

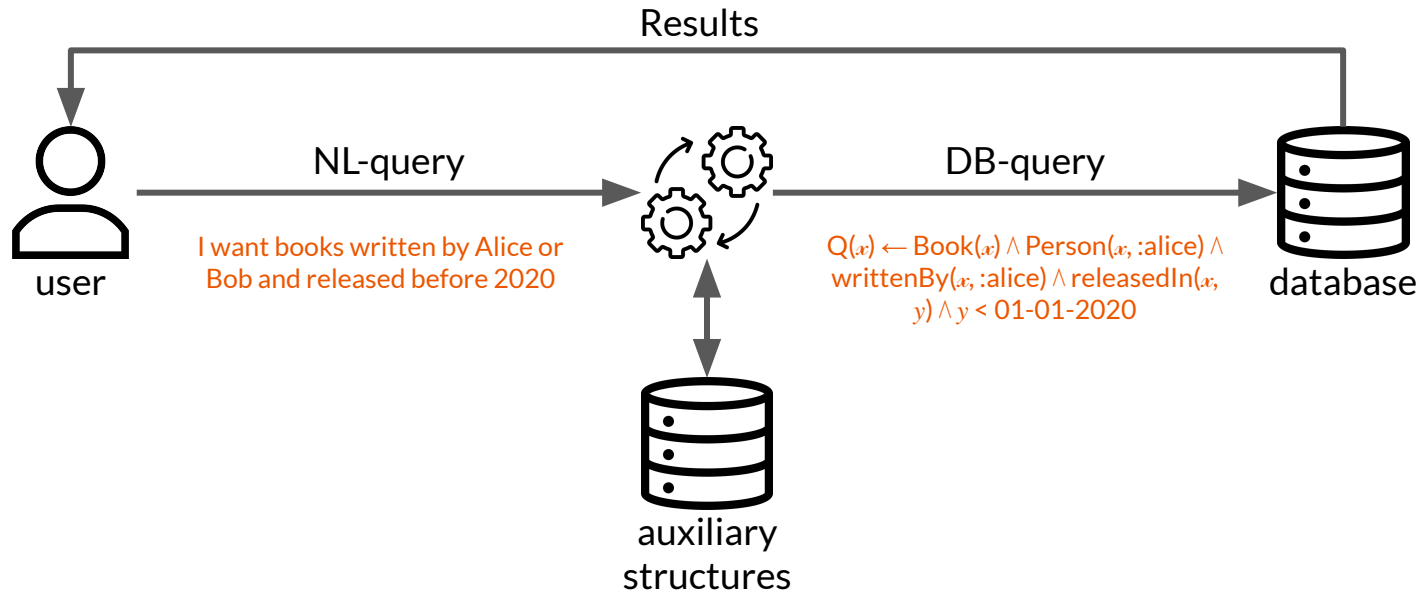


Natural Language Querying System through Entity Enrichment

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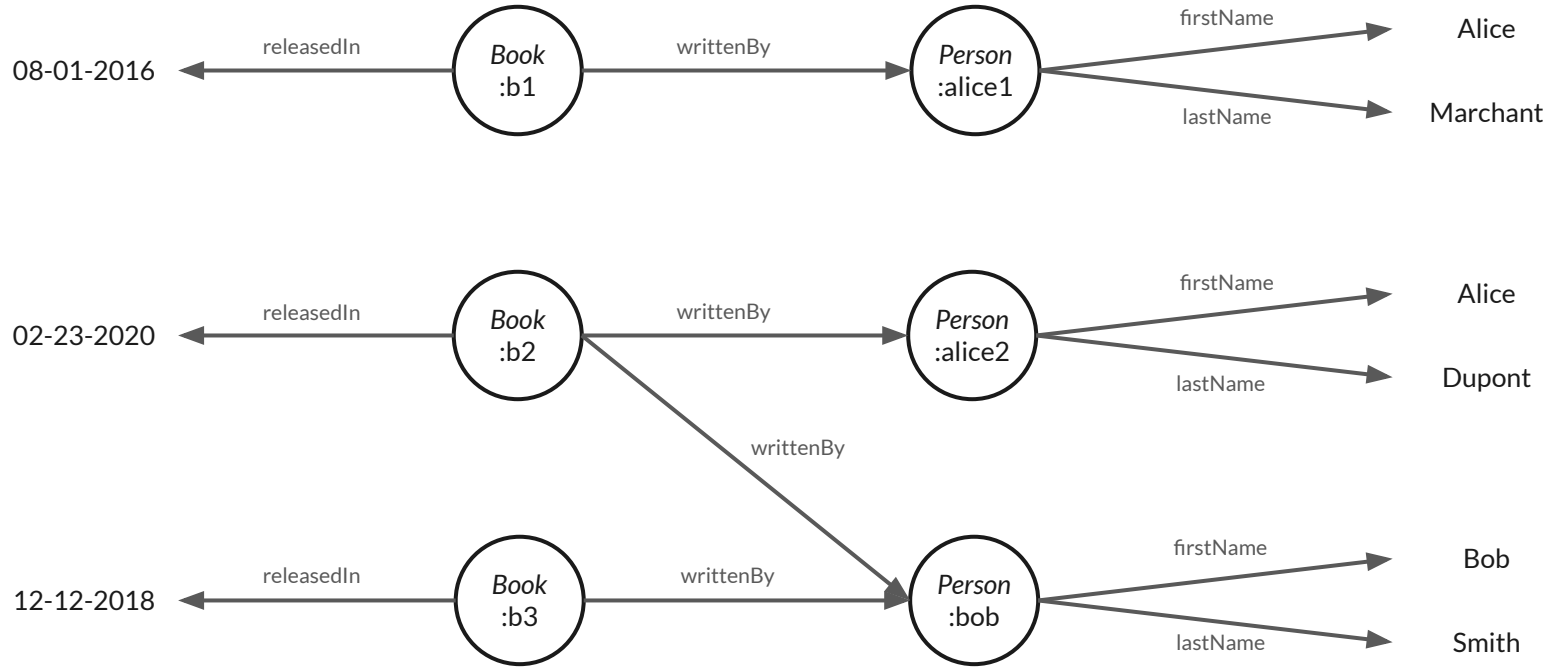
Natural Language Interface on database





Summary

- Entity Extraction and Enrichment
- Building DB-query from Enriched Entities
- Experimental Results
- Conclusion



Database example

**I want books written by Alice or
Bob and released before 2020**

I want **books** written by **Alice** or
Bob and released **before 2020**



Entity Extraction: which method?

Why we use Lexicons/Grammars approach instead of Deep Learning:

- Work well for limited knowledge
- Can be generated from the database
- Easy to update
- Use grammar for values of huge/infinite domains (dates, numbers, ...)

**How to deal with ambiguity
when instances share lexemes
in lexicons?**



Entity Extraction: how to deal with ambiguity ?

- Interaction with the user
 - Not transparent for the user
- Reduce ambiguity
 - Costly
 - Error prone
- Include the ambiguity in the query
 - Use extended entity representation to keep the value/type ambiguity

I want **books** *written by Alice* or **Bob** and *released before 2020*

Entity	Entity Value	lexT
E1	Book	Class
E2	:alice1	Person
	:alice2	Person
E3	:bob	Person

Entity	Entity Value	lexT
E4	lowerThan	Operator
E5	01-01-2020	Date
<i>E6</i>	<i>Author</i>	<i>Context</i>
<i>E7</i>	<i>ReleaseDate</i>	<i>Context</i>

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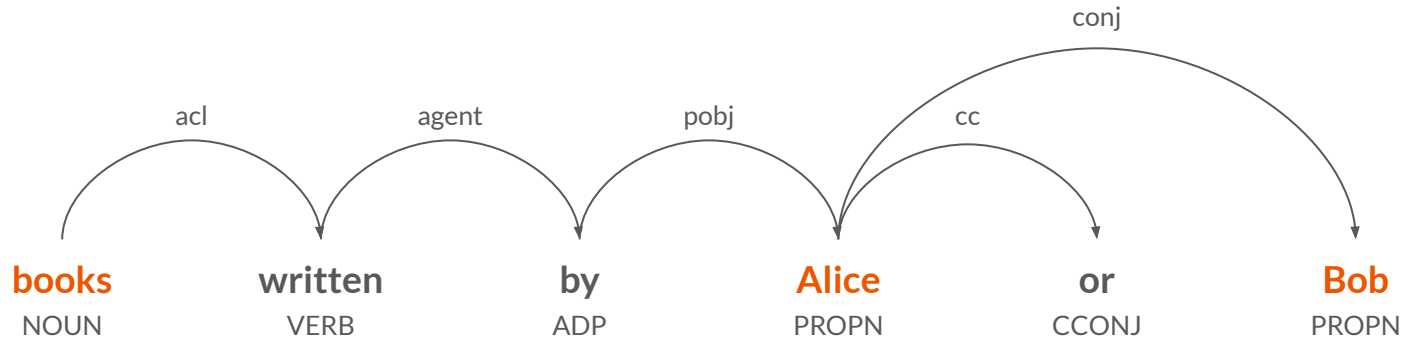
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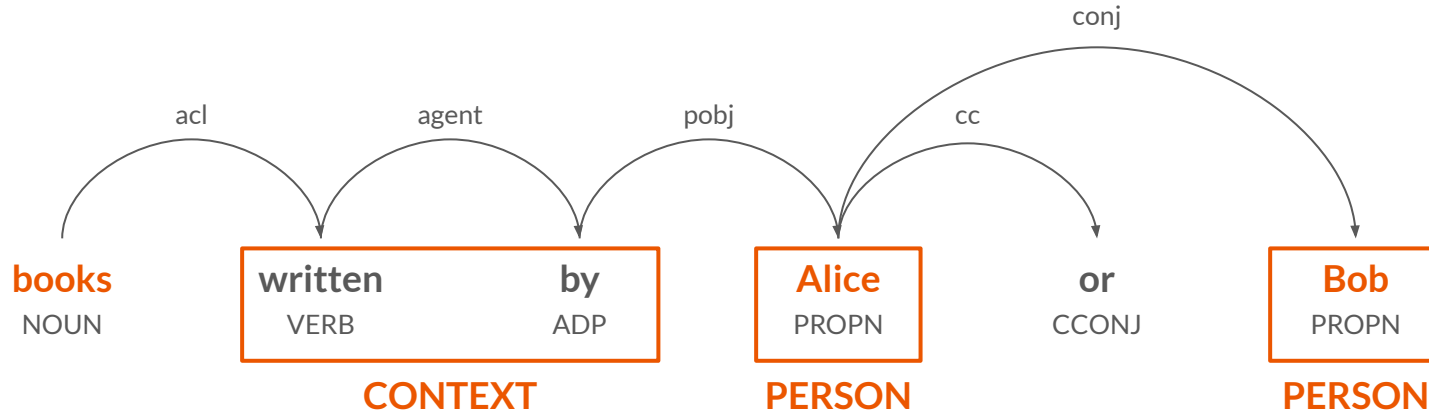
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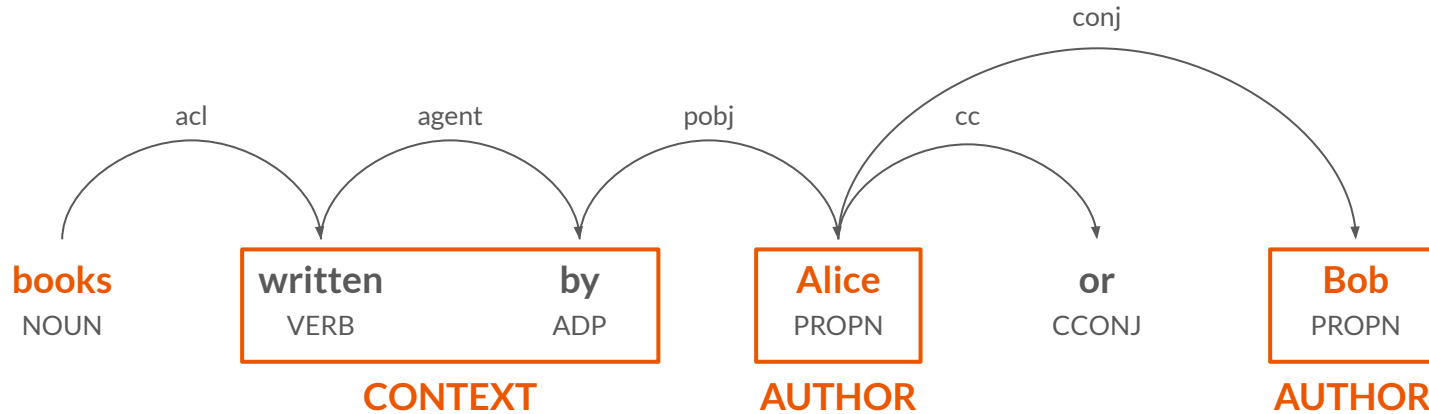
Entity Enrichment: include contexts/operators



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Entity Enrichment: include contexts/operators



I want **books** written by **Alice** or **Bob** and released **before 2020**

Entity	Entity Value	dbT	op
Ee1	Book	Class	=
Ee3	:bob	Person	=
	:bob	Author	=
Ee4	01-01-2020	ReleaseDate	<

Entity	Entity Value	dbT	op
Ee2	:alice1	Person	=
	:alice2	Person	=
	:alice1	Author	=
	:alice2	Author	=

**How to deal with coordinating
conjunctions? (“Alice *or* Bob”)**



Entity Enrichment: conjunction management

- *AND* simply give different information so no need for treatment
- *OR* add ambiguity and need to be computed
 - Ambiguity on the value

I want **books** written by **Alice** or **Bob** and released **before 2020**

Entity	Entity Value	dbT	op
Ee1	Book	Class	=
Ee2	:alice1	Person	=
	:alice2	Person	=
	:bob	Person	=

Entity	Entity Value	dbT	op
Ee2	:alice1	Author	=
	:alice2	Author	=
	:bob	Author	=
Ee4	01-01-2020	ReleaseDate	<



Build the DB-query

Entity	Entity Value	dbT
Ee2	:alice1	Person
		Author
	:alice2	Person
		Author

- Direct translation using a mapping between dbT and predicates
- We have an ambiguity (multiple query) when an entity contains multiple values

I want **books** written by **Alice** or **Bob** and released **before 2020**

$Q1(x) \leftarrow \text{Book}(x)$
 $\wedge \text{Person}(x, \text{:alice1})$
 $\wedge \text{writtenBy}(x, \text{:alice1})$
 $\wedge \text{releasedIn}(x, y)$
 $\wedge y < \text{01-01-2020}$

$Q2(x) \leftarrow \text{Book}(x)$
 $\wedge \text{Person}(x, \text{:alice2})$
 $\wedge \text{writtenBy}(x, \text{:alice2})$
 $\wedge \text{releasedIn}(x, y)$
 $\wedge y < \text{01-01-2020}$

$Q3(x) \leftarrow \text{Book}(x)$
 $\wedge \text{Person}(x, \text{:bob})$
 $\wedge \text{writtenBy}(x, \text{:bob})$
 $\wedge \text{releasedIn}(x, y)$
 $\wedge y < \text{01-01-2020}$



Experimentals Results

Database:

- 66 classes
- 29327 instances

Dataset:

- 113 NL-queries
- 10 classes used

dbT	precision	recall	f1-score	support
Class	1.00	0.62	0.77	82
ApplicationDate	1.00	0.62	0.76	13
ArchiveDate	0.50	0.67	0.57	3
CreationDate	0.60	0.60	0.60	5
ExpirationDate	1.00	0.50	0.67	2
Customers	1.00	0.60	0.75	5
Department	1.00	0.11	0.20	9
Sector	1.00	0.95	0.98	21
Author	0.77	0.63	0.70	38
Signatory	0.90	0.86	0.88	21
Status	1.00	0.50	0.67	6
Unit	1.00	0.27	0.43	11
...
Weighted avg.	0.86	0.59	0.67	295



Conclusion

- Information Extraction based approach: database/query language independent
- Domain independent approach
- Good precision (>80%) thanks to lexicons
- Weaker recall (<60%) because of evaluating method (ambiguity)
- Current work: evaluation method for ambiguity and deep learning entities enrichment