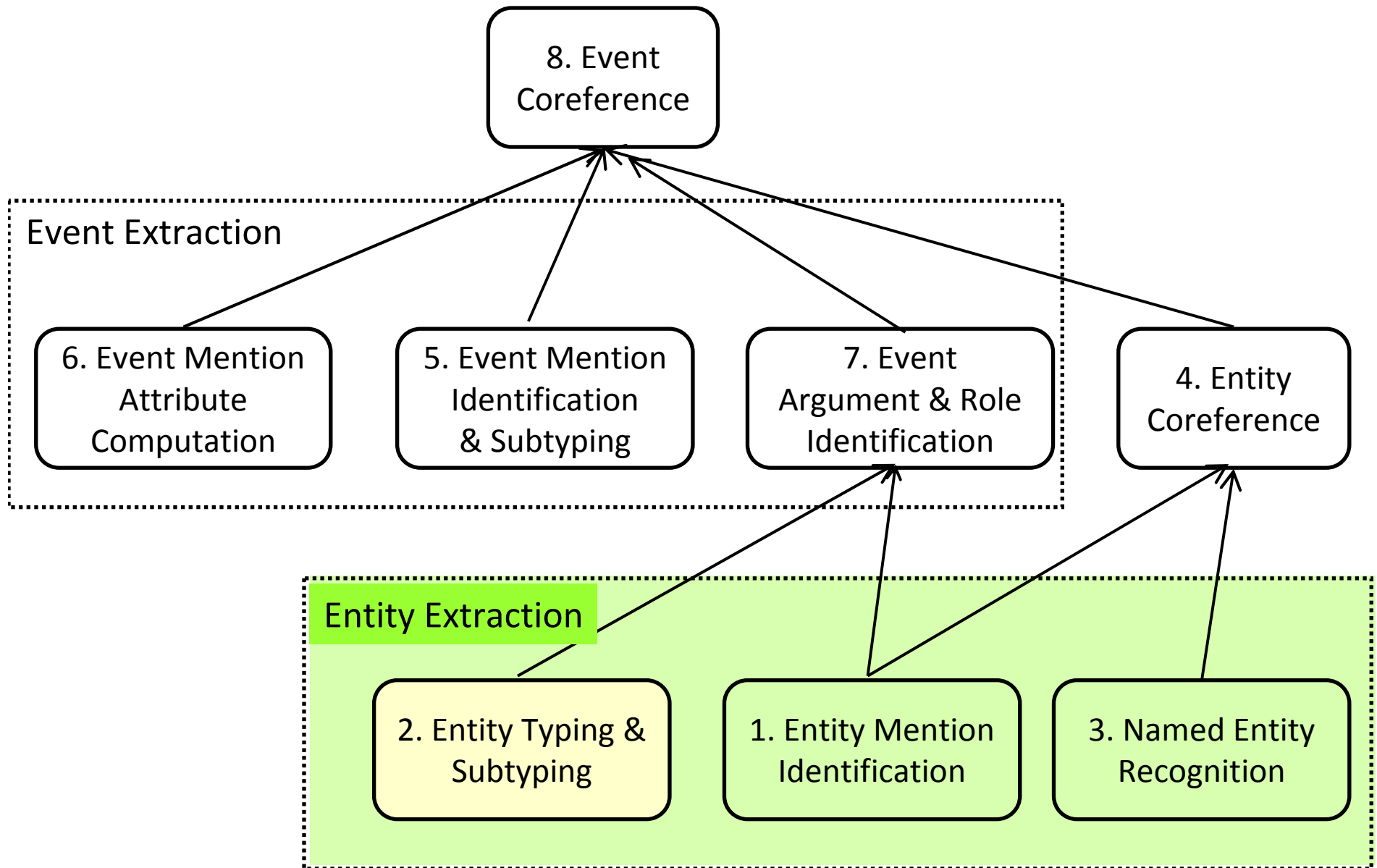
Entity Mention Identification

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 - We train CRF classifiers to extract entity mentions

Entity Mention Identification

- How to implement this component?
 - We train CRF classifiers to extract entity mentions
- achieves an overall F-score of 84.7% on ACE 2005

SinoCoreferencer: System Architecture



Entity Typing & SubTyping

- Goal
 - Determine the type and subtype of entity mention

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- Why is this component useful for event coreference?
 - Provide features needed to train a classifier for classifying the role of an event argument in event extraction
 - Indirect influence on event coreference

Entity Typing & SubTyping

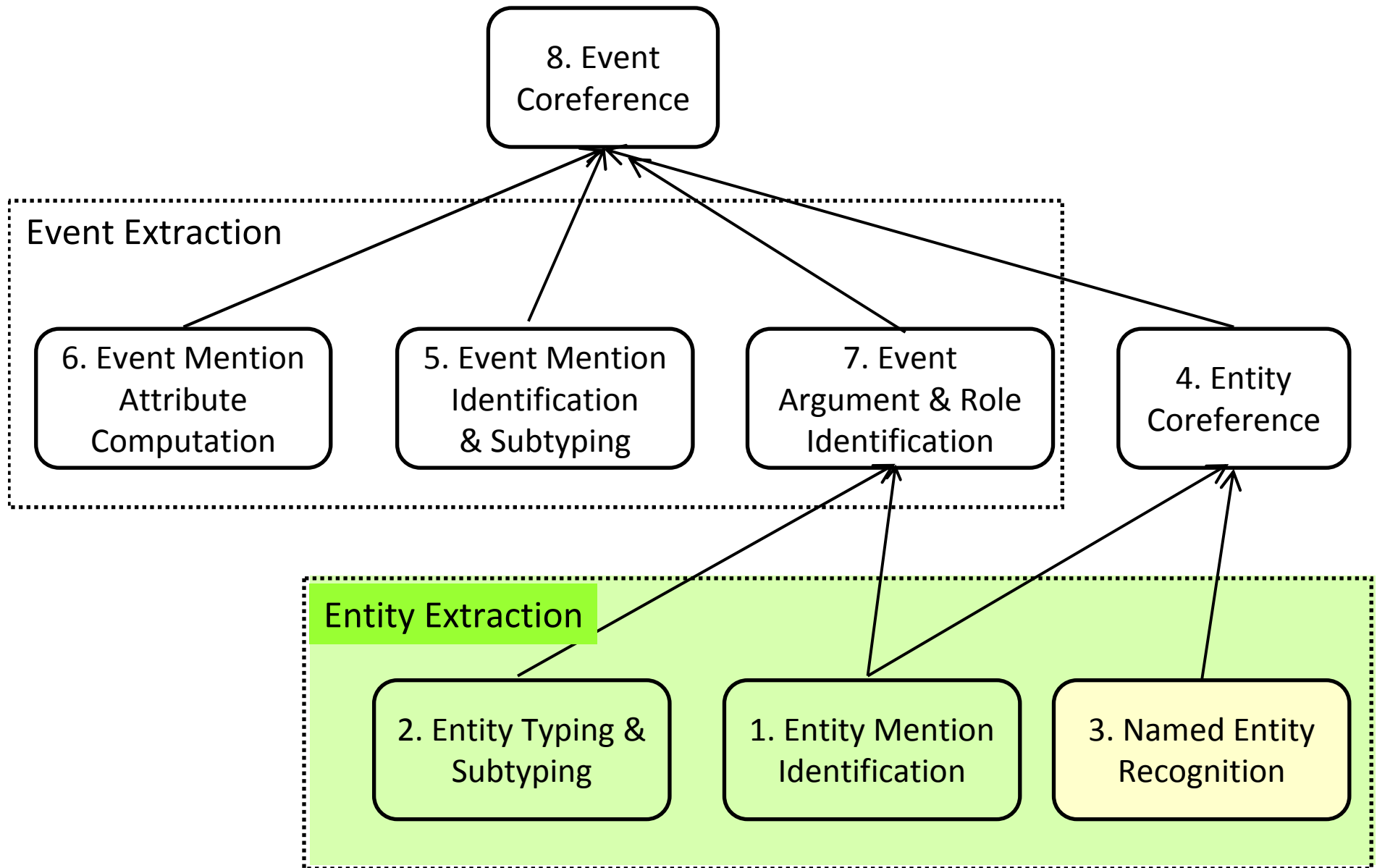
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Entity Typing & SubTyping

- How to implement this component?
 - train two SVM classifiers
 - One for classifying types and the other subtypes
- **Results** (in terms of R, P, and F)
 - Perfect vs. predicted entity mention boundaries

	Entity Typing			Entity SubTyping		
Input	R	P	F	R	P	F
Perfect	90.1	90.1	90.1	81.6	81.6	81.6
Predicted	80.5	77.6	79.0	73.1	70.4	71.7

SinoCoreferencer: System Architecture



Named Entity Recognition

- Goal
 - recognize named entities

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- Why is this component useful for event coreference?
 - Provide features for entity coreference resolution
 - Indirect influence on event coreference

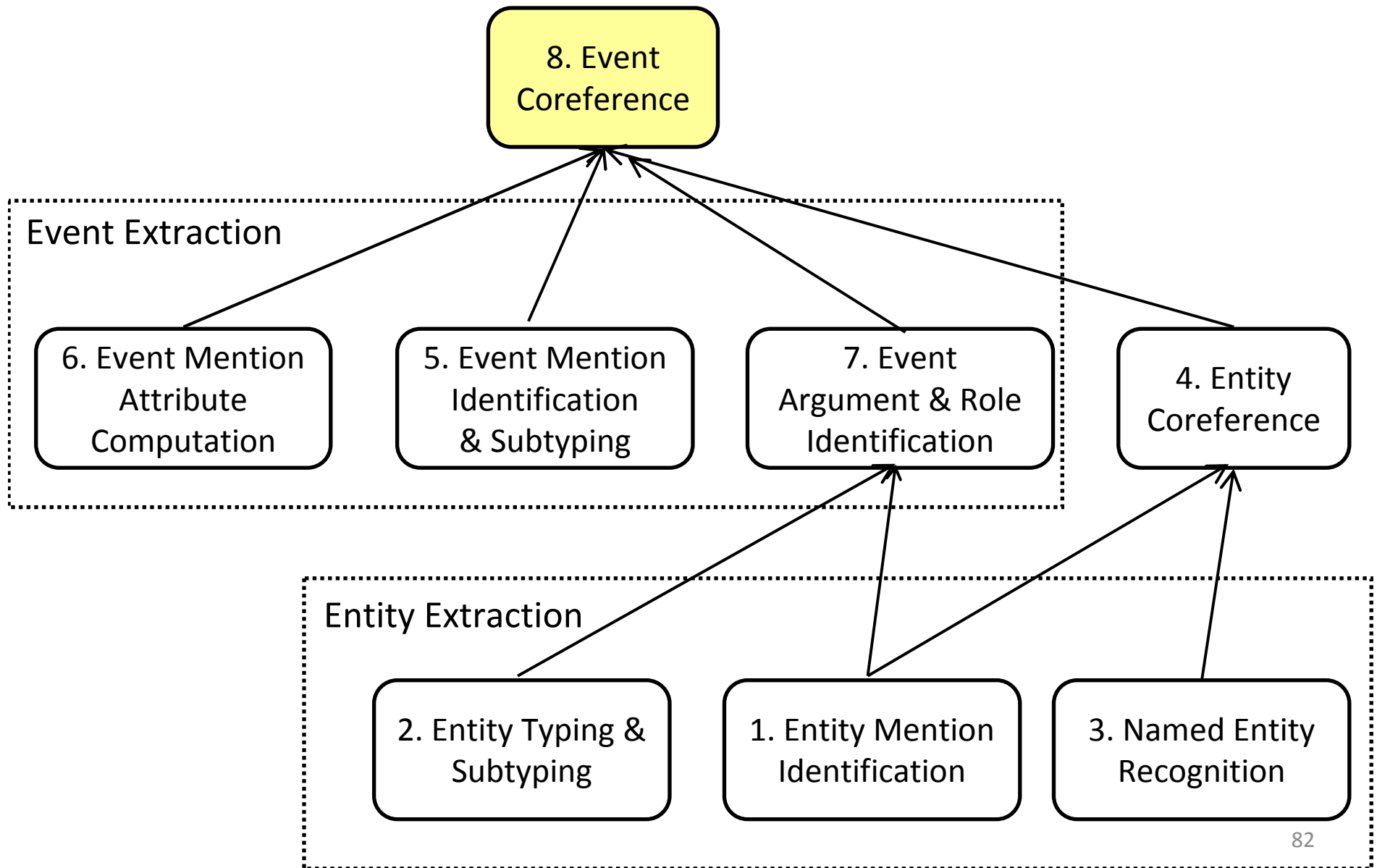
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- Overall F-score is 66.4%

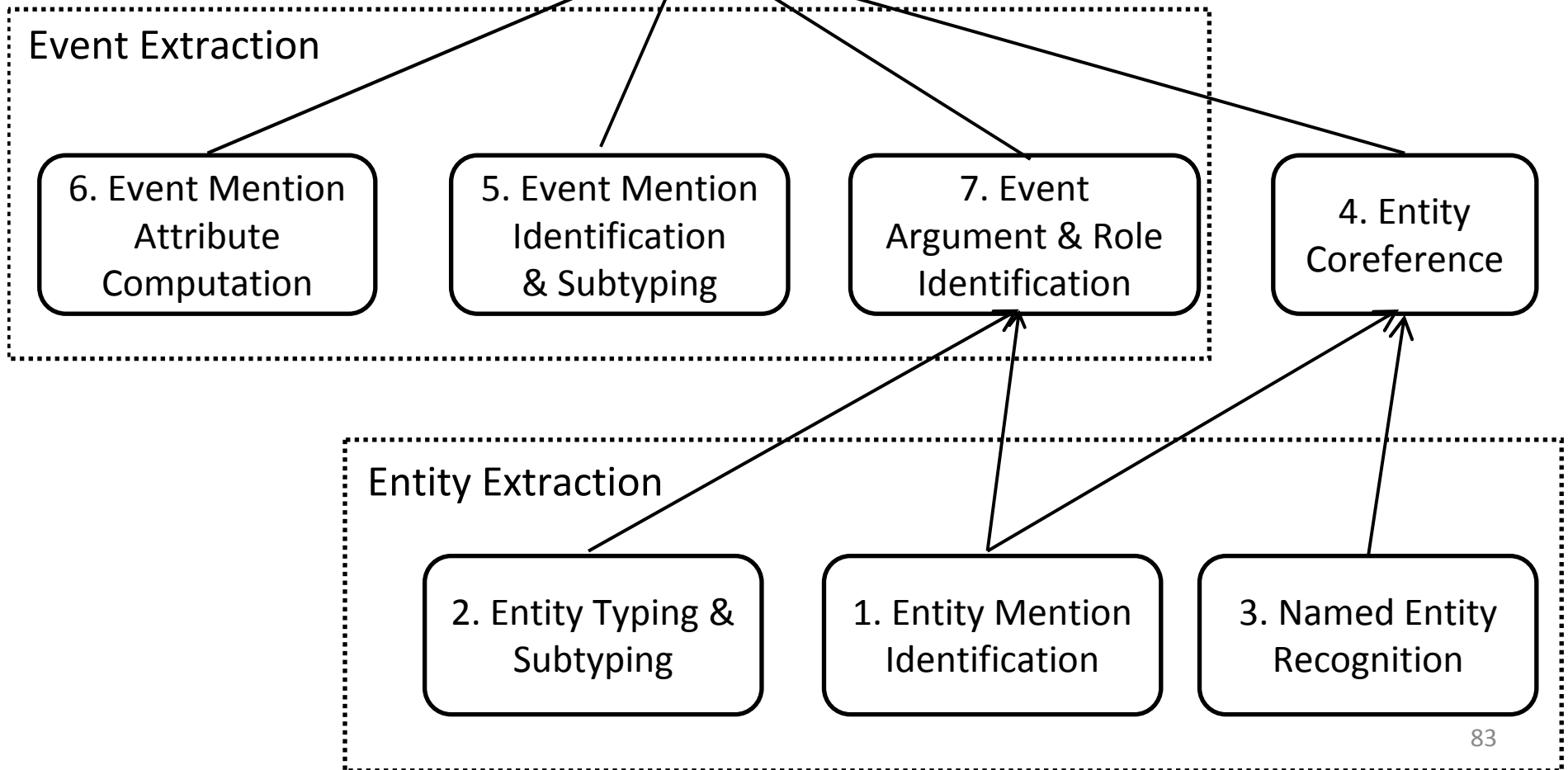
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8. Event Coreference

Takes as input the output of event extraction and entity coreference



Event Coreference

- Encode the output of the entity coreference subsystem and the event extraction subsystem as **features** for training an event coreference classifier
 - Pairwise classifier that determines whether two event mentions are coreferent

Event Coreference

- Encode the output of the entity coreference subsystem and the event extraction subsystem as **features** for training an event coreference classifier
 - Pairwise classifier that determines whether two event mentions are coreferent
- During testing, it selects as the antecedent of each event mention the closest preceding event mention that is classified as coreferent with it

Results of Event Coreference

		MUC				B ³				CEAF _e			AvgF
Input	R	P	F	R	P	F	R	P	F	R	P	F	F
Perfect	80.4	70.0	74.8	88.4	79.7	83.8	57.3	66.8	61.7	73.4			
Predicted	37.4	36.7	37.1	72.8	71.1	71.9	40.6	41.1	40.8	49.9			

Evaluation

- To what extent is the noisy output of each of its upstream components limiting the performance?

Ablation Experiments

- Start with an event coreference resolver that assumes all upstream components are error free
- **Replace** each oracle component with its predicted (i.e., automatically computed) counterpart one by one

Summary of Ablation Results

- Components whose noise has big impact on event coreference performance:
 - those in event extraction and entity coreference
- Components whose noise has little impact:
 - those in entity extraction

Same as English Event Coreference?

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- Chinese event coreference is more challenging in part because the upstream components may have lower accuracies
 - Chinese has **no morphology**, so in Chinese it is hard to correctly classify verb tenses
 - Chinese requires **segmentation**. Segmentation errors affect event mention detection
 - Chinese has **zero pronouns**, which make it harder to find the arguments of events

Summary

- Analyzed an ACE-style Chinese event coreference resolver
- Made our implementation publicly available
 - Facilitate the development of Chinese NLP applications

Future Work

- Examine **partial** coreference relations (Hovy et al., 2013)
 - **subevent**
 - Subevent relations form a stereotypical sequence of events
 - e.g., bombing → destroyed → wounding
 - **membership**
 - multiple instances of the same event
 - e.g., injury events corresponding to different people

NAACL HLT 2013 and ACL 2014 workshops