



Coreference Resolution in Russian: State-of-the-Art Approaches Application and Evolvment

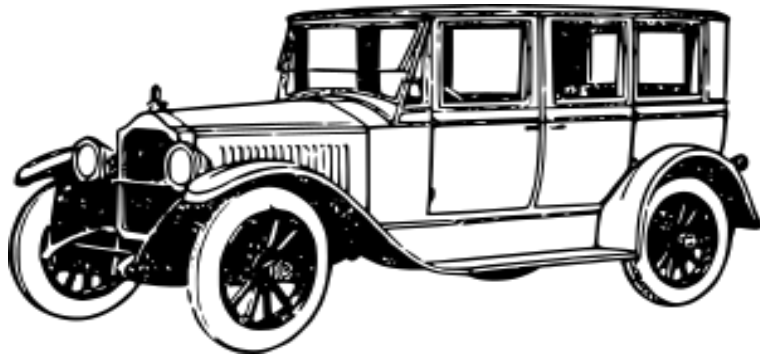
Speaker: Andrey Sysoev
Ivan Andrianov
Alexandra Khadzhiiskaia

Moscow, 2017

Let's start with an example

If a **bulb** in your **car** burned out – change **it**.

Если в **машине** перегорела **лампа** – замените **её**.

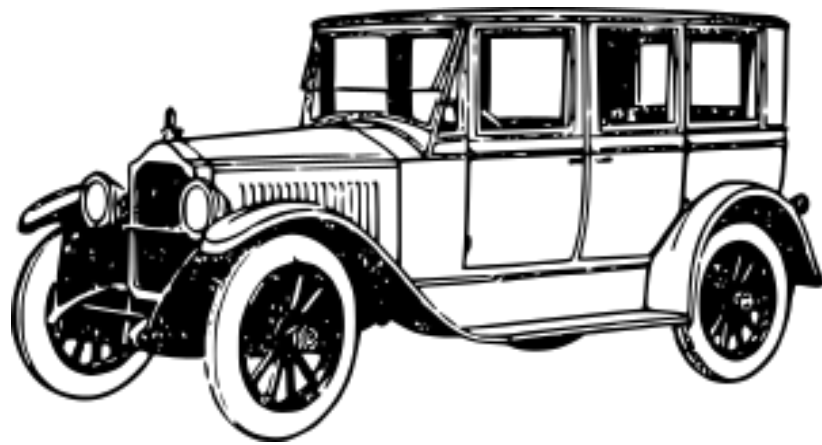


What should I actually change?

Let's start with an example

If a **bulb** in your **car** burned out – change **it**.

Если в **машине** перегорела **лампа** – замените **её**.

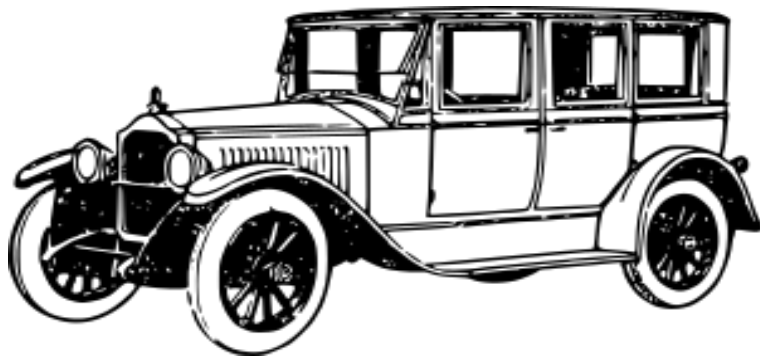


What should I actually change? A car?

Let's start with an example

If a **bulb** in your **car** burned out – change **it**.

Если в **машине** перегорела **лампа** – замените **её**.



What should I actually change? A car? Or just a bulb?

Coreference resolution. Where to use?

- Relation extraction
- Question-answering systems
- Sentiment analysis

Coreference resolution in practice

Mikhail Lomonosov is a famous Russian scientist.

One of his discoveries is the atmosphere of Venus.

Михаил Васильевич Ломоносов – выдающийся русский ученый.

Одно из его открытий – атмосфера Венеры.

Coreference resolution in practice


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Coreference resolution in practice

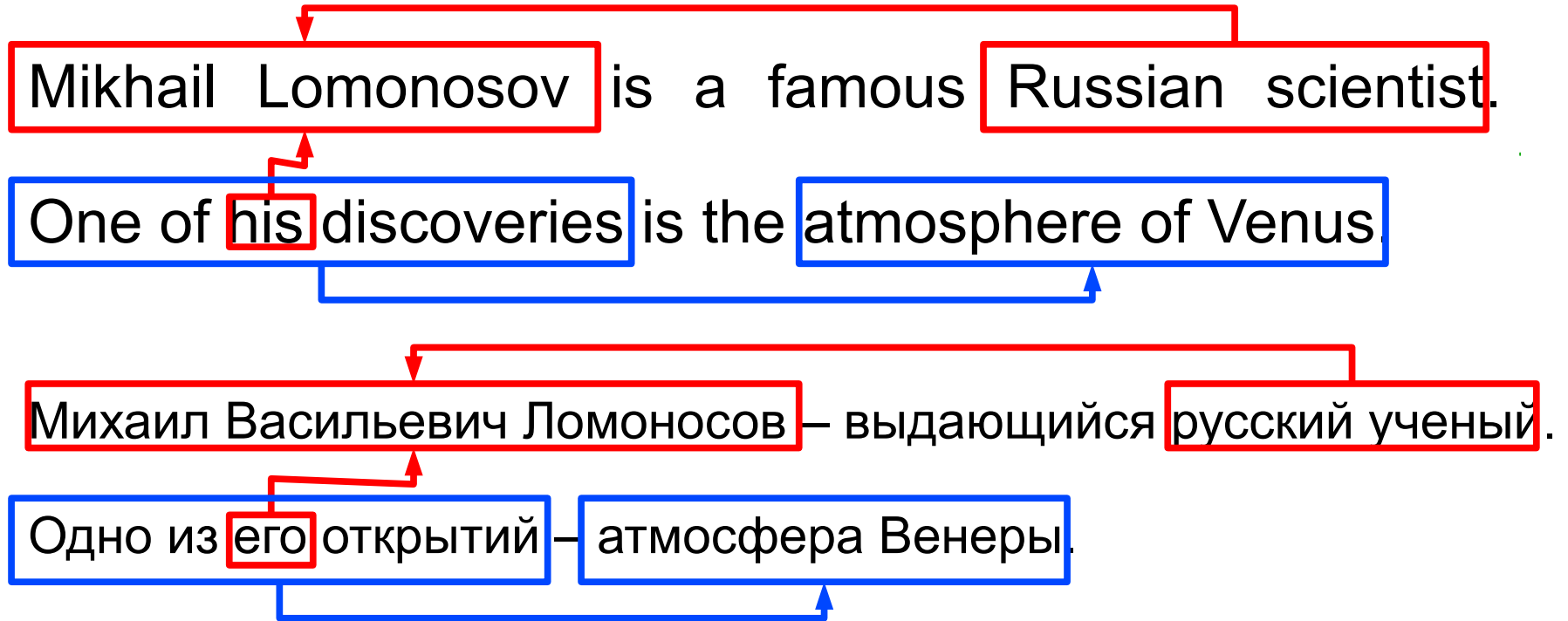
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Coreference resolution in practice



Coreference resolution in practice

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Mikhail Lomonosov
Russian scientist
his



One of his discoveries
atmosphere of Venus

Antecedent and anaphor

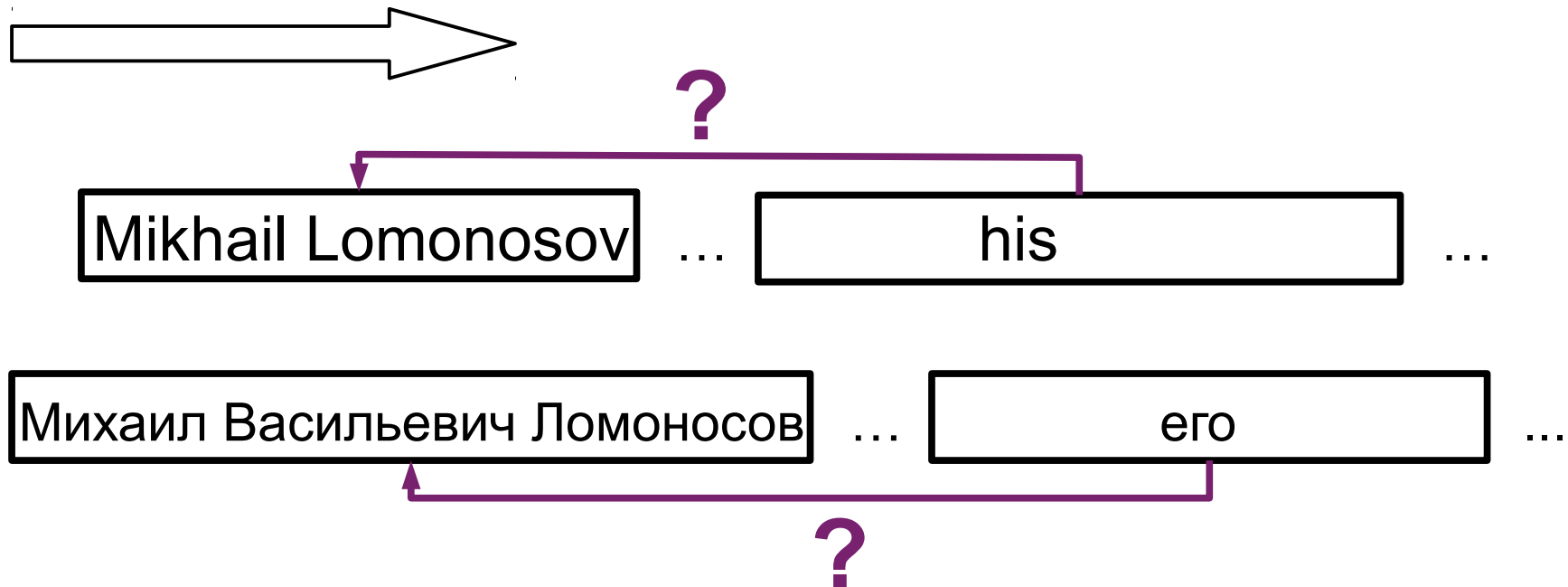
Antecedent - the mention, which already has some meaning (within the text).

Anaphor - the mention, which borrows its meaning from corresponding antecedent.

Antecedent and anaphor

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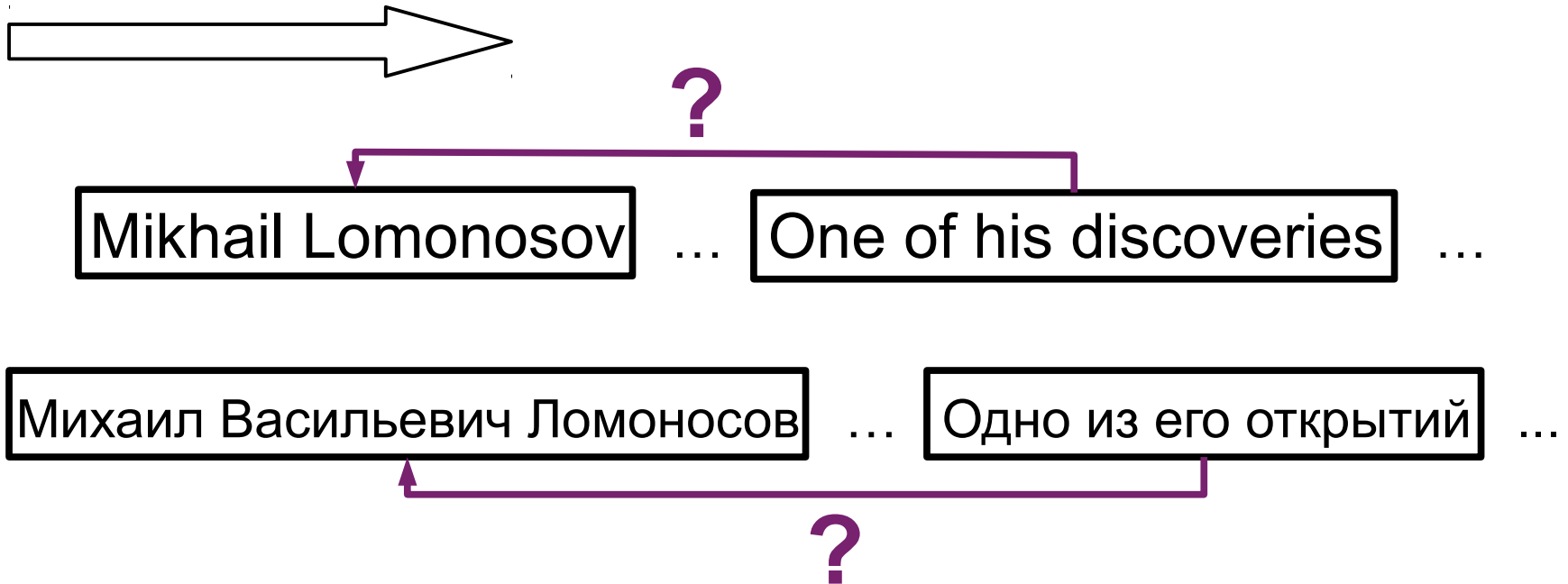
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Antecedent and anaphor

Antecedent - the mention, which already has some meaning (within the text).

Anaphor - the mention, which borrows its meaning from corresponding antecedent.



Evaluation.

Test corpus and metrics

Corpus:

- RuCor from RuEval-2014
- 181 (179 after fixing conflicting markup) documents

Quality measures:

- Precision / Recall / F1
- MUC / B3 / $CEAF_{\text{entity}}$ / $CEAF_{\text{mention}}$

10-fold crossvalidation

Preprocessing

One	of	his	discoveries	is	the	atmosphere	of	Venus	.
OTHER	IN	OTHER	NNS	VBZ	DT	NN	IN	NNP	PUNCT

0.13
0.01
0.04
...

...

0.31
0.11
0.17
...

...

LOCATION



Одно	из	его	открытий	-	атмосфера	Венеры	.
NUM_NEUT_SIN	PR	S_MAS_SIN	S_NEUT_PL	PUNCT	S_FEM_SIN	S_FEM_SIN	PUNCT

0.11
0.42
0.02
...

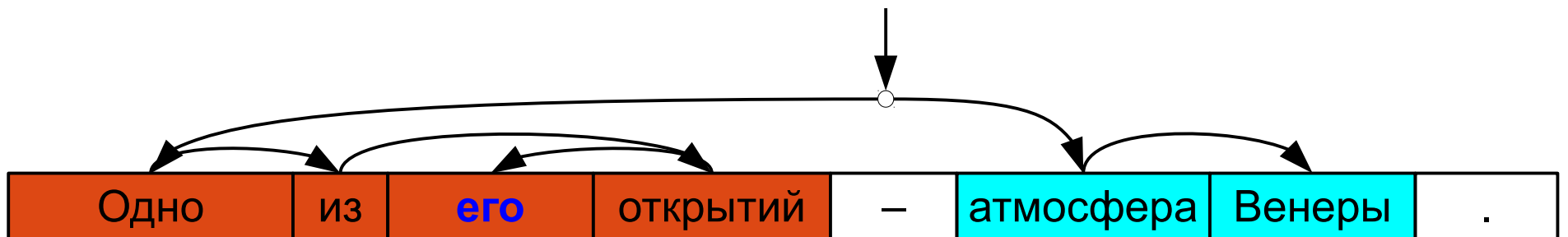
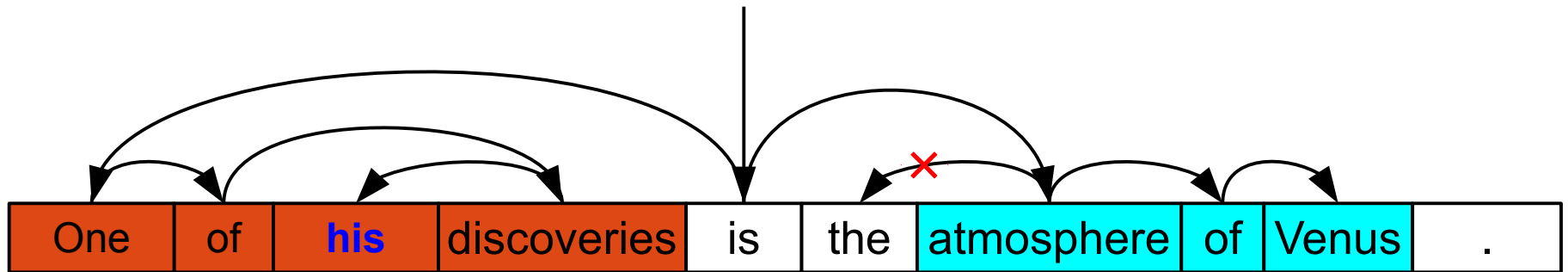
...

0.87
0.03
0.02
...

...

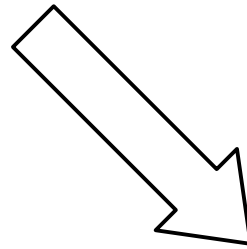
LOCATION

Mention detection. What is mention?



Mention detection

Identify mention heads



Expand heads to full mentions

Peng H., Chang K.-W., Roth D. (2015)

A Joint Framework for Coreference Resolution and Mention Head Detection

Head identification

T

One of **his** discoveries is the atmosphere of **Venus**.

PRONOUN

T

LOCATION

T

Одно из **его** открытий – атмосфера **Венеры**.

PRONOUN

T

LOCATION

Head identification

F T

F F

F T

One **of his** discoveries **is the** atmosphere **of Venus**.

PRONOUN

LOCATION

F T

T

Одно **из его** открытий – атмосфера **Венеры**.

PRONOUN

LOCATION

Head identification. Features

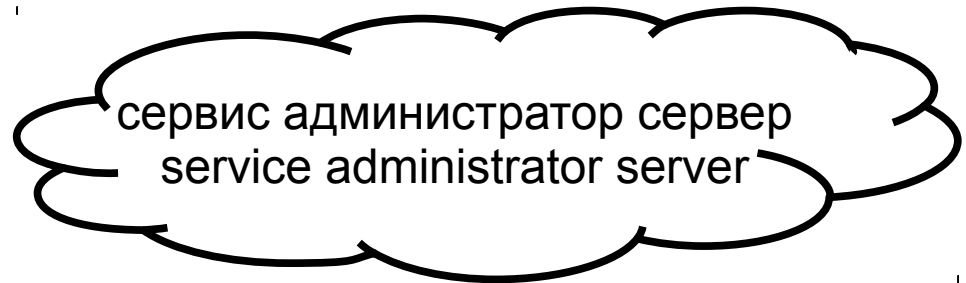
- Internal morphological
 - POS-tag
 - number
 - gender
 - animacy
- Syntax
 - position within sentence
 - relations of a token
- Syntactic context - morphological features for syntactic parent
- Context - basic morphological features for neighbors
- Frequency - TF weighting
- **Semantic**

Head identification. Semantic features

- Groundtruth heads from training documents are clustered (word2vec, k-means).



суд юрист свидетель
court lawyer witness



сервис администратор сервер
service administrator server



университет исследование
студент
university research student

Head identification. Semantic features

- Groundtruth heads from training documents are clustered (word2vec, k-means).
- Features - distance and similarity from head candidate to clusters centroids.



Head identification

T F T F F F T F T

One **of his** discoveries **is the** atmosphere **of Venus**.

PRONOUN

LOCATION

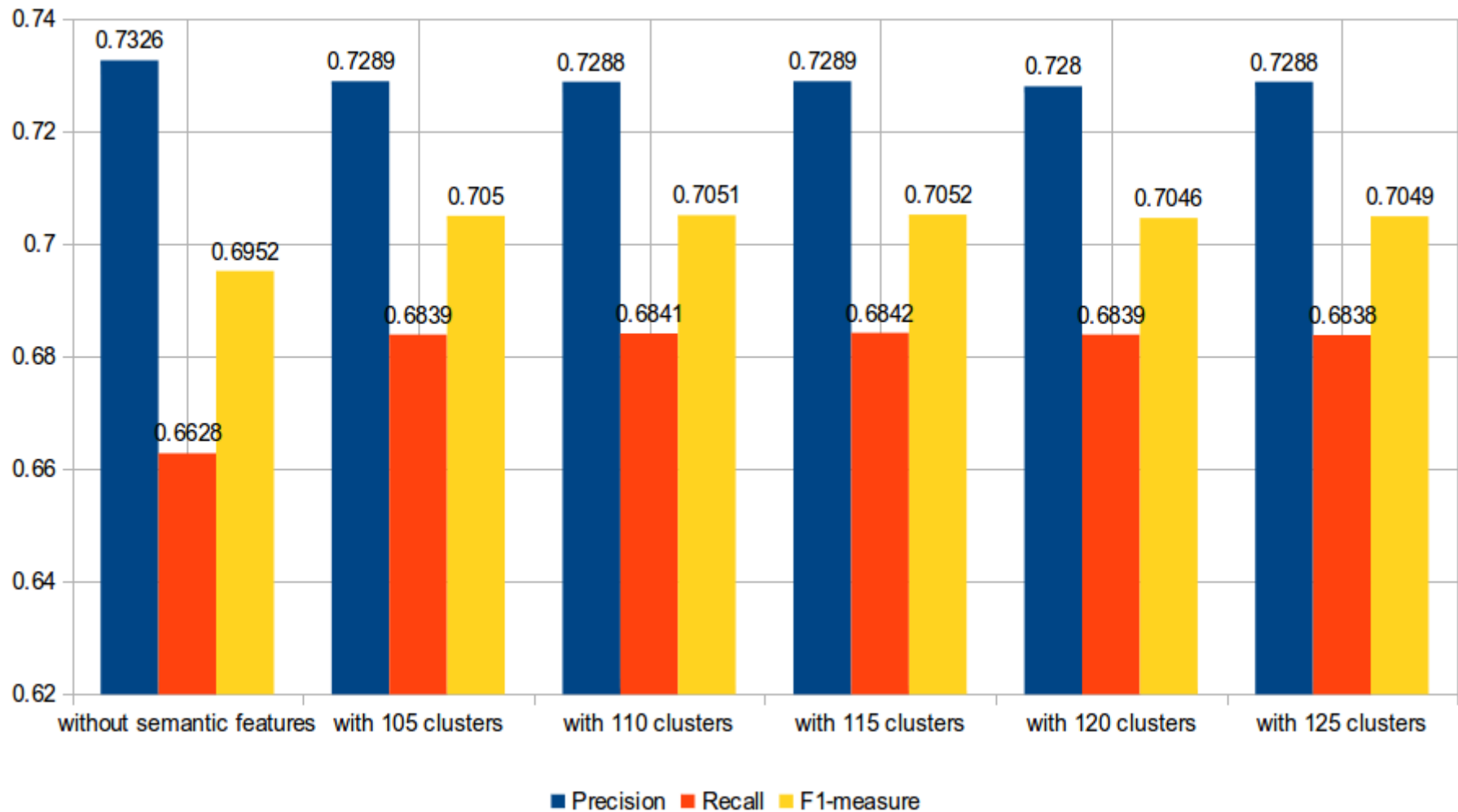
T F T F T T

Одно **из его** открытий – атмосфера **Венеры**.

PRONOUN

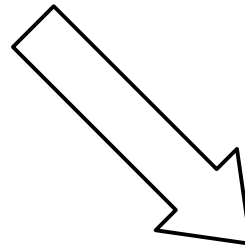
LOCATION

Head identification. Evaluation



Mention detection

Identify mention heads



Expand heads to full mentions

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Head expansion

One of his discoveries is the **atmosphere** of Venus.

Одно из его открытий – **атмосфера** Венеры.

Head expansion



One of his discoveries is the **atmosphere** of Venus.



Одно из его открытий – **атмосфера** Венеры.

Head expansion

One of his discoveries is the **atmosphere** *of* Venus.




Одно из его открытий – **атмосфера** Венеры.

Head expansion

One of his discoveries is the **atmosphere of Venus**.



Одно из его открытий – **атмосфера Венеры**.




Head expansion

One of his discoveries is the **atmosphere of Venus**.



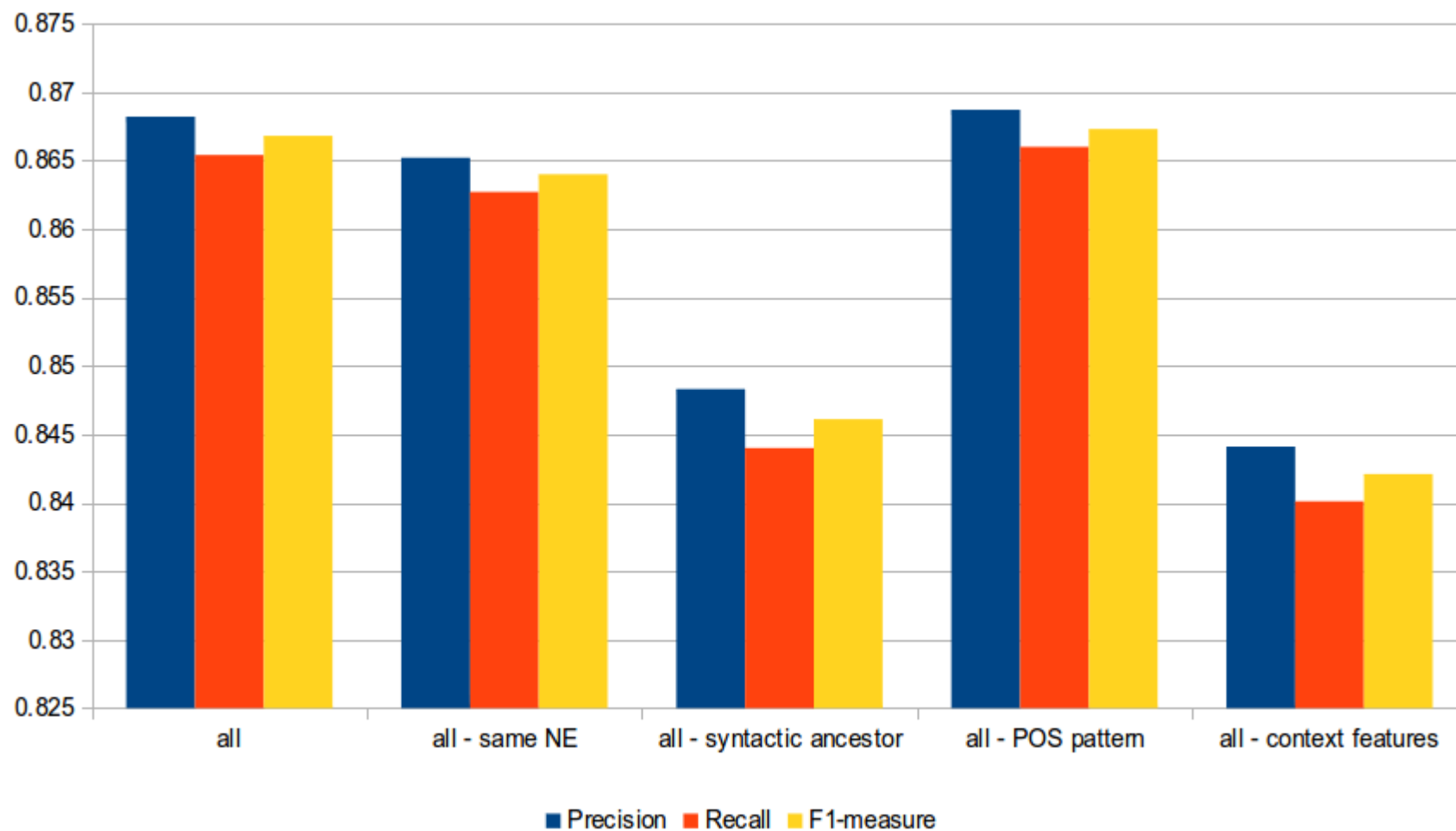
Одно из его открытий – **атмосфера Венеры**.



Head expansion. Features

- Token-based [head, candidate token, nearest neighbours]
 - word form
 - lemma
 - POS-tag
- Position-based
 - direction from head to candidate
 - distance between head and candidate
 - head/candidate is the first/last token of the sentence
- Context-based
 - head and candidate are parts of the same named entity
 - head/candidate is a syntactic ancestor of candidate/head
 - POS-tag pattern for words between head and candidate

Head expansion. Evaluation



Coreference resolution. Easy-First Mention Pair algorithm

*Uryupina O., Moschitti A. (2015)
A State-of-the-Art Mention-Pair Model for Coreference Resolution*

Coreference resolution.

Easy-First Mention Pair algorithm

Classified pairs
(antecedent-anaphor)

Mikhail Lomonosov - Russian scientist

One of his discoveries - his

One of his discoveries - atmosphere of Venus

Mikhail Lomonosov - his

Russian scientist - his

Mikhail Lomonosov - atmosphere of Venus

Mikhail Lomonosov - One of his discoveries

Russian scientist - One of his discoveries

Russian scientist - atmosphere of Venus

his - atmosphere of Venus

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Mikhail
Lomonosov

Russian
scientist

Mikhail Lomonosov - Russian scientist

One of his discoveries - his

One of his discoveries - atmosphere of Venus

Mikhail Lomonosov - his

Russian scientist - his

Mikhail Lomonosov - atmosphere of Venus

Mikhail Lomonosov - One of his discoveries

Russian scientist - One of his discoveries

Russian scientist - atmosphere of Venus

his - atmosphere of Venus

his

atmosphere of
Venus

One of his
discoveries

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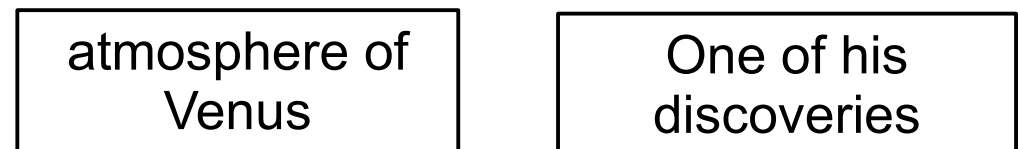
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Classified pairs
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his

- One of his discoveries - his
- One of his discoveries - atmosphere of Venus
- Mikhail Lomonosov - his
- Russian scientist - his
- Mikhail Lomonosov - atmosphere of Venus
- Mikhail Lomonosov - One of his discoveries
- Russian scientist - One of his discoveries
- Russian scientist - atmosphere of Venus
- his - atmosphere of Venus



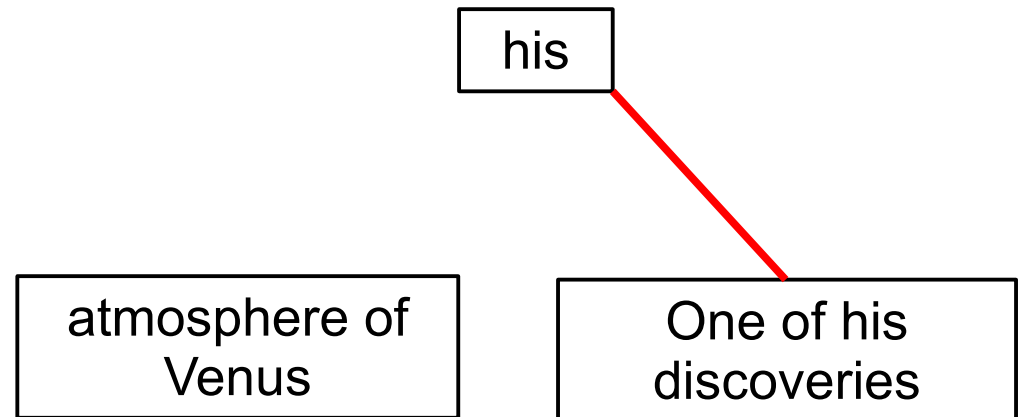
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One of his discoveries - atmosphere of Venus
Mikhail Lomonosov - his
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Russian scientist - One of his discoveries
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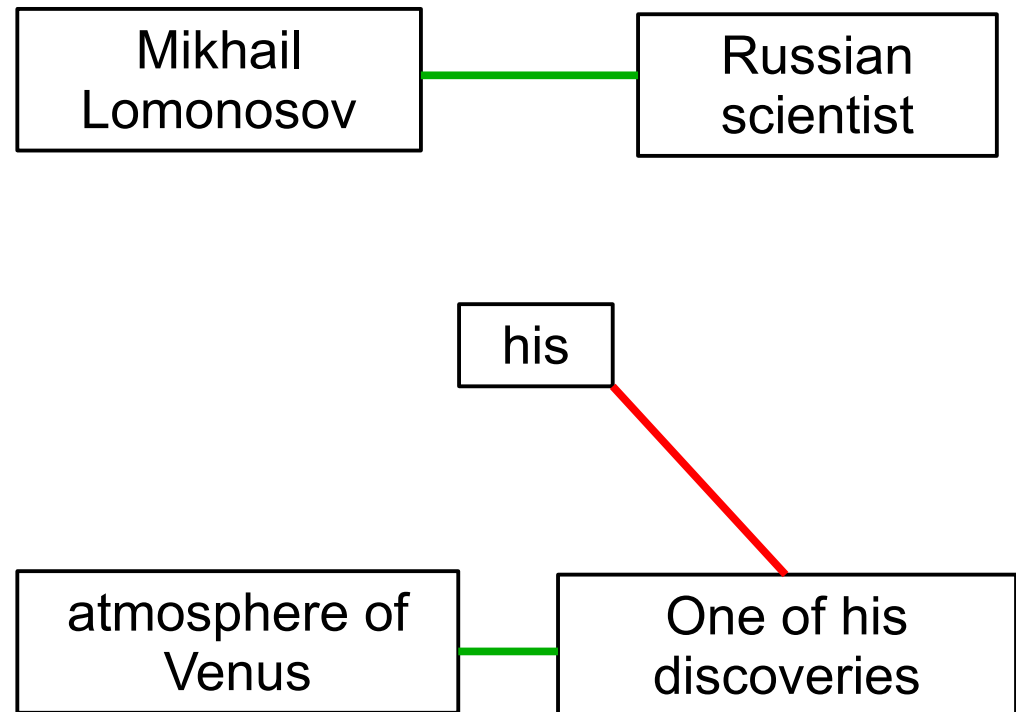


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Russian scientist - One of his discoveries
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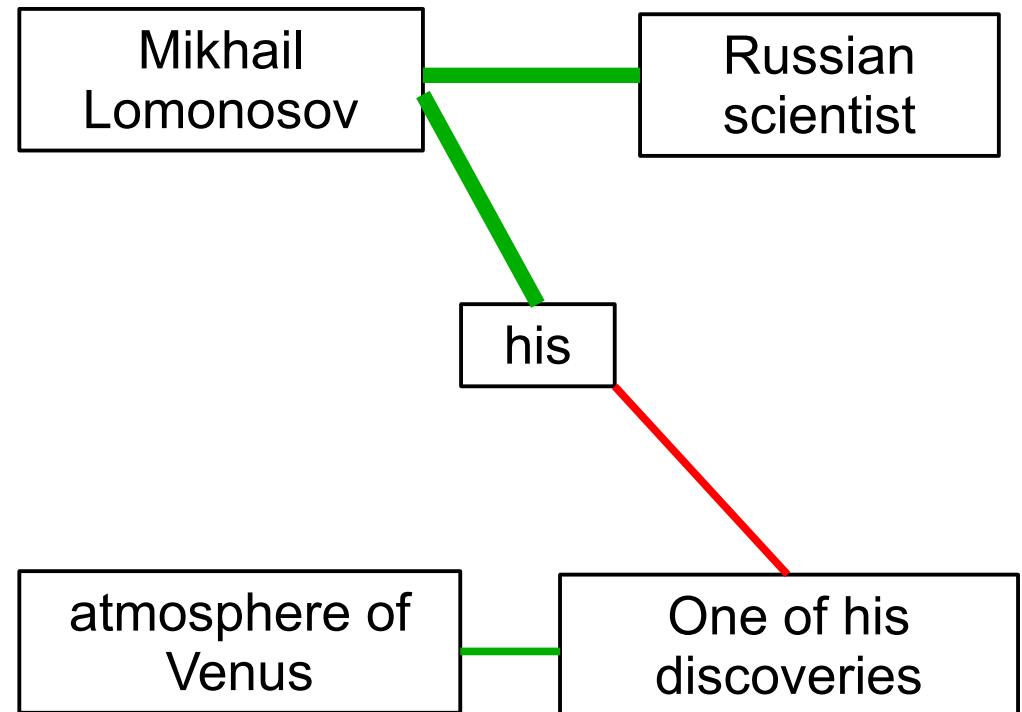


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Classified pairs
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Russian scientist – his
Mikhail Lomonosov - atmosphere of Venus

Mikhail Lomonosov - One of his discoveries
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Russian scientist - atmosphere of Venus
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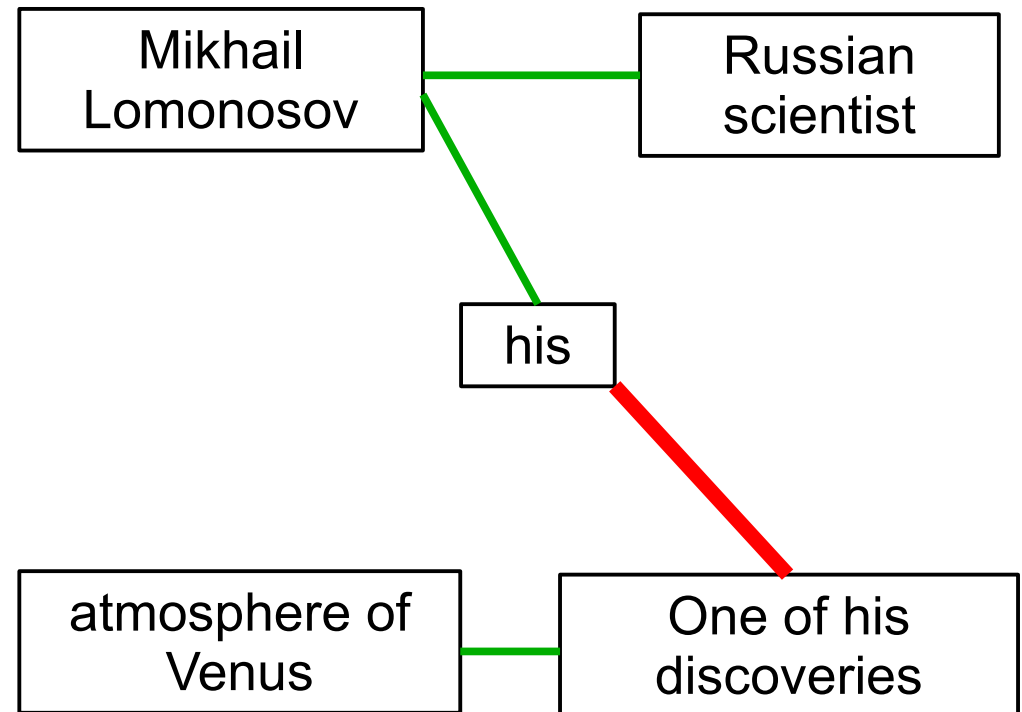
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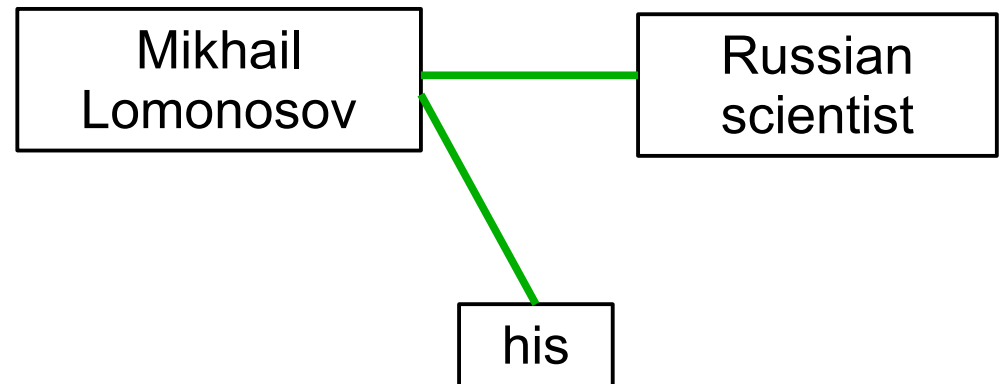


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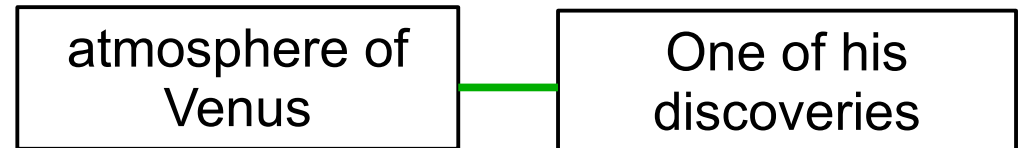
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Russian scientist - One of his discoveries
Russian scientist - atmosphere of Venus
his - atmosphere of Venus



Coreference resolution.

Mention-pair classifier features

- Basic linguistic: word forms, lemmas, part-of-speech tags, grammemes (gender, number, animacy) for mention head and context words.
- Grammmemes agreement: mention heads share the same key grammemes (number, gender, animacy, pro-nominality).
- Positional: distance, place within sentence boundaries.
- Named entity: mention types and their agreement.
- Structural: mention size and interrelation with other mentions of the text.
- Surface form matching: lexicographic similarity and textual representation equality indicators.
- Syntactic: grammar role, sharing same parent node or clause.

Jaccard Item Set mining

Uryupina O., Moschitti A. (2015)
A State-of-the-Art Mention-Pair Model for Coreference Resolution

Segond M., Borgelt C. (2011)
Item Set Mining Based on Cover Similarity

Jaccard Item Set mining

	ANT_PER	ANAF_PER	LEX_SIM>0.5	NUM	ANIM	SENT_ST
<i>TRUE</i>	F	F	T	F	F	T
<i>FALSE</i>	T	T	F	F	T	F
<i>TRUE</i>	T	F	F	F	T	T
<i>FALSE</i>	F	F	F	T	F	F
<i>TRUE</i>	T	F	T	F	T	T
<i>FALSE</i>	T	T	F	F	T	T
<i>TRUE</i>	F	T	T	F	F	F

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TRUE	F	F	T	F	F	T
FALSE	T	T	F	F	T	F
TRUE	T	F	F	F	T	T
FALSE	F	F	F	T	F	F
TRUE	T	F	T	F	T	T
FALSE	T	T	F	F	T	T
TRUE	F	T	T	F	F	F

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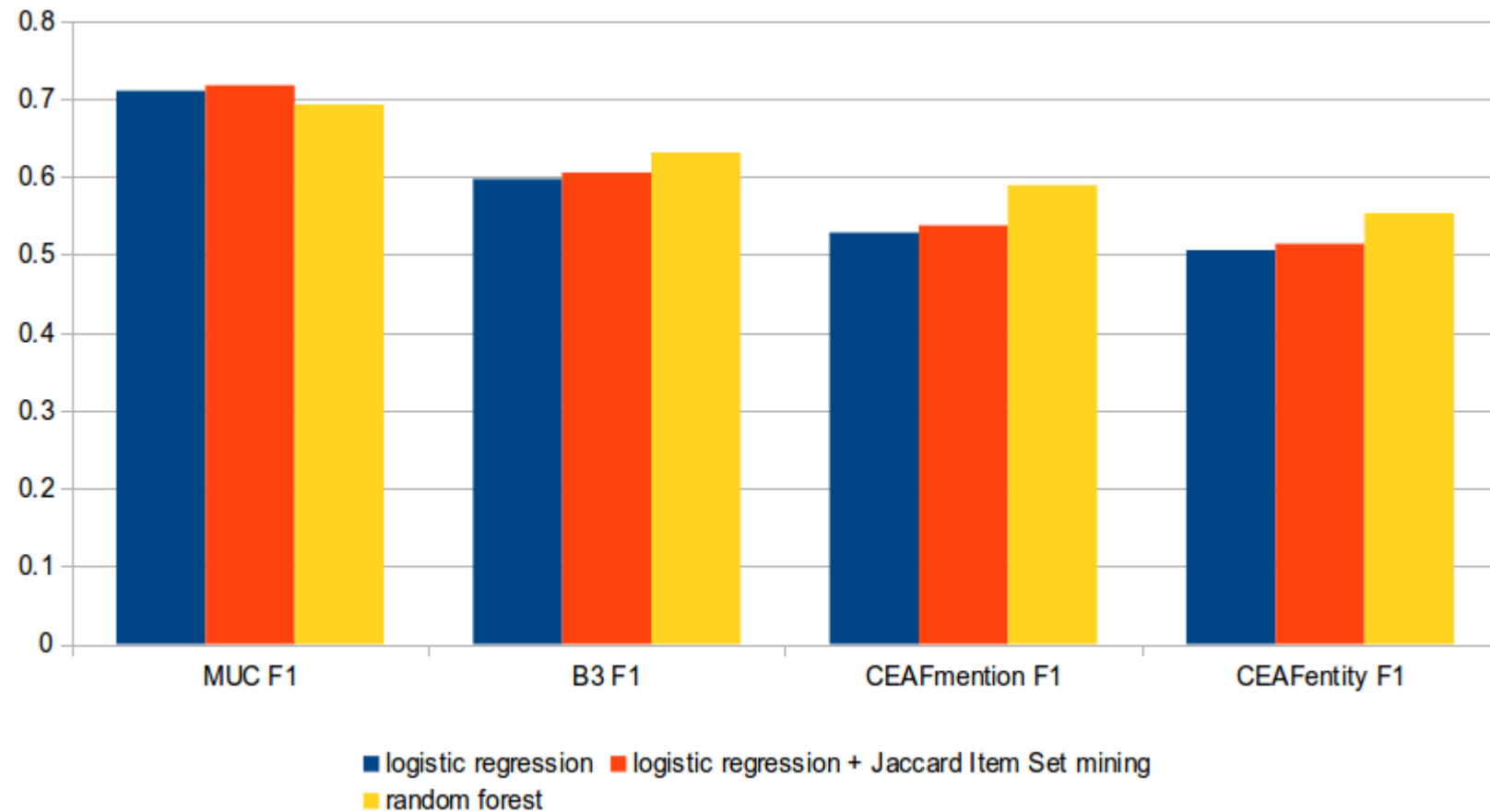
	ANT_PER	ANAF_PER	LEX_SIM>0.5	NUM	ANIM	SENT_ST
TRUE	F	F	T	F	F	T
FALSE	T	T	F	F	T	F
TRUE	T	F	F	F	T	T
FALSE	F	F	F	T	F	F
TRUE	T	F	T	F	T	T
FALSE	T	T	F	F	T	T
TRUE	F	T	T	F	F	F

ANT_PER & ANAF_PER & ~(LEX_SIM>0.5)

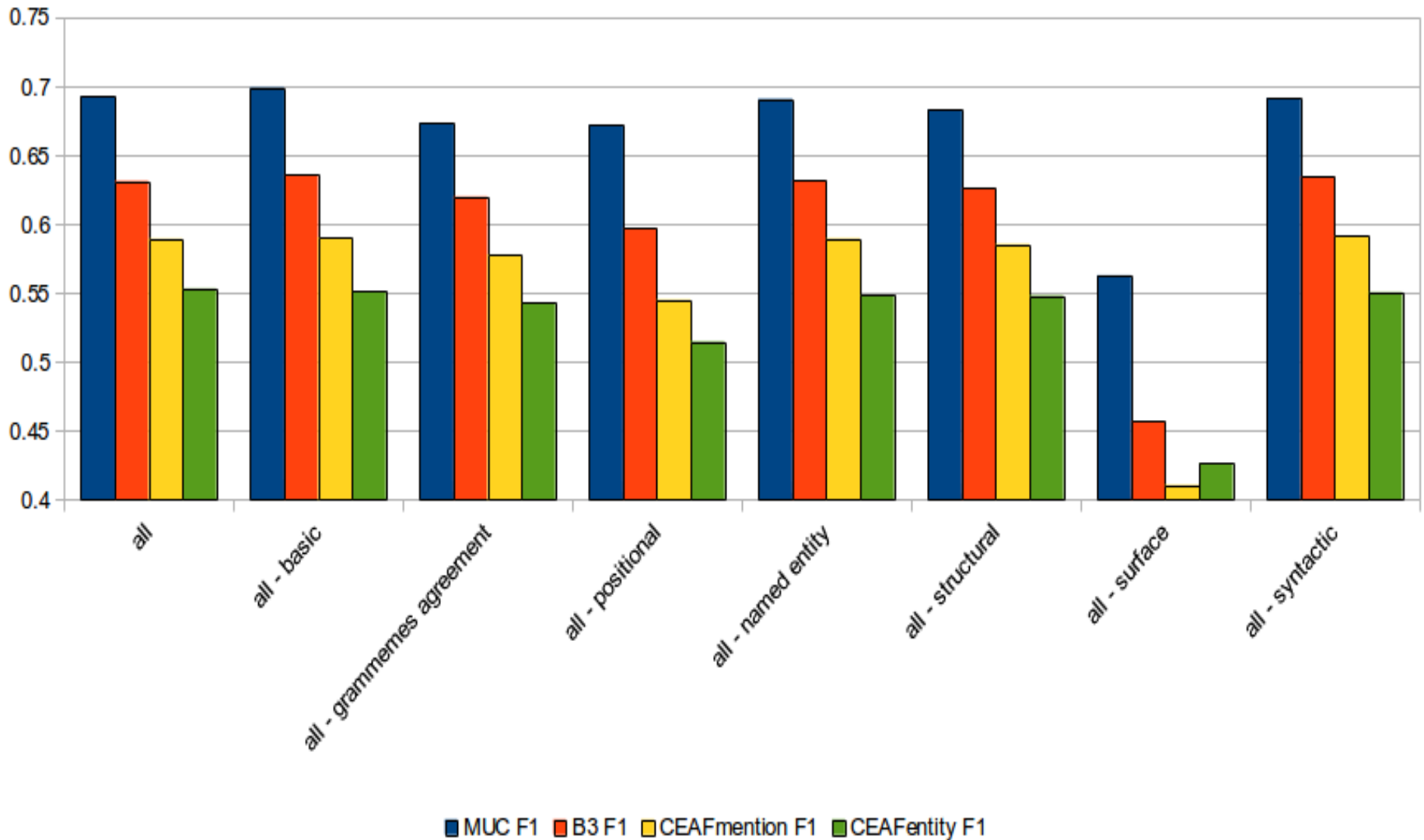
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Coreference resolution. Selecting classifier



Coreference resolution. Ablation analysis



Coreference resolution. In the wild

	Precision	Recall	F1
MUC	0.4768	0.3741	0.4189
B3	0.4104	0.2957	0.3431
CEAF _{mention}	0.4024	0.3702	0.3854
CEAF _{entity}	0.2525	0.3433	0.2906

Future work

- Experiments with more machine learning algorithms and approaches.
- Using various clustering algorithms for word embeddings.
- Detailed analysis of features, assumed useless in ablation experiments.
- Tuning coreference resolution algorithm for different mention types.

Credits

Alexandra Khadzhiiskaia

Ivan Andrianov

Thank you!