

# Anaphora Resolution in Biomedical Literature: A Hybrid Approach

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# What is Anaphora Resolution?

{FK506} suppressed {the transcriptions through the AP-1 or kappa B-like sites} induced by {PMA} plus {Ca(2+)-mobilizing agents,} but not those induced by Ca(2+)-independent stimuli.

- Task: identify an antecedent for each anaphor
- 3 subtasks
  1. Identify all the anaphors
  2. Identify all the candidate antecedents for each anaphor
  3. Determine which of these candidate antecedents is the correct antecedent for each anaphor

# Our Evaluation Data-set

- from BioNLP 2011 Coreference Task

# Why Coreference?

Useful for Event Extraction

# BioNLP Event Extraction

## Event Cause

A mutant of KBF1/p50(delta SP), unable to bind to DNA but able to form homo-  
or heterodimers, has been constructed. This protein **Negative Regulation Event**  
**(reduces or abolishes)** in vitro  
the DNA binding activity of wild-type proteins of the same family...

# Previous Approaches to Coreference

- Rule-Based or Learning-Based

## Our Approach: Hybrid Approach

- Use different approaches to resolve different classes of anaphors.

# Different classes of anaphors?

Anaphor Type	Examples	Training	Development
Relative Pronoun	that, which, who, where, etc.	54.3%	56.9%
Personal Pronoun	it, they	26.6%	26.0%
Definite Noun Phrase	the genes, this protein, etc.	15.4%	14.0%
Demonstrative & Indefinite Pronoun	this, those, both, etc.	2.4%	2.1%
Others		1.3%	1.1%

- Why no statistics on the test set?
- How then do we evaluate?  
The test set is not available to system developers.

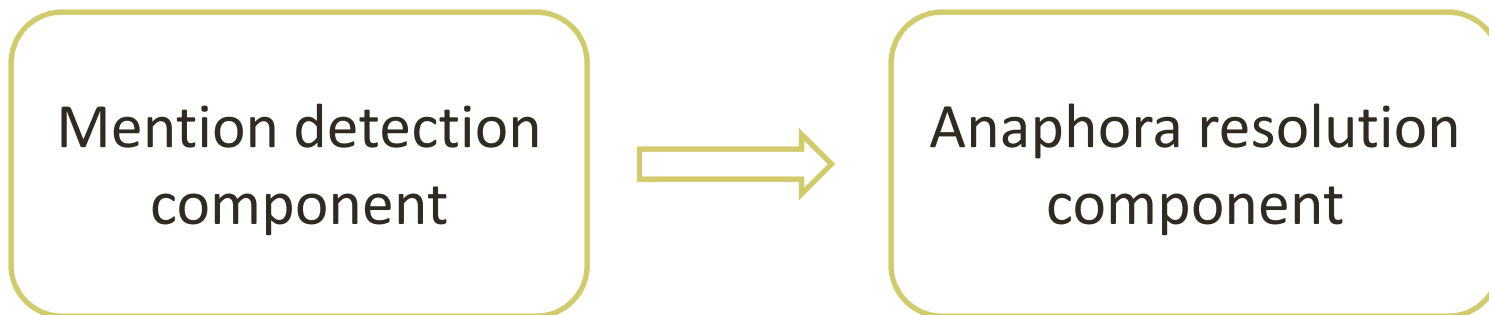
# Motivation for Hybrid System

- Hypothesis: Different classes of anaphors might be better resolved using different approaches.
- Basis of Hypothesis?
  - Linguistic properties
    - Different features for different anaphor types?
  - Data-set distributions
    - Rule-based versus learning-based approaches?



# System Architecture

- A pipeline architecture



FK506 suppressed the transcriptions through the AP-1 or kappa B-like sites induced by PMA plus Ca(2+)-mobilizing agents, but not those induced by Ca(2+)-independent stimuli.

Mention detection component

{ FK506 } suppressed { the transcriptions through the AP-1 or kappa B-like sites } induced by { PMA } plus { Ca(2+)-mobilizing agents }, but not those induced by Ca(2+)-independent stimuli.

**Candidates**

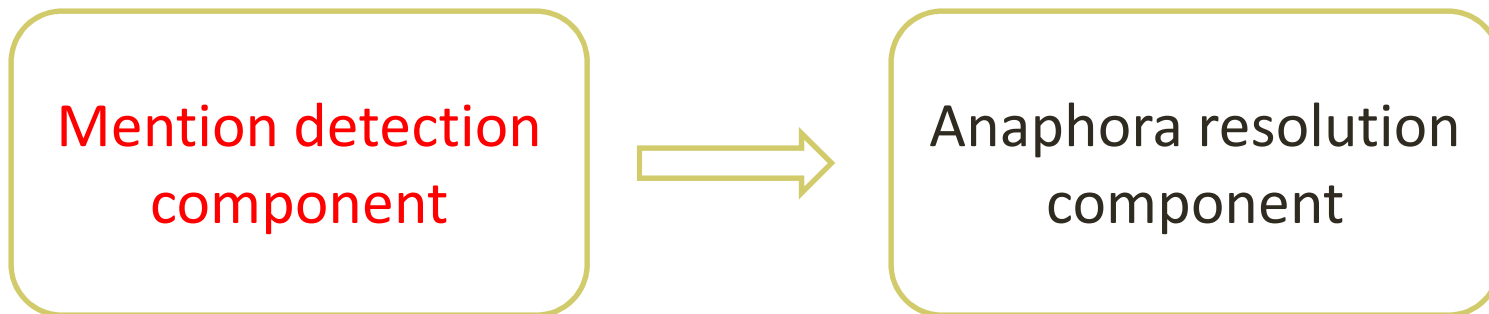
**Anaphor**

Anaphora resolution component

FK506 suppressed { the transcriptions through the AP-1 or kappa B-like sites } induced by PMA plus Ca(2+)-mobilizing agents, but not those induced by Ca(2+)-independent stimuli.

# System Architecture

- A pipeline architecture



Goal: Extract Anaphors &  
Candidate Antecedents

## 2 Approaches to Mention Detection

1. Learning-Based Approach
2. Heuristic-Based Approach

# Learning-Based Mention Detection

- Sequential Labeling Task – CRF
- Class Values: given a sentence token, does it begin the mention (B), or is it inside the mention (I), or is it outside a mention (O)?
- Features: Token, POS, word shape information, etc.
- Separate Anaphor & Candidate Antecedent Classifiers [Kim et al., 2011]
- Limitation:
  - Insufficient training instances for sparse anaphor classes

# Heuristic-Based Mention Detection

- **Anaphor Extractor**

- Step1: List-Based Extraction

- Use pre-created lists to extract anaphors

- Step 2: Prune Extracted Non-Anaphors with Heuristics

- E.g.s. of non-anaphors are complementizers as in “found that”, “suggests that”, or pleonastic pronouns as in “It is found that”, “It was possible that”, etc.

- **Antecedent**

- List synt

Anaphor type	Before Pruning	After Pruning
	TP/FP	TP/FP
Relative pronoun	269/313	262/22
Personal pronoun	123/235	120/5
D&I pronoun	32/19	32/13
Definite NP	10/12	10/2

from the mention)

# Combinations of Mention Extraction Methods

- We now have 2 methods for extracting candidate antecedents (1 learning-based, 1 heuristic-based)
- We now have 2 methods for extracting anaphors (1 learning-based, 1 heuristic-based)
- We can mix learning-based and heuristic-based methods for extracting anaphors and candidate antecedents
  - 4 possible ways:
    - CRF Anaphors + CRF Antecedents
    - CRF Anaphors + Heuristic Antecedents
    - Heuristic Anaphors + Heuristic Antecedents
    - Heuristic Anaphors + CRF Antecedents

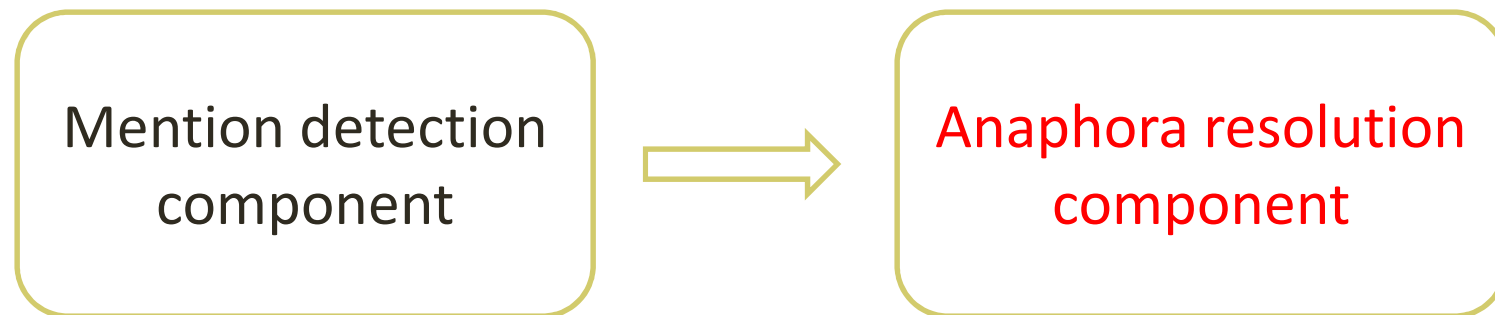
# Which combination should we use?

- Development data helps us decide...



# System Architecture

- A pipeline architecture



Goal: To find the antecedent for an anaphor

# 6 Anaphora Resolution Methods

1. Reconcile Features
2. Sentence-Based Flat Parse Features
3. Document-Based Flat Parse Features
4. Sentence-Based Structured Parse Feature
5. Document-Based Structured Parse Feature
6. Rule-Based Method

**Learning  
-Based  
Methods**

- Why 6 methods?
  - Hypothesis: Different methods may work well for different anaphor types

# Resolution Method 1

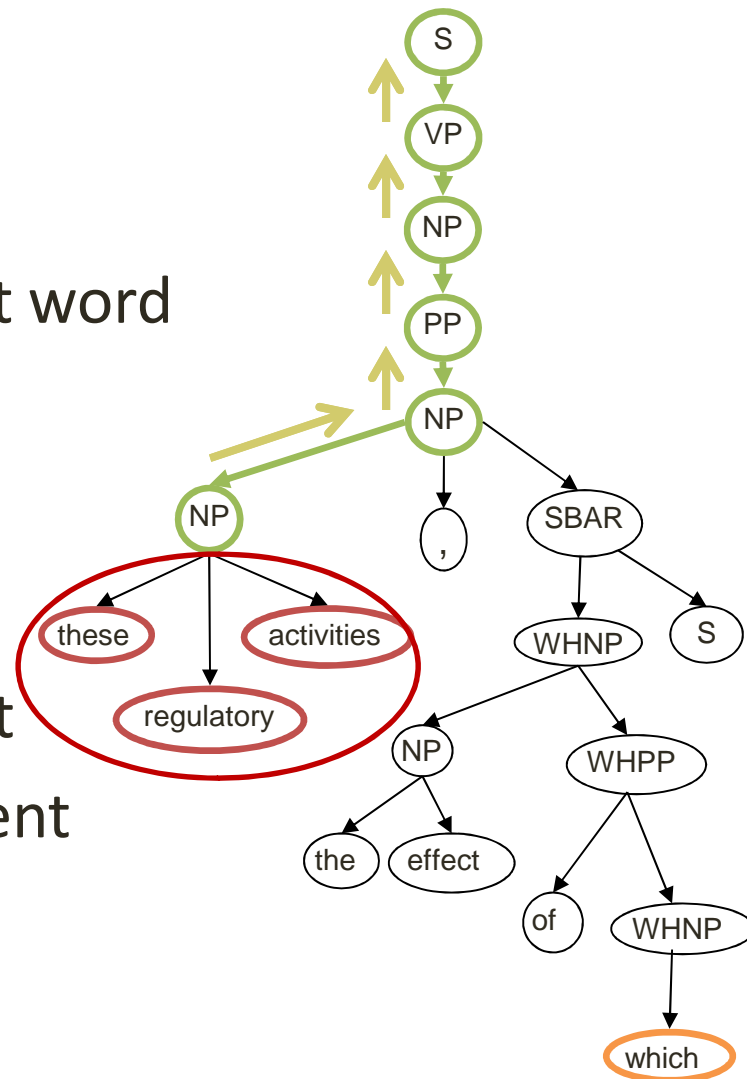
- Goal
  - using a ranker trained on Reconcile features to obtain the correct antecedent for an anaphor
    - 66 string-matching, grammatical, positional, and semantic features from Reconcile
    - ranker aims to rank the candidate so the correct one has highest rank
- How do we train this ranker?
  - generate a feature vector for anaphor paired with a candidate from the list

# Resolution Method 2

- Weakness of Method 1
  - need to design potentially complex heuristics for encoding parse tree information as features
- Solution
  - train a ranker on **path-based features** extracted from sentence parse trees (i.e. features derived from paths in a parse tree)
- 6 path-based features

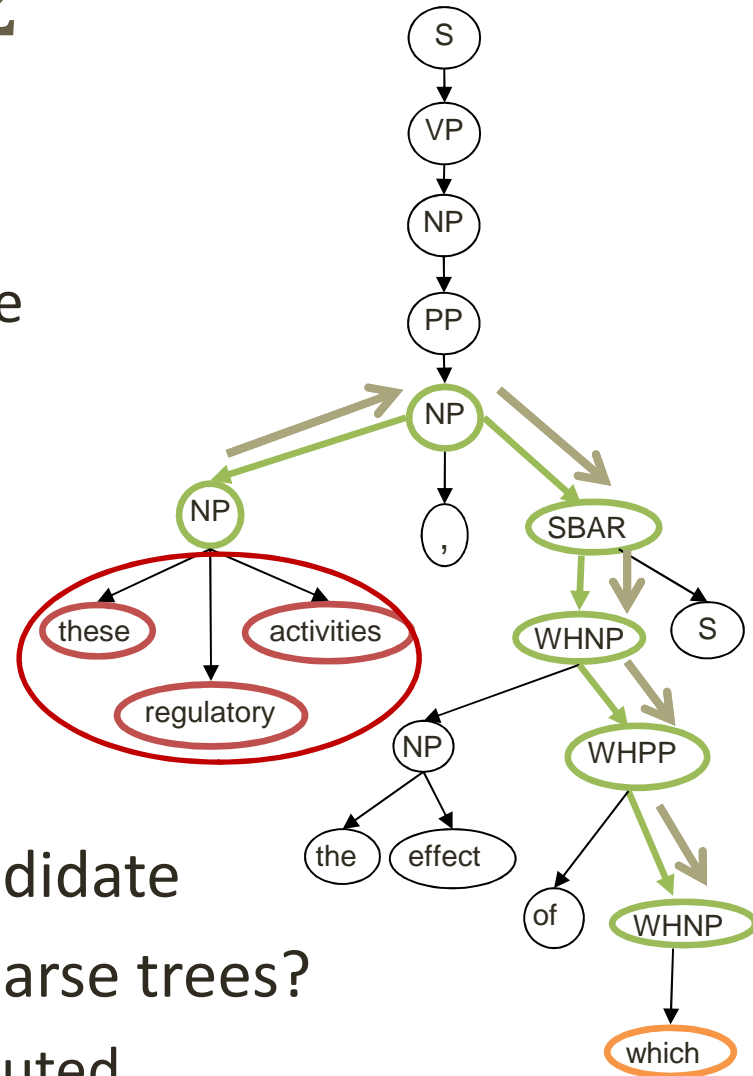
# Resolution Method 2

- Feature 1
  - Path from the parent of first candidate antecedent word to the root of the tree
- Motivation
  - Captures syntactic context of the candidate antecedent



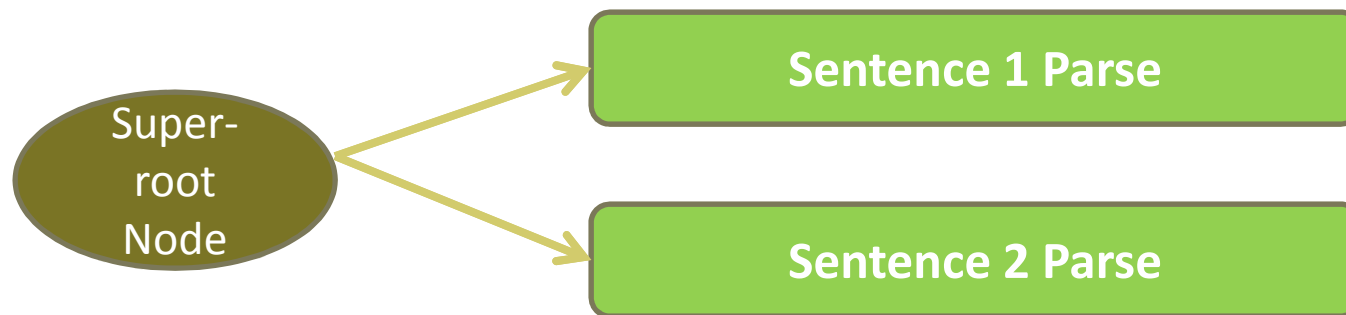
# Resolution Method 2

- Feature 6
  - Directed path from candidate antecedent to anaphor
- Motivation
  - Captures syntactic context
- What if the anaphor and candidate antecedent are in different parse trees?
  - This feature cannot be computed



# Resolution Method 3

- Addresses this problem by using document based rather than sentence based parse trees
- What are document based parse trees?
  - sentence parses are connected by a pseudo link



- Ranker trained on the same 6 features as in method 2 except that they are computed on document parse trees

# Resolution Method 4

- Weakness of methods 2 & 3
  - Need to manually determine which paths in a parse tree to use as features
- Solution
  - Use a sentence-based parse tree as a **structured** feature
- What is a structured feature?
  - A feature whose value is a linear or hierarchical structure, as opposed to a flat feature, which has a discrete or real value



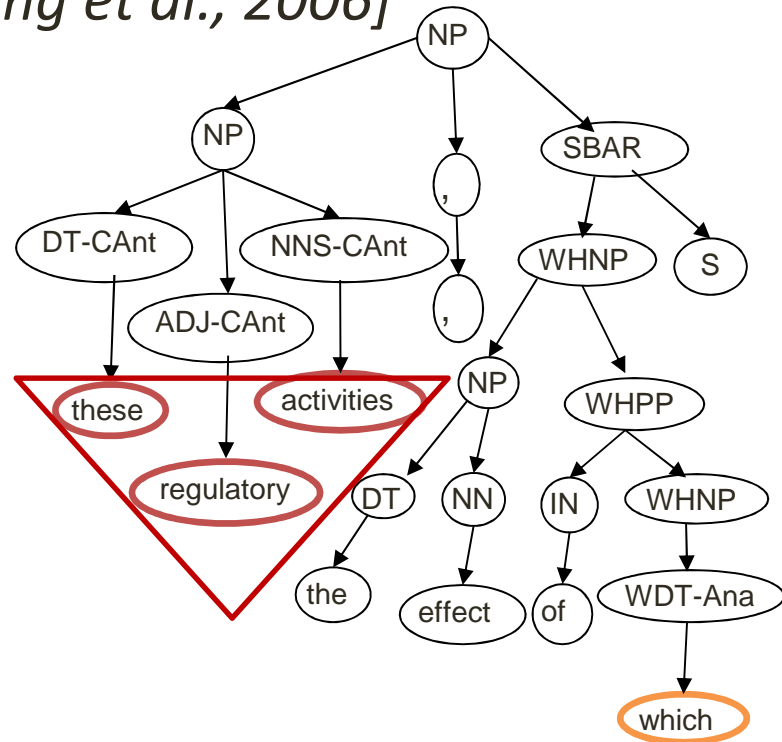
# Resolution Method 4

- But we cannot use the entire parse tree...
  - the learner cannot generalize well
  - so we extract a **parse substructure** (i.e. subtree) and use as a structured feature
  - But which parse substructure do we extract?

# Structured Tree Feature

- Simple Expansion Tree [Yang et al., 2006]

- includes all nodes in path from candidate antecedent to anaphor and the nodes first level children

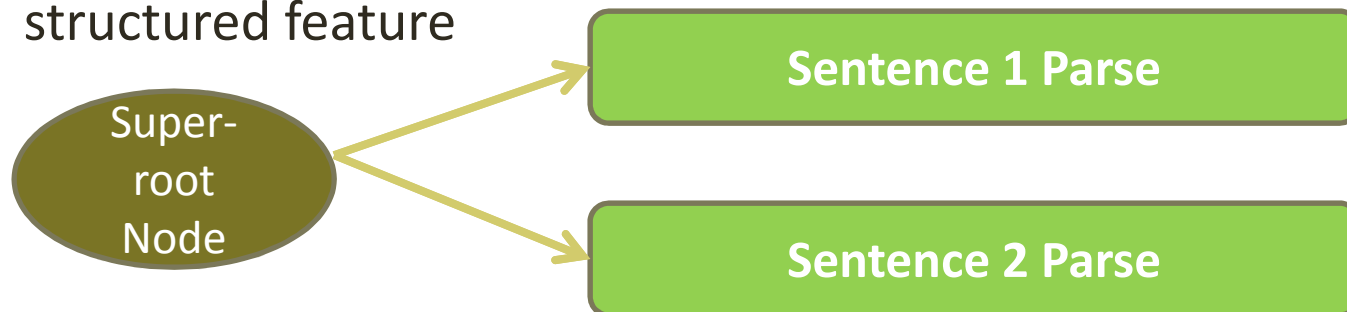


# Resolution Method 4

- Use this sentence-based structured feature to train a classifier

# Resolution Method 5

- Weakness of method 4
  - The sentence-based structured feature cannot be computed if the candidate antecedent and the anaphor are not in the same sentence
- Solution
  - Same as method 4 except that we connect sentence-based parse trees by a pseudo link to create a document-based structured feature



# Resolution Method 6

- **Rule-based method**
- Each rule specifies which candidate antecedent an anaphor should be resolved to.
- Each type of anaphors has its own set of resolution rules.
  - Each set of resolution rules is **ordered**
    - So that the second rule is applied only if the first rule is not applicable

# Rules for Resolving Personal Pronouns

Rule 1: Resolve anaphor to candidate if (1) the two agree in number and are in the same sentence; and (2) candidate contains a protein name or one of its words satisfies the three conditions in the Pattern rule.

Rule 2: Resolve anaphor to candidate if the two agree in number and are in the same sentence.

Rule 3: Resolve anaphor to candidate if candidate contains a protein name or one of its words satisfies the three conditions in the Pattern rule.

Rule 4: Resolve anaphor to candidate if the two are in the same sentence.

Rule 5: Resolve anaphor to candidate if the two agree in number.

# Rule for Resolving Relative Pronouns

Resolve anaphor to the closest candidate.

- For each type of anaphors, we have 24 method combinations, because we have:
  - 2 candidate antecedent extraction methods
  - 2 anaphor extraction methods
  - 6 resolution methods
- Which combination should we use?
  - We use the **development set** to determine the best combination of anaphor extraction method, antecedent extraction method, and resolution method for **each of the 4 types of anaphors**.



## Relative Pronoun Resolution Results on Development Set

Resolution Method	CRF anaphors						Heuristic anaphors					
	CRF candidates			Heuristic candidates			CRF candidates			Heuristic candidates		
	R	P	F	R	P	F	R	P	F	R	P	F
Ranking-based Reconcile	21.3	60.6	31.5	13.4	47.4	20.8	21.3	62.3	31.7	14.9	53.6	23.3
Sentence-based flat	19.8	83.3	32.0	28.2	83.8	42.2	18.8	84.4	30.8	25.2	91.1	39.5
Document-based flat	19.3	83.0	31.3	28.2	78.0	41.4	19.3	84.8	31.5	24.3	90.7	38.3
Sentence-based structured	21.3	75.4	33.2	22.8	79.3	35.4	20.8	77.8	32.8	22.3	78.9	34.7
Document-based structured	21.3	69.4	32.6	22.3	77.6	34.6	20.8	72.4	32.3	22.3	81.8	35.0
Rule-based	—	—	—	27.2	75.3	40.0	—	—	—	27.7	77.8	40.8

- Best combination for relative pronouns:
  - CRF anaphors, heuristic candidates and learning method using sentence-based flat features.

## Personal Pronoun Resolution Results on Development Set

Resolution Method	CRF anaphors						Heuristic anaphors					
	CRF candidates			Heuristic candidates			CRF candidates			Heuristic candidates		
	R	P	F	R	P	F	R	P	F	R	P	F
Ranking-based Reconcile	3.5	24.1	6.1	19.3	63.9	29.7	5.0	40.0	8.8	19.8	59.7	29.7
Sentence-based flat	3.5	53.8	6.5	21.8	74.6	33.7	3.5	63.6	6.6	21.3	76.8	33.3
Document-based flat	3.0	54.5	5.6	19.8	80.0	31.7	3.5	63.6	6.6	19.8	81.6	31.9
<b>Sentence-based structured</b>	3.5	53.8	6.5	24.3	73.1	36.4	5.0	66.7	9.2	<b>26.3</b>	<b>77.9</b>	<b>39.3</b>
Document-based structured	3.5	26.9	6.1	21.8	75.9	33.8	5.0	34.5	8.7	23.8	76.2	36.2
Rule-based	—	—	—	13.9	75.7	23.4	—	—	—	16.3	71.7	26.6

- Best combination for personal pronouns:
  - Heuristic anaphors, heuristic candidates and learning method using sentence-based structured feature.

## Demonstrative & Indefinite Pronoun Resolution Results on Development Set

Resolution Method	CRF anaphors						Heuristic anaphors					
	CRF candidates			Heuristic candidates			CRF candidates			Heuristic candidates		
	R	P	F	R	P	F	R	P	F	R	P	F
Ranking-based Reconcile	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN
<b>Sentence-based flat</b>	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	<b>2.0</b>	<b>12.9</b>	<b>3.4</b>
Document-based flat	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	0.0	0.0	NaN
Sentence-based structured	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	0.0	0.0	NaN
Document-based structured	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN
Rule-based	—	—	—	0.0	NaN	NaN	—	—	—	1.0	100	2.0

- Best combination for demonstrative and indefinite pronouns:
  - Heuristic anaphors, heuristic candidates and learning method using sentence-based flat features.

## Definite Noun Phrase Resolution Results on Development Set

Resolution Method	CRF anaphors						Heuristic anaphors					
	CRF candidates			Heuristic candidates			CRF candidates			Heuristic candidates		
	R	P	F	R	P	F	R	P	F	R	P	F
Ranking-based Reconcile	0.0	NaN	NaN	0.5	100	1.0	0.5	11.1	0.9	1.0	50.0	1.9
Sentence-based flat	0.0	NaN	NaN	0.5	7.1	0.9	0.0	NaN	NaN	2.5	14.7	4.2
Document-based flat	0.0	NaN	NaN	1.0	12.5	1.8	0.0	NaN	NaN	0.0	0.0	NaN
Sentence-based structured	0.0	NaN	NaN	0.0	0.0	NaN	0.0	NaN	NaN	0.0	NaN	NaN
Document-based structured	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN	0.0	NaN	NaN
Rule-based	—	—	—	5.0	38.5	8.8	—	—	—	6.9	58.3	12.4

- Best combination for definite noun phrases:
  - Heuristic anaphors, heuristic candidates and rule-based method.

# Observation

- Different combination methods work best for different types of anaphors on development set
  - Provides empirical support for a hybrid approach to anaphora resolution
- We employ the best combination learned for each anaphor type from the development set to resolve the anaphors in the test documents.

# Results Using the Best Combination on Development and Test Sets

System	Development Set			Test Set		
	R	P	F	R	P	F
Reconcile	26.7	74.0	39.3	22.2	73.3	34.1
EventMine	53.5	69.8	60.5	50.4	62.7	55.9
Our system	59.9	77.1	<b>67.4</b>	55.6	67.2	<b>60.9</b>

# Error Analysis

- Definite Noun Phrases:
  - Our mention detection method is constrained to only extract the seen anaphors in the training set.
- Personal Pronouns:
  - Our system only accounts for intra-sentential pronouns. This affects both precision and recall.

# Conclusion

- Substantiated our hypothesis that different methods are needed for resolving different types of anaphors.
- Proposed a hybrid approach to coreference resolution.