

GLAD: Groningen Lightweight Authorship Detection

PAN, Authorship verification, 2015



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The challenge

- given: a set of **Known documents** written by the **same Author A_K** ,
- given: **one Unknown document** written by an **unknown Author A_U** ,
- task: determine whether **$A_U = A_K$**



Gentle Tony



Fat Vinny



The Weasel

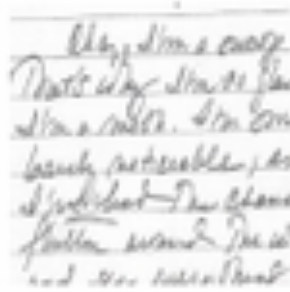
How can we recognise different authors?



Gentle Tony



Fat Vinny



The Weasel

How can we recognise different authors?

Unusual word choice?

Shorter sentences?

More complex grammar?



Gentle Tony



Fat Vinny



The Weasel

How can we recognise different authors?

`individual_vector(feet1, feet2...)`

`individual_vector(feet1, feet2...)`

`individual_vector(feet1, feet2...)`



Gentle Tony



Fat Vinny



The Weasel

**How can we then differentiate
between authors?**

Our approach

- machine learning approach training on PAN (2015) data
- using SVM to do two-class classification task
- a set of features
- feature ablation studies to tune the system to each different language

The core aim

- A lightweight system!



The aim

training
instance

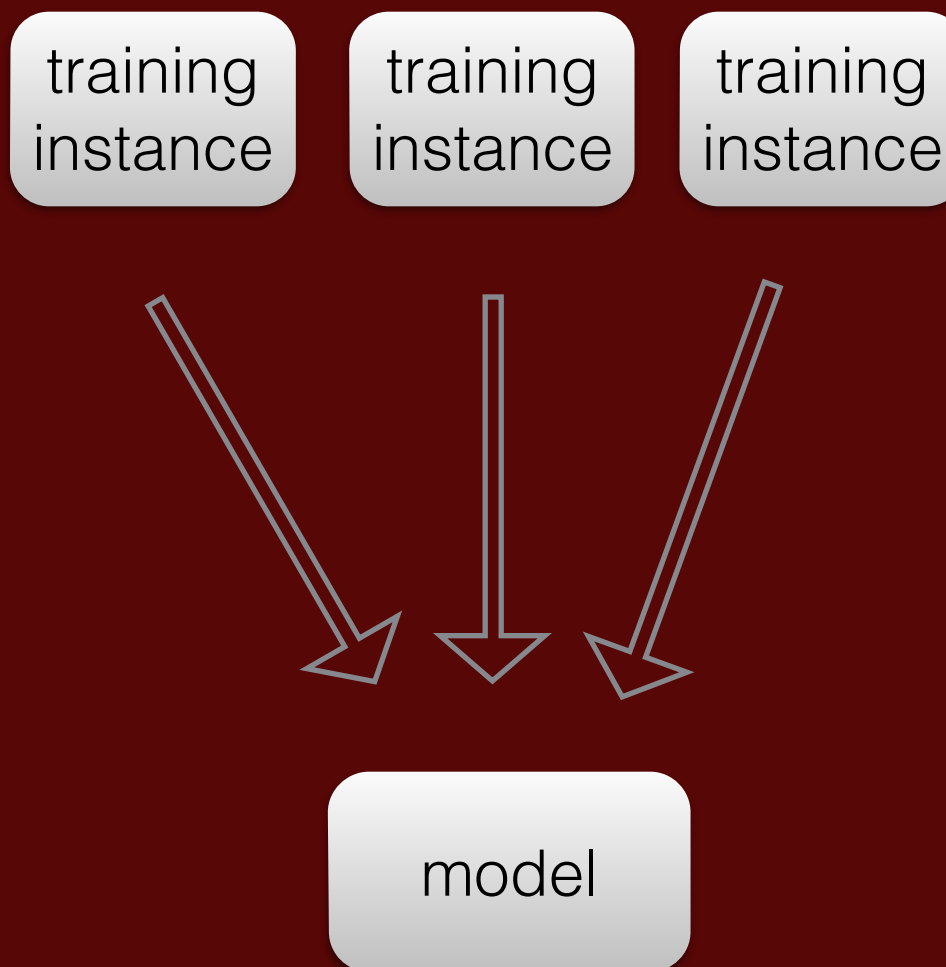
training
instance

training
instance

**Input in
any
language**



The aim

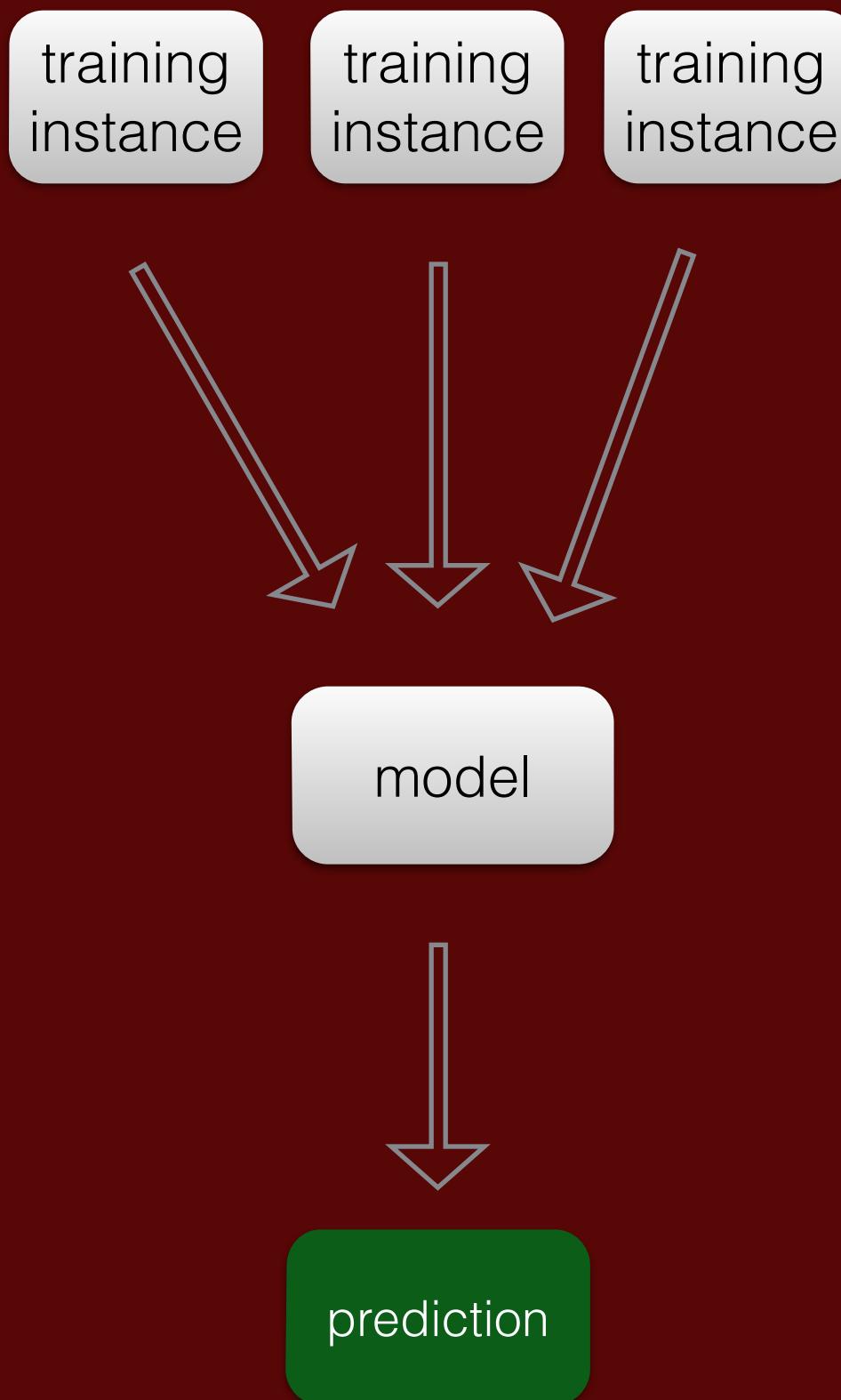


**Input in
any
language**

**Features
should be
easy to
extract**



The aim



**Input in
any
language**

**Features
should be
easy to
extract**

**Training &
Testing
time
should be
fast**


Our features



entropy
of known



entropy
of unknown



joint entropy



n-gram overlap



morpho-
syntactic
similarity



sentence length
of known



sentence length
of unknown



visual features

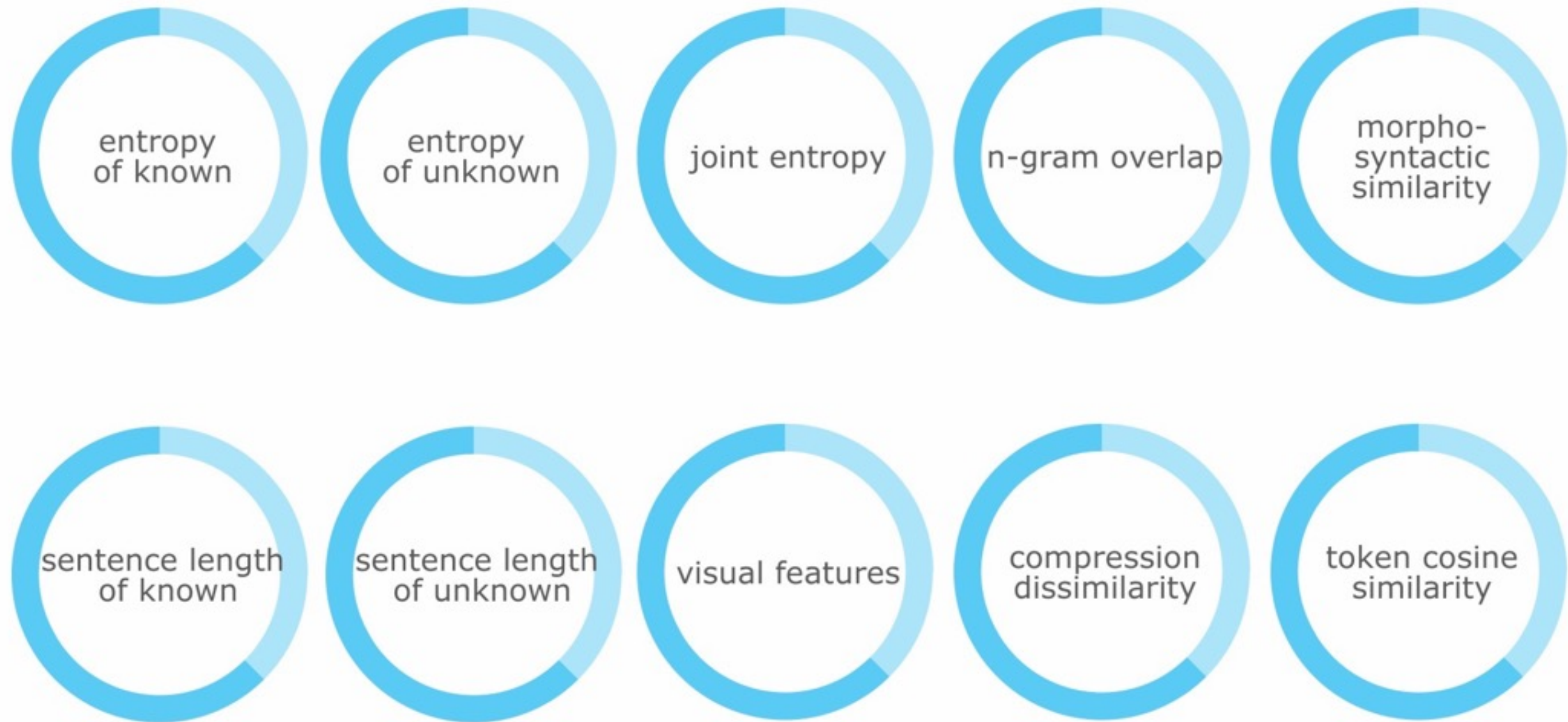


compression
dissimilarity



token cosine
similarity

Our features



`similarity_vector(entropy_of_known, visual_features, ...)`

Our features



entropy
of known

entropy
of unknown

joint entropy

n-gram overlap

morpho-
syntactic
similarity

sentence length
of known

sentence length
of unknown

visual features

compression
dissimilarity

token cosine
similarity

To determine relevance: grouping

Our features

entropy
of known

entropy
of unknown

joint entropy

n-gram overlap

morpho-
syntactic
similarity

sentence length
of known

sentence length
of unknown

visual features

compression
dissimilarity

token cosine
similarity

Individual



Vector_K(feats1, feats2)

Individual



Vector_U(feats1, feats2)

Joint

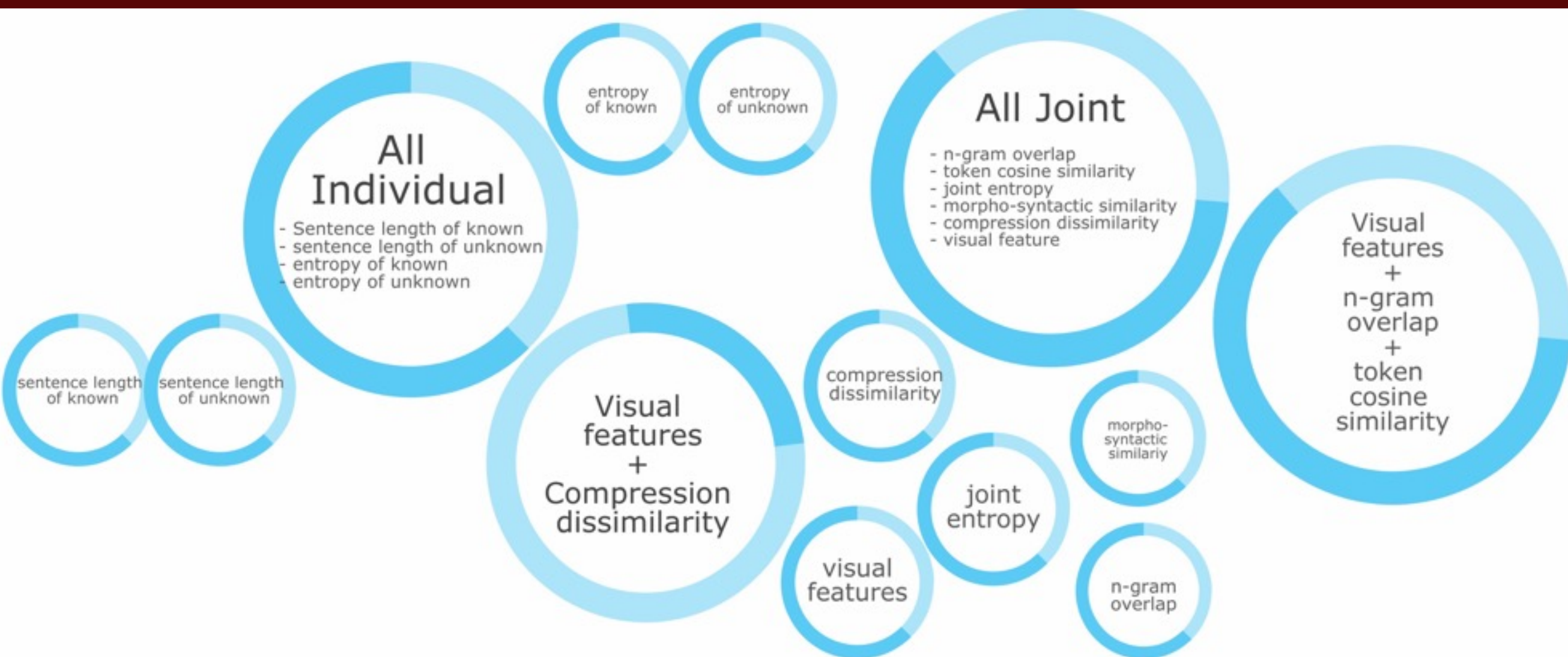


Vector_Joint(feats1, feats2)

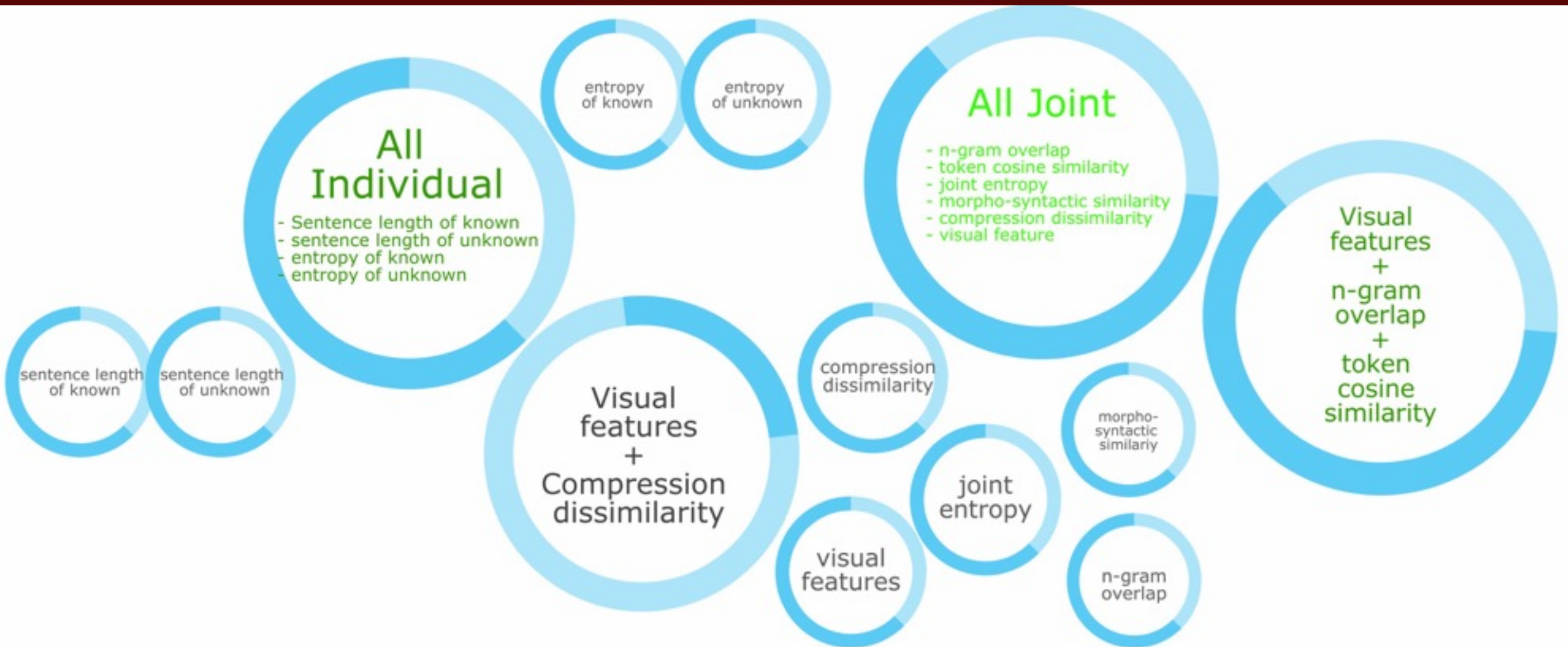
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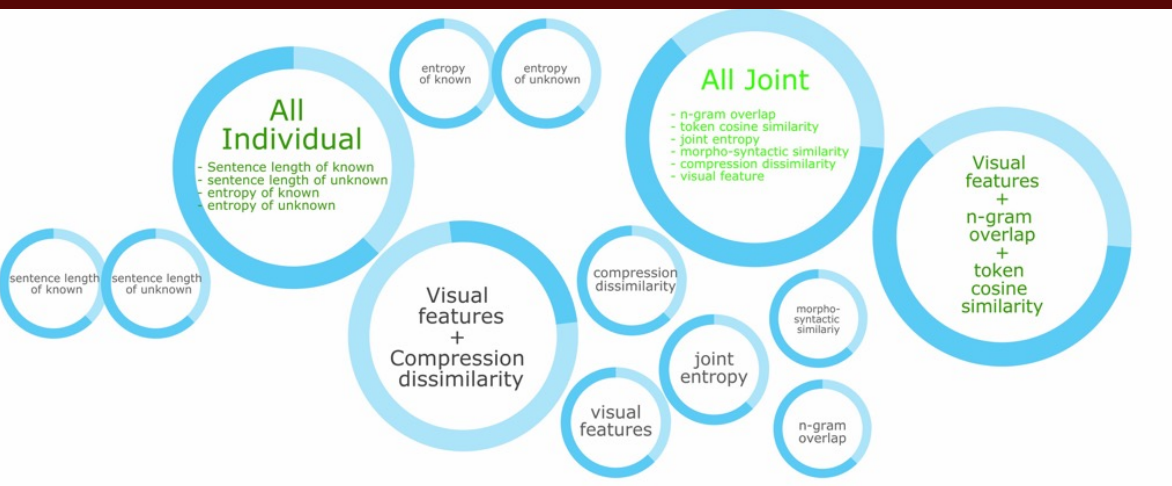
Comparing features



Comparing features

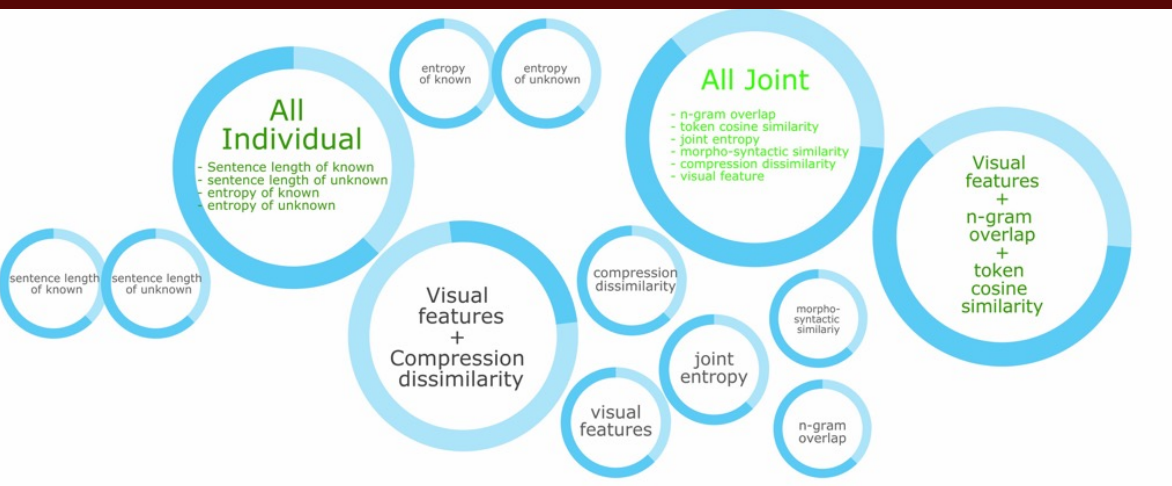


Results of ablation & single-feature experiments:
Helpful features



Side note: Visual features

- Punctuation
- Line ending
- Letter case
- Ling length
- Block size

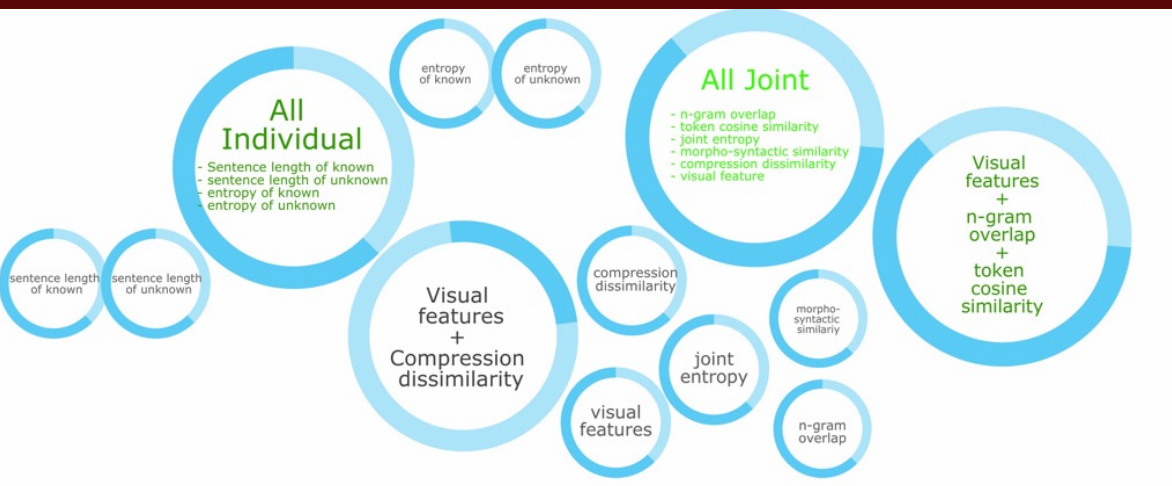


Side note:
Visual features

- Punctuation
- Line ending
- Letter case
- Ling length
- Block size

Con

- **Not a characteristic of the author**
- **Not a linguistic feature**



“Pa-pa, pa-pa, pa-pa!

Here, stop her. She'll fall down.
Here, turn around. Walk this
way.

Ma-ma, ma-ma, ma-ma;

Oh, I think you are a darling.

Mer-ry Christ-mas! Mer-ry
Christmas.”

Side note:
Visual features

- Punctuation
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- Letter case
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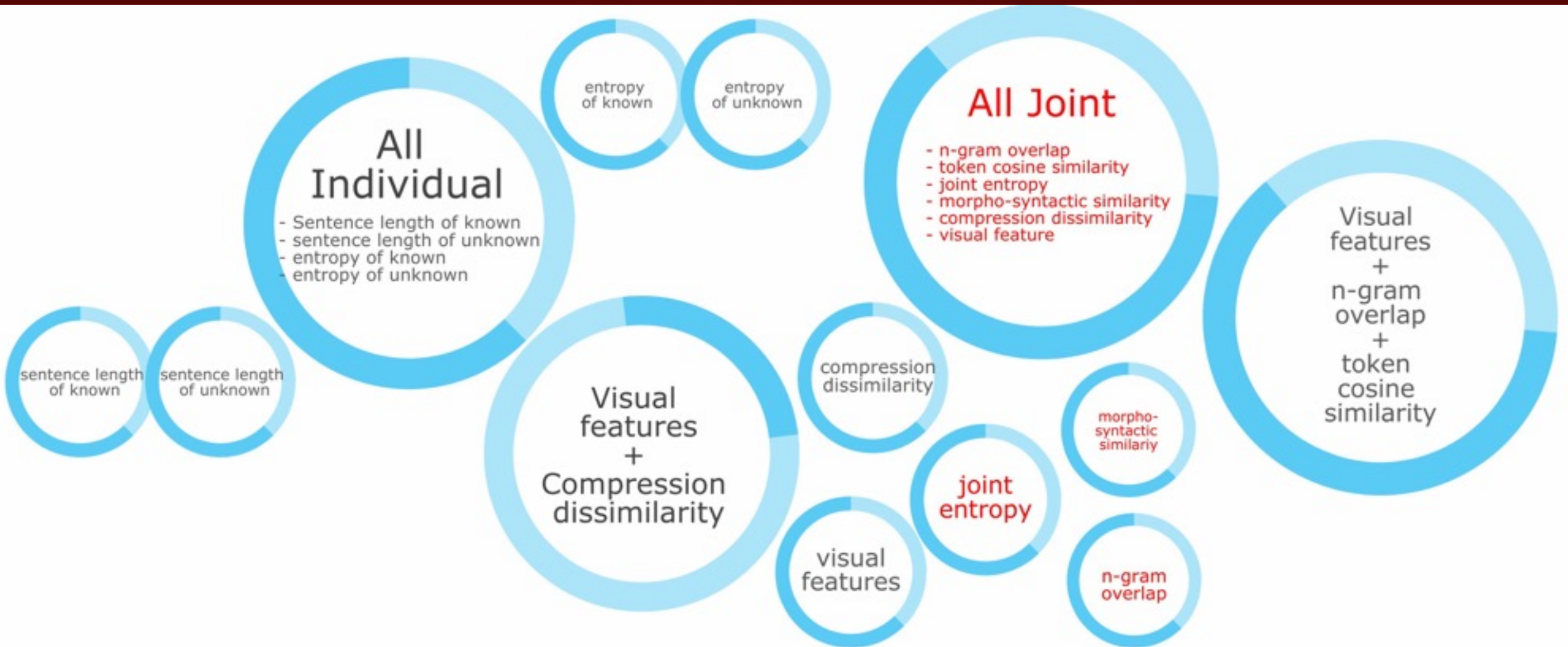
Con

- **Not a characteristic of the author**
- **Not a linguistic feature**

Pro

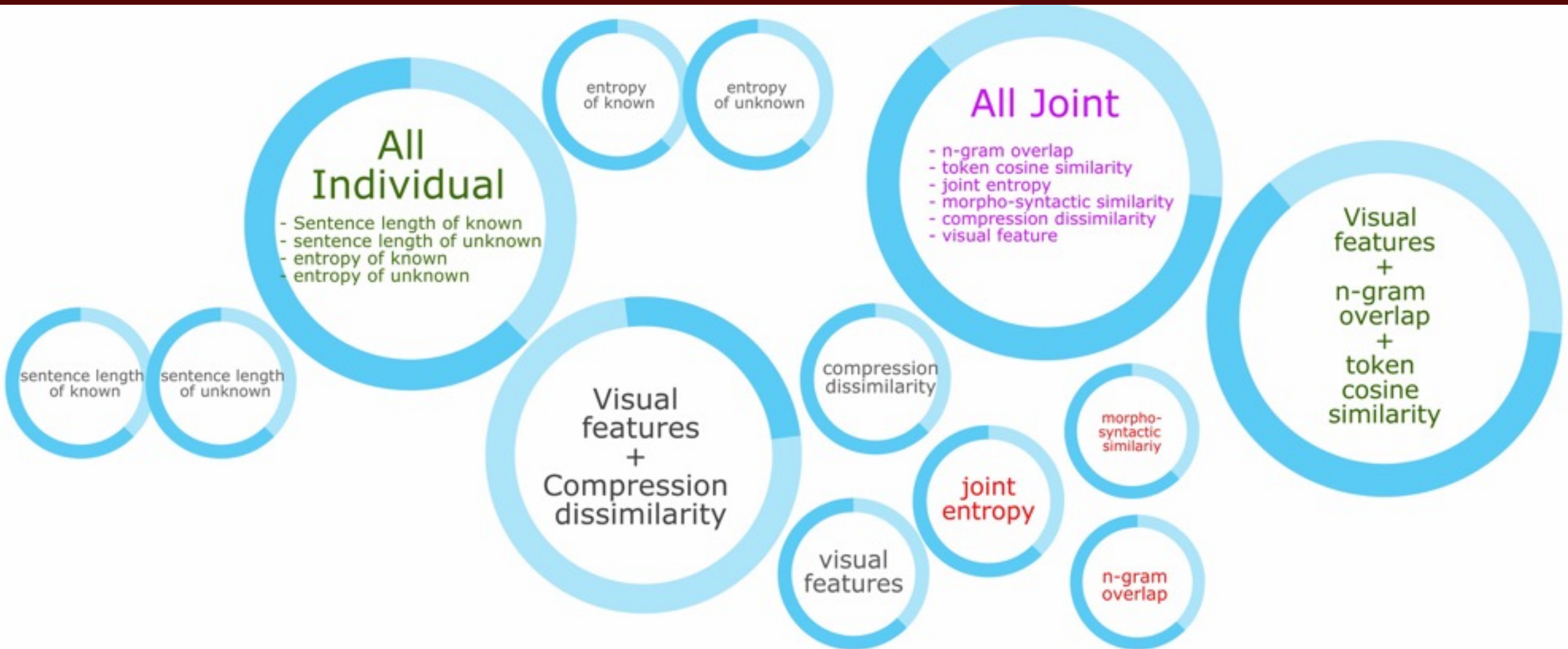
- **Can be author-specific for some genres**
- **If it works...**

Comparing features



Results of ablation & single-feature experiments:
Harmful features

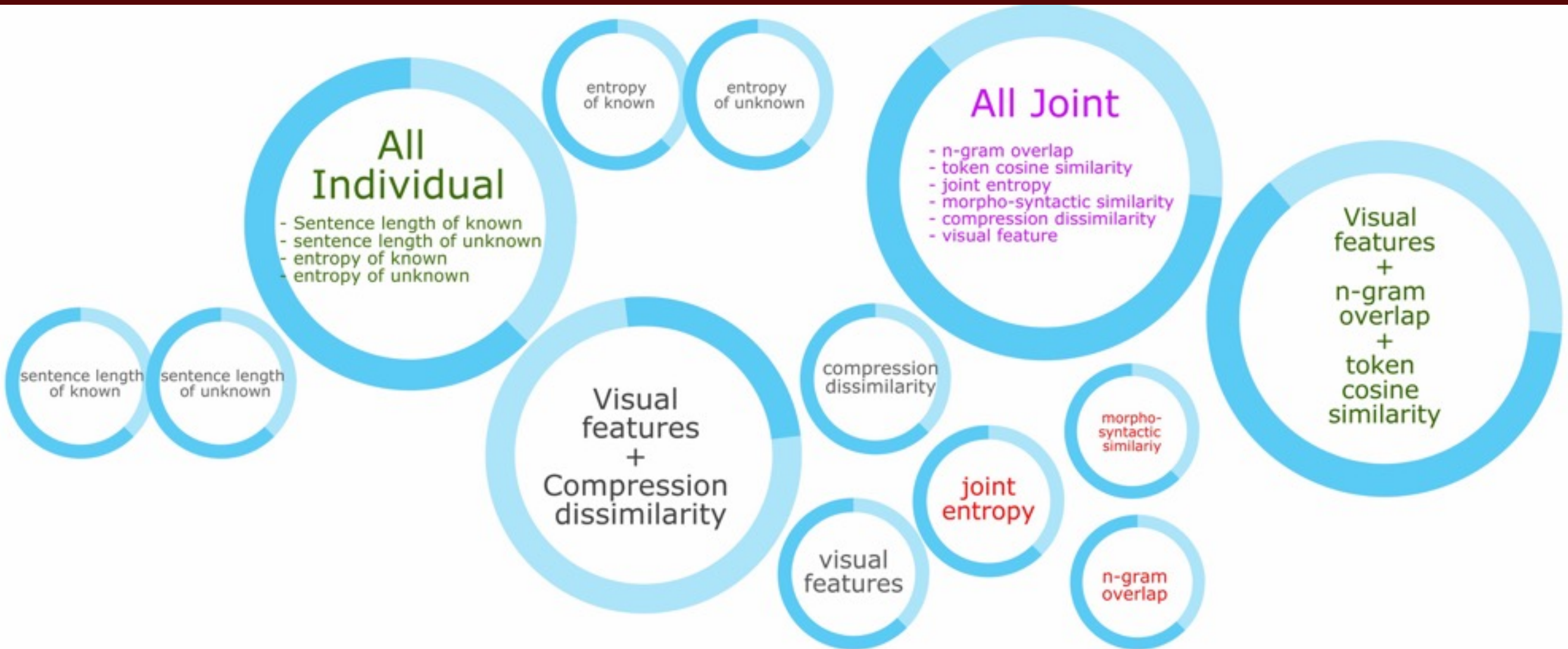
Comparing features



Results of ablation & single-feature experiments:

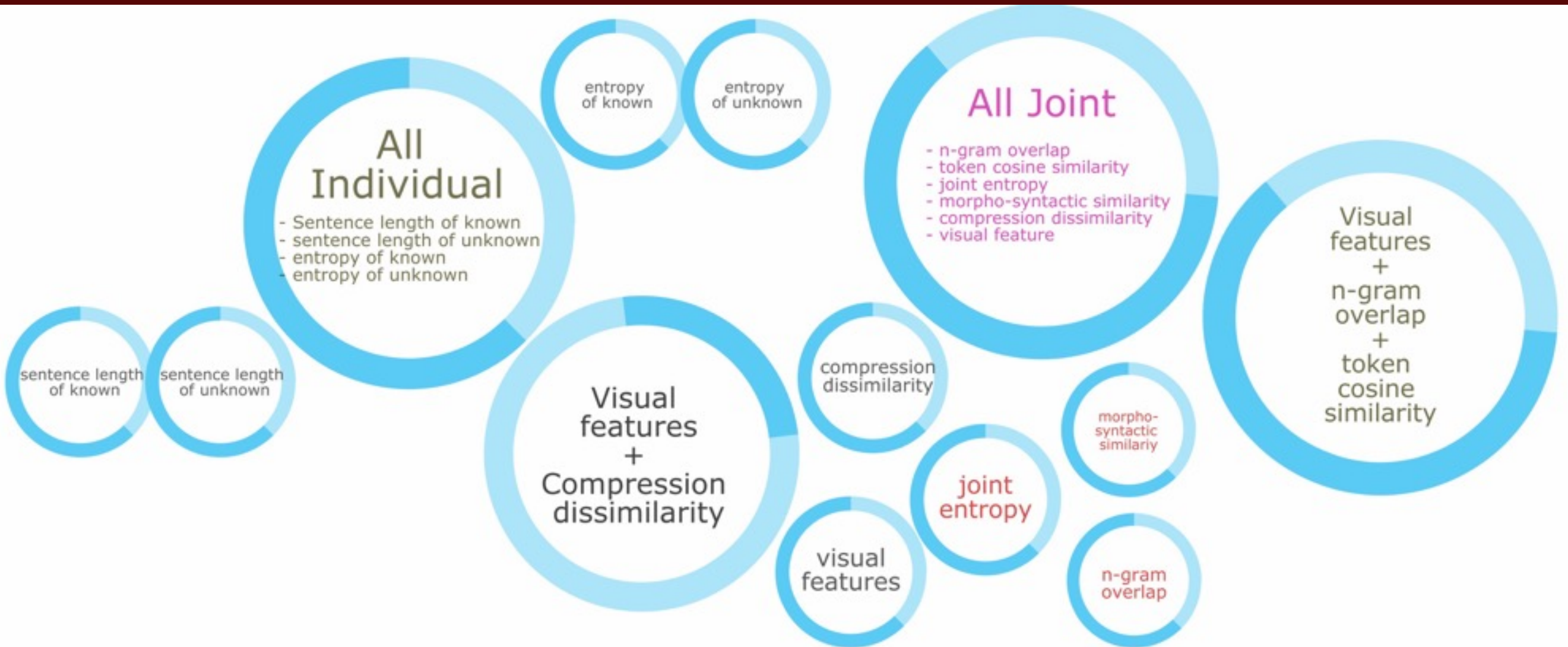
Features that are harmful, helpful, or helpful-depending-on-the-language

Comparing features



Results of ablation & single-feature experiments:
Features that are harmful, helpful, or
helpful-depending-on-the-language

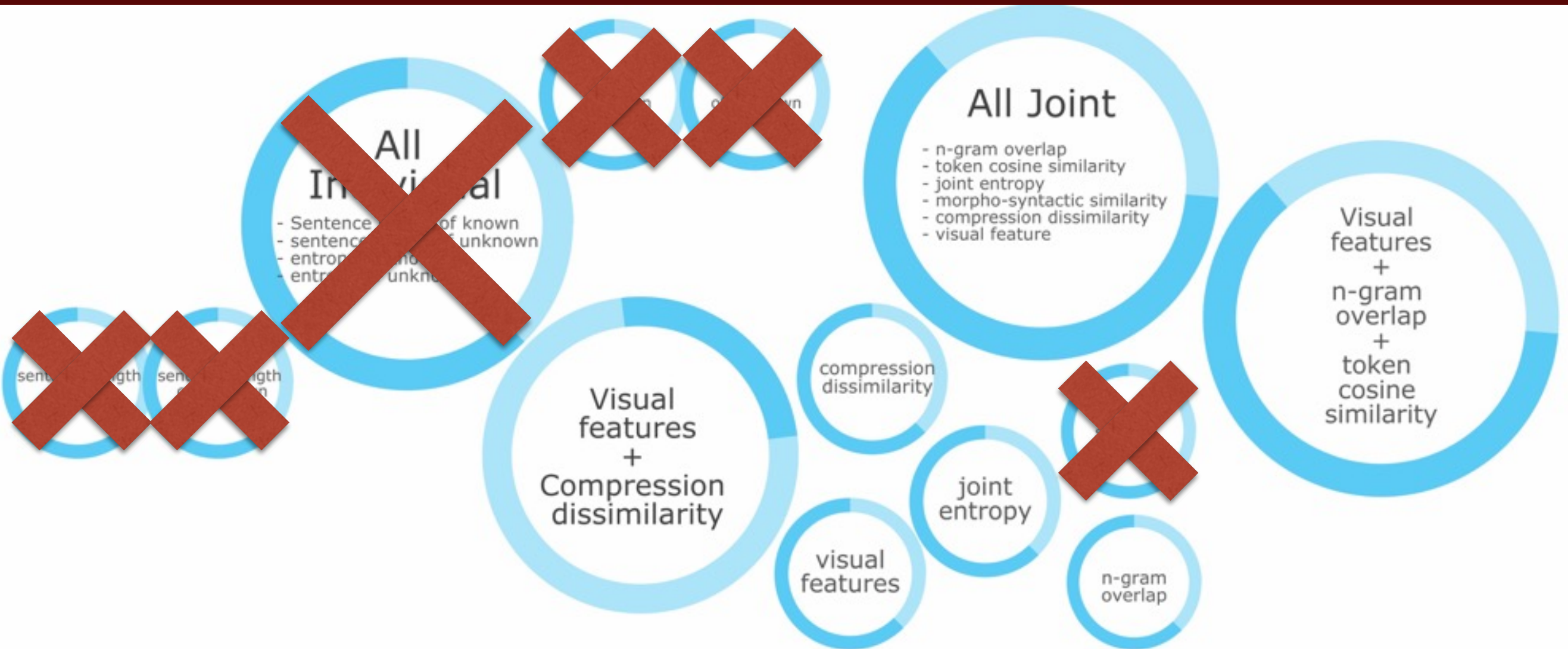
Comparing features



Results of ablation & single-feature experiments:

Differences are subtle

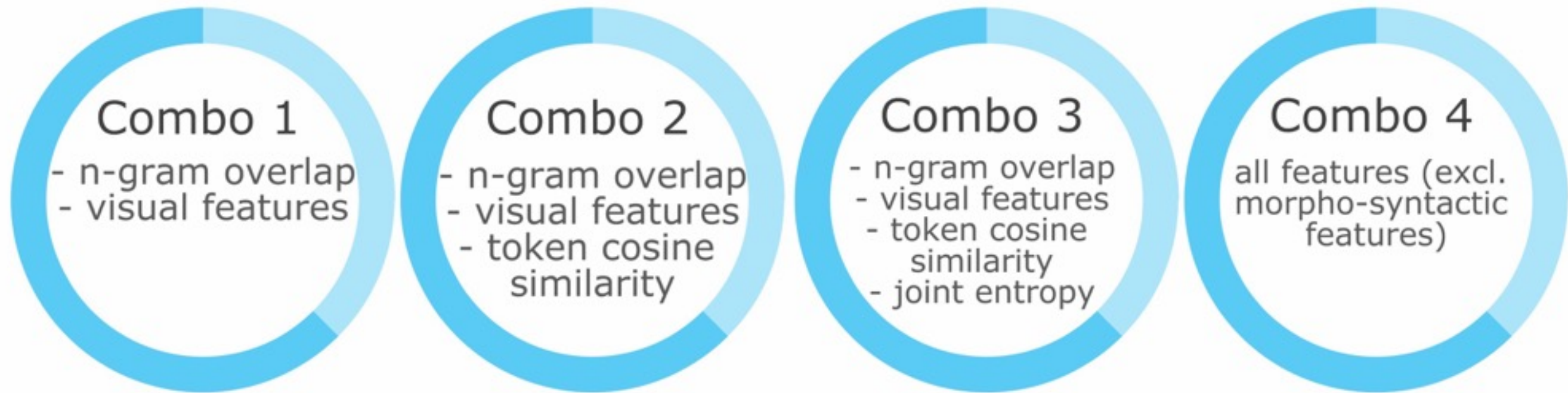
Comparing features



Results of ablation & single-feature experiments:

Differences are subtle

Resulting groups



Results

Language	Training	Test	Ranking
Dutch (full set)	.55	.62	3
English (full set)	.56	.41	7
Greek (combo2)	.54	.60	4
Spanish (full set)	.90	.54	5

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- Simple similarity features work

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- Simple similarity features work **in unison**

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- Simple similarity features work **in unison** *independent of language (except greek)*

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- Simple similarity features work **in unison** *independent of language (except greek)*
- System works fast (runtime av. 1 minute)

Final conclusion

GLAD

... is a light and fast language-independent system

... allows language adaptation done via feature selection

... involves innovative visual features which appear useful (especially for English data) and could be investigated further

