To Extend or Not to Extend? Context-driven Corpus Enrichment

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General Setting



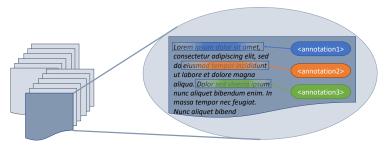
• Corpus = set of text documents

General Setting



- Corpus = set of text documents
- Each document has a set of annotations

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- Corpus = set of text documents
- Each document has a set of annotations
- Each annotation is associated with words at a specific location

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- Each document has a set of annotations
- Each annotation is associated with words at a specific location

Assumption

Annotations are relevant for a given task, i.e., reflect the context.

Formalisation of Recommending Documents



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Extend a corpus with a new document

Formalisation of Recommending Documents



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Extend a corpus with a new document only if the document

provides additional data relevant for a given task, i.e., adds value in a given context.

Formalisation of Recommending Documents



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→ Corpus enrichment

Formalisation of Recommending Documents



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Extend a corpus with a new document only if the document

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→ Corpus enrichment

Make decision based on

words, BUT: not context-specific

Formalisation of Recommending Documents



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Extend a corpus with a new document only if the document

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→ Corpus enrichment

Make decision based on

- words, BUT: not context-specific
- topics, BUT: possibly inconclusive

Formalisation of Recommending Documents



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Extend a corpus with a new document only if the document

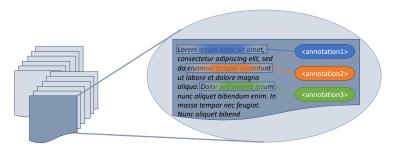
provides additional data relevant for a given task, i.e., adds value in a given context.

→ Corpus enrichment

Make decision based on

- words, BUT: not context-specific
- topics, BUT: possibly inconclusive
- annotations?

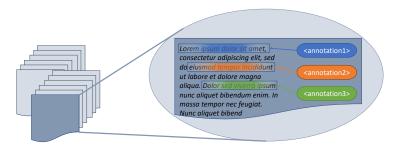
Foundation



Assumption

Annotations are relevant for a given task, i.e., reflect the context.

Foundation



Assumption

Annotations are relevant for a given task, i.e., reflect the context.

Proposition

Annotations generate the words in a document.

Foundation





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Foundation





Proposition

Annotations generate the words in a document.

Question

How much of the document can we generate with high probability given the annotations in the corpus?

How does this question help to decide "To extend or not to extend?"?





Decision

Based on answer to how much is generated with high probability: decide extension (IN/OUT)

How does this question help to decide "To extend or not to extend?"?





Decision

Based on answer to how much is generated with high probability: decide extension (IN/OUT)

• Generate large part with high probability: OUT (\rightarrow known).

How does this question help to decide "To extend or not to extend?"?





Decision

Based on answer to how much is generated with high probability: decide extension (IN/OUT)

- Generate large part with high probability: OUT (\rightarrow known).
- Probability low: OUT (→ unrelated).

How does this question help to decide "To extend or not to extend?"?





Decision

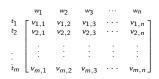
Based on answer to how much is generated with high probability: decide extension (IN/OUT)

- Generate large part with high probability: OUT (\rightarrow known).
- Probability low: OUT (→ unrelated).
- Generate only some parts with high probability: IN (\rightarrow extension).



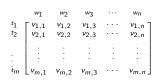


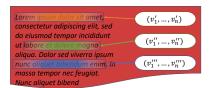
Approach



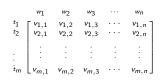


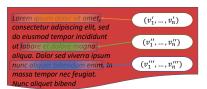
• Corpus documents (offline): build vector representation of annotations with respect to words occurring with annotations





- Corpus documents (offline): build vector representation of annotations with respect to words occurring with annotations
- New document: for word chunks, build vector representation of the words occurring in the chunk

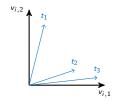


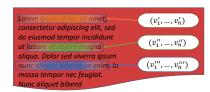


- Corpus documents (offline): build vector representation of annotations with respect to words occurring with annotations
- New document: for word chunks, build vector representation of the words occurring in the chunk
- Use cosine similarity to find annotation whose vector representation is most similar to the words of a chunk:

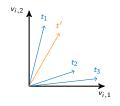
$$sim(A, B) = cos(\angle A, B) = cos(\theta) = \frac{A \cdot B}{\|A\| \cdot \|B\|}$$

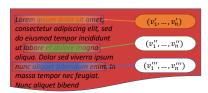
Approach



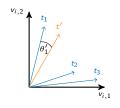


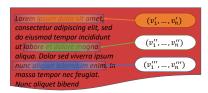
 Simplified representation of corpus annotations t_i with two words in the vocabulary



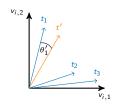


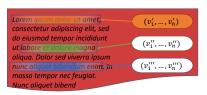
- Simplified representation of corpus annotations t_i with two words in the vocabulary
- Representation of vector representation of word chunk t'





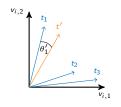
- Simplified representation of corpus annotations t_i with two words in the vocabulary
- Representation of vector representation of word chunk t'
- Angle θ'_1 between t_1 and t' smallest compared to t_2 , t_3

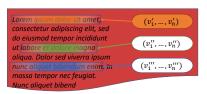




- Simplified representation of corpus annotations t_i with two words in the vocabulary
- Representation of vector representation of word chunk t'
- Angle θ'_1 between t_1 and t' smallest compared to t_2, t_3
- \rightarrow Find t_i with smallest angle for each word chunk

Approach



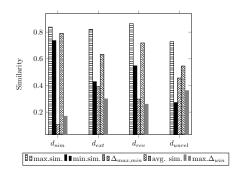


- Simplified representation of corpus annotations t_i with two words in the vocabulary
- Representation of vector representation of word chunk t'
- Angle θ'_1 between t_1 and t' smallest compared to t_2, t_3
- \rightarrow Find t_i with smallest angle for each word chunk

Use set of t_i 's for all word chunks t' in the new document and their similarities for decision

Results

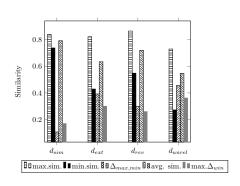
- Corpus: Wikipedia articles about cities
- New document:
 - d_{sim} : known
 - d_{ext} : extended
 - d_{rev} : revised
 - d_{unrel} : unrelated



Challenges

Influencing factors

- Corpus size
- Quality of annotations
- Indicators
- → No single indicator to rule them all!
- → Limited transfer between corpora!



Challenges

	city corpus				president corpus					
Indicator I	dsim	d_{ext}	d_{rev}	d_{unrel}	d _{sim}	d_{ext}	d_{rev}	d_{unrel}		
Max Sim.	+	+	+	0	+	+	+	0		
Min Sim.	+	0	0	_	0	0	0	_		
$\Delta_{max,min}$	_	0	_	0	_	0	_	0		
Avg. Sim.	+	0	+	0	+	+	+	0		
$Max.\Delta_{\mathit{win}}$	_	0	_	0	_	0	-	0		

"+":
$$I \ge 0.7$$
, "-": $I \le 0.3$, "0": $0.3 < I < 0.7$

Challenges

	city corpus				president corpus					
Indicator I	dsim	d_{ext}	d_{rev}	d_{unrel}	d_{sim}	d_{ext}	d_{rev}	d _{unrel}		
Max Sim.	+	+	+	0	+	+	+	0		
Min Sim.	+	0	0	_	0	0	0	_		
$\Delta_{max,min}$	_	0	_	0	_	0	_	0		
Avg. Sim.	+	0	+	0	+	+	+	0		
$Max.\Delta_{\mathit{win}}$	_	0	-	0	_	0	-	0		

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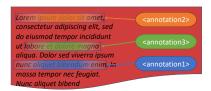
Challenges

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Indicator I	dsim	d_{ext}	d _{rev}	d _{unrel}	dsim	d_{ext}	d _{rev}	d_{unrel}	
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Avg. Sim.	+	0	+	0	+	+	+	0	
$Max.\Delta_{\mathit{win}}$	-	0	_	0	-	0	_	0	

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Subsequent Uses





- Enrich corpus with documents that add value
 → Recommendations
- Use similarities as guideline to unknown portions
- Use annotations as a starting point for annotating new document
- Augment annotations of corpus documents with new annotations of unknown document portions