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# Latin Embeddings and the LiLa Knowledge Base of Interlinked Resources for Latin

Marco Passarotti, Rachele Sprugnoli

Computational Approaches to ancient Greek and Latin  
Groningen, 2 November 2021



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# Overview



## The LiLa Knowledge Base

### Latin Word Embeddings

Training and Testing Lemma Embeddings

Diachronic Analysis

Induction of Sentiment Lexicons

Modeling

### Conclusions



## ERC Consolidator Grant 2018-2023

A collection of multifarious, interoperable linguistic resources described with the same vocabulary for knowledge description (by using common data categories and ontologies)

### Interlinking as a Form of Interaction



Common Language Resources and  
Technology Infrastructure

Infrastructure



Interoperability

# Why LiLa?

State of Affairs



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- ▶ Resources disconnected from each other (silos of LRs)

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- ▶ Proprietary and heterogeneous formats
- ▶ Different representation schemes, query languages, annotation criteria and tagsets

# The Linked Data Principles

...just to be FAIR



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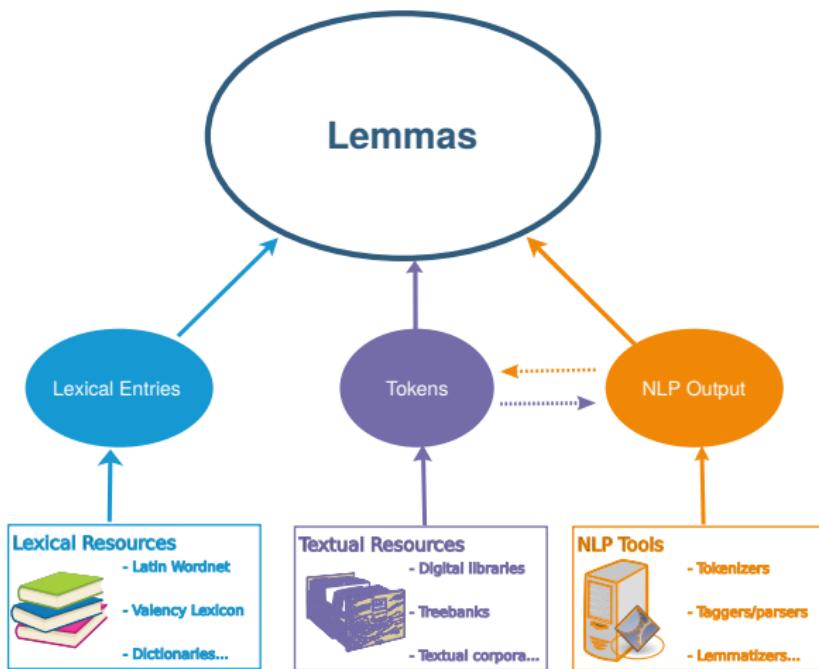
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- ▶ Include links to other URIs

# LiLa Knowledge Base

Lexically-based architecture and (meta)data sources

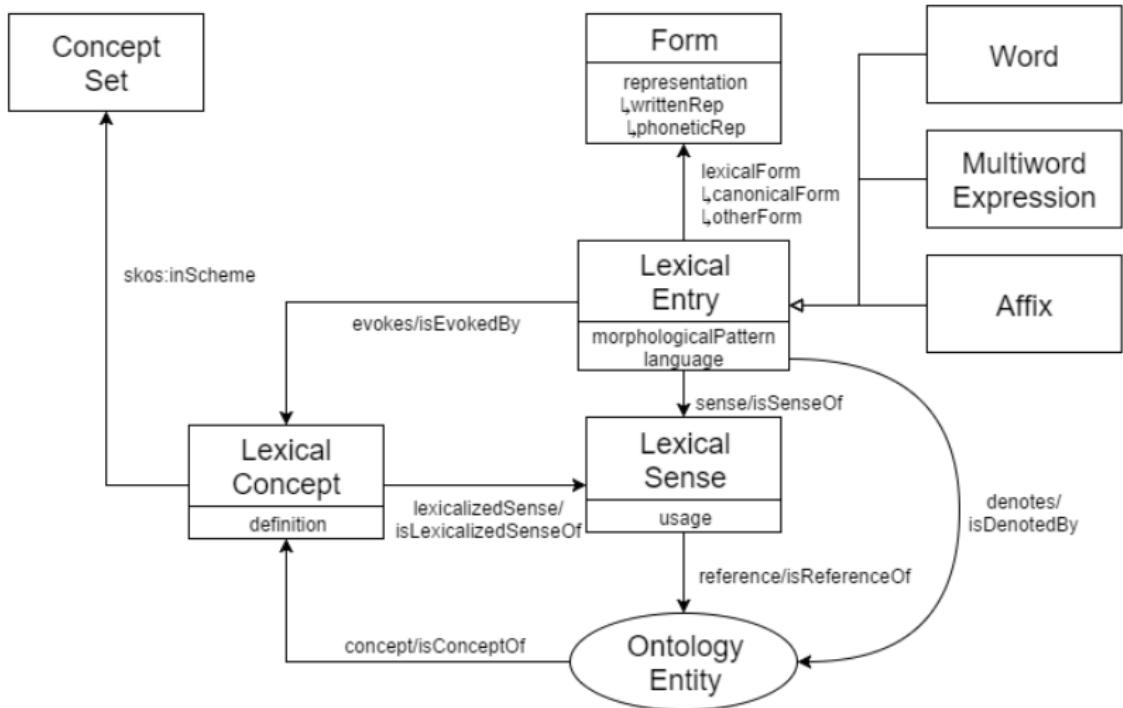


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# LiLa and Ontolex Lemon

A *de facto* W3C standard for publishing lexical data as LLOD



# Lemma Bank and Lexical Resources

The core of LiLa



Lemma *admiror* ‘to admire, to respect’

<http://lila-erc.eu/data/id/lemma/87541>

- ▶ Lemma Bank
- ▶ A derivational lexicon (Word Formation Latin)
- ▶ A polarity lexicon (LatinAffectus)
- ▶ An etymological dictionary (De Vaan)
- ▶ A Valency Lexicon (Latin Vallex)
- ▶ A manually checked subset of the Latin WordNet
- ▶ A bilingual Latin-English dictionary (Lewis & Short)

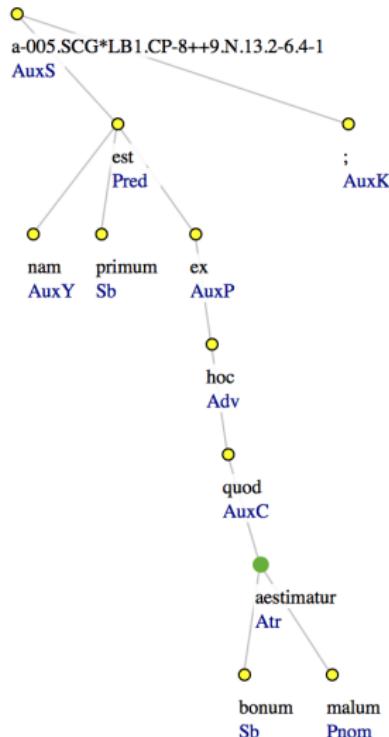
# Textual Resources

Source: the *Index Thomisticus* Treebank (original scheme)



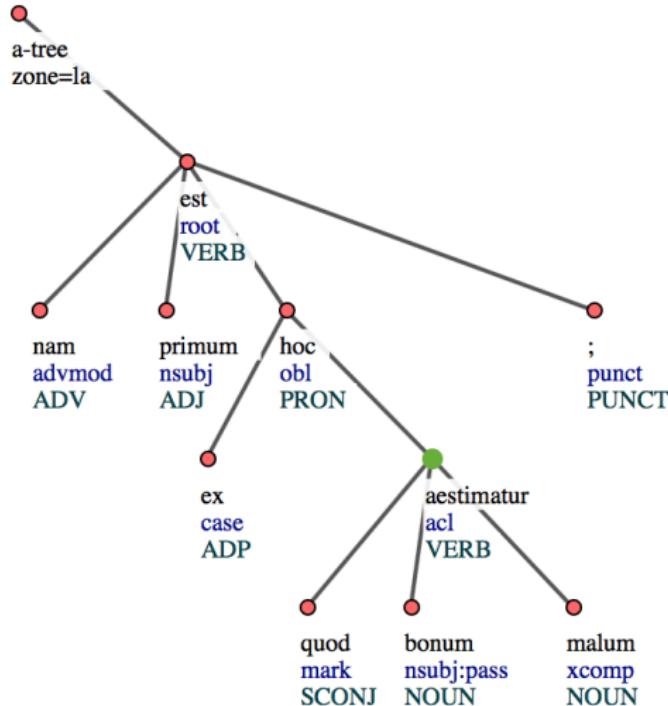
*nam primum est ex hoc  
quod bonum **aestimatur**  
malum; (IT-TB: SCG, lib. 1,  
cap. 89, n. 13)*

*for the first arises because  
the good **is judged** to be  
evil; (Trans. Anton C. Pegis)*



# Textual Resources

Source: the *Index Thomisticus* Treebank (UD scheme)



# Textual Resources

A token from the *Index Thomisticus* Treebank in LiLa



Token *aestimatur*

[http://lila-erc.eu/lodview/data/corpora/  
ITTB/id/token/005.SCG\\*LB1.CP-8++9.N.13.](http://lila-erc.eu/lodview/data/corpora/ITTB/id/token/005.SCG*LB1.CP-8++9.N.13.)

2-6.4-1W8

## ► Textual Resources

- Index Thomisticus Treebank (*Summa contra Gentiles*): ca. 450,000 nodes
- Dante Search (700th death anniversary): ca. 46,000 tokens
- Liber Abaci, Chapter VIII*: ca. 30,000 tokens
- Querulus sive Aulularia*: ca. 17,000 tokens
- PROIEL and LLCT treebanks
- Computational Historical Semantics, LASLA and CroALa Corpora

## ► Lexical Resources

- Word Formation Latin: ca. 46,000 lemmas (Classical Latin)
- Etymological dictionary of Latin & the other Italic Langs.: ca. 1,400 entries
- LatinAffectus: ca. 4,000 entries
- Index Graecorum Vocabulorum in Linguam Latinam: ca. 1,800 entries
- Latin WordNet: ca. 1,000 manually checked entries
- Latin Vallex 2.0: Valency Lexicon
- Lewis & Short Dictionary
- Lemma Embeddings

## ► NLP tools

- LEMLAT (lemma bank): ca. 150,000 lemmas

## ► TOTAL: approximately 15 million triples

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# Lemma Embeddings of Classical Latin: Why



1. Supporting data-driven **socio-cultural studies** of the Latin world
2. Fostering the **interdisciplinary collaboration** between Computational Linguistics and Classical Studies
3. **Filling a void** in the literature:

	Word2Vec	FastText	Clean	Download	Evaluation
CoNLL	✓			✓	
Facebook		✓		✓	
Bamman	✓			✓	
CompHistSem	✓	✓	✓	✓	
LiLa	✓	✓	✓	✓	✓

Sprugnoli, R., Passarotti, M., & Moretti, G. (2019). Vir is to Moderatus as Mulier is to Intemperans - Lemma Embeddings for Latin. In Proceedings of the Sixth Italian Conference of Computational Linguistics (CLiC-it 2019).

# Lemma Embeddings of Classical Latin



Texts taken from the **LASLA corpus**:

- ▶ manually annotated since 1961
- ▶ lemmas, PoS tags, inflectional features
- ▶ multi-genre
- ▶ 158 texts, 20 authors
- ▶ 1.7M words

Text pre-processing:

- ▶ conversion to CoNLL-U
- ▶ extraction of lemmas
- ▶ lower-casing
- ▶ conversion: v → u

## Vector representations:

- ▶ Word2vec: treats each word in corpus like an atomic entity
- ▶ FastText: treats each word as composed of character ngrams

## Models:

- ▶ Skip-gram: the distributed representation of the input word is used to predict the context
- ▶ CBOW: the distributed representations of context (or surrounding words) are combined to predict the word in the middle

## Dimensions:

- ▶ 100
- ▶ 300

# Lemma Embeddings: Intrinsic Evaluation



- ▶ **Synonym Selection Task:** select the correct synonym of a target lemma out of a set of possible answers
- ▶ **Benchmark:** 2,759 multiple-choice questions each involving 5 terms
  - 1 target lemma
  - 1 synonym of the target lemma taken from Latin synonym dictionaries
  - 3 decoy lemmas

TARGET WORDS	SYNONYM	DECAY WORDS		
<i>exilis/thin</i>	<i>macer/emaciated</i>	<i>moles/pile</i>	<i>mortalitas/mortality</i>	<i>audens/daring</i>
<i>globus/ball</i>	<i>sphaera/sphere</i>	<i>patronus/defender</i>	<i>brevitas/brevity</i>	<i>apex/cap</i>
<i>cunctus/doubt</i>	<i>haesito/hesitate</i>	<i>uito/avoid</i>	<i>conflo/compose</i>	<i>pondero/weigh</i>

# Lemma Embeddings: Intrinsic Evaluation



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## ► Results:

- calculate the cosine similarity between the vector of the target lemma and that of the other lemmas
- pick the candidate with the largest cosine
- measure the correct-answer accuracy

	word2vec		fastText	
	cbow	skip-gram	cbow	skip-gram
100	81.14%	79.83%	80.57%	<b>86.91%</b>
300	80.86%	79.48%	79.43%	86.40%

## ► Errors:

- meronymy: TARGET: *annalis* - SYN: *historia* - ANSWER: *charta*
- morphological derivation: TARGET: *consors* - SYN: *particeps* - ANSWER: *sors*

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# Lemma Embeddings: Diachronic Analysis



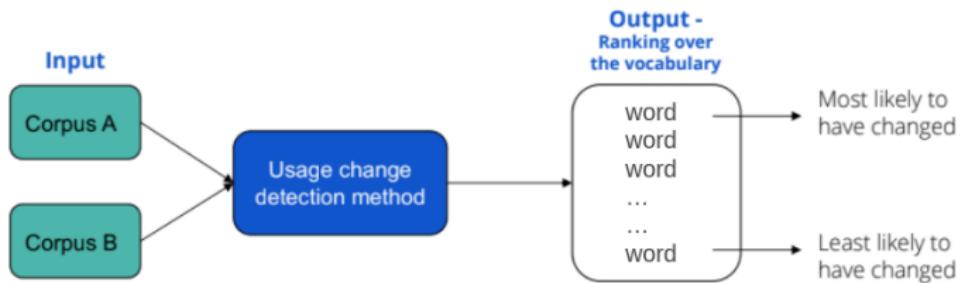
- ▶ The use of Latin spans more than two millennia
  - Classical Latin ≠ Medieval Latin
- ▶ New embeddings to compare:
  - *Opera Maiora* of Thomas Aquinas:
  - philosophical and religious works
  - 13th century
  - manually lemmatized in the *Index Thomisticus* project
  - 4.5 million words

Sprugnoli, R., Passarotti, M., & Moretti, G. Building and Comparing Lemma Embeddings for Latin. Classical Latin versus Thomas Aquinas.IJCoL. Italian Journal of Computational Linguistics, 6(6-1), 29-45.

# Lemma Embeddings: Diachronic Analysis



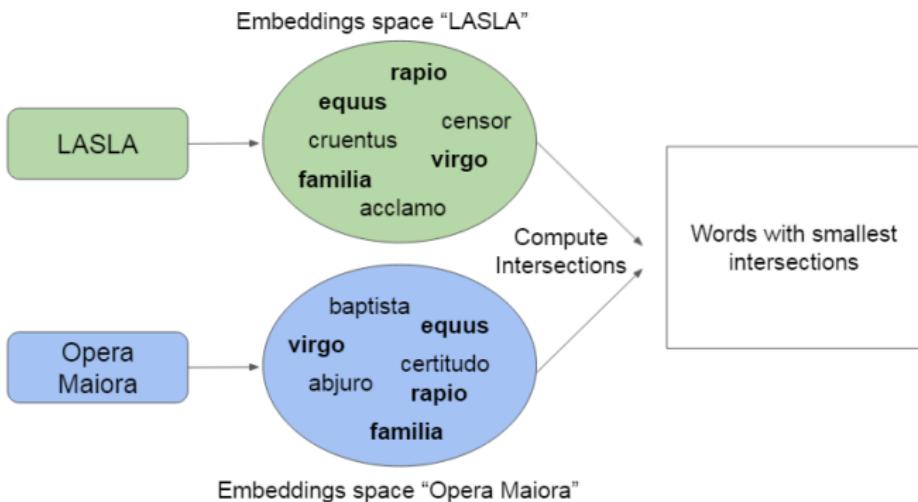
- ▶ USAGE CHANGE: identify words that are used differently over time periods



Gonen, H., Jawahar, G., Seddah, D., & Goldberg, Y. (2020). Simple, interpretable and stable method for detecting words with usage change across corpora. In Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics (pp. 538-555).

# Lemma Embeddings: Diachronic Analysis

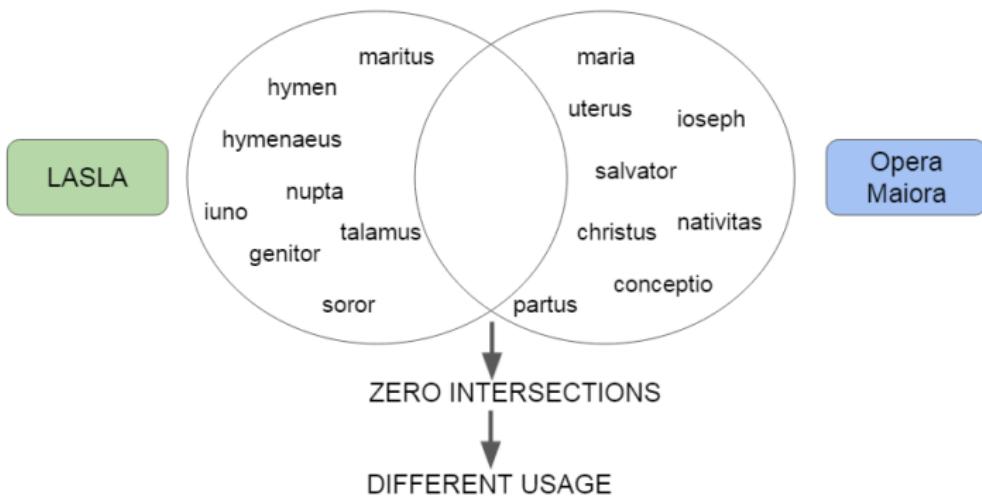
- ▶ Nearest neighbors are a proxy for meaning



# Lemma Embeddings: Diachronic Analysis



- ▶ Example: *virgo* = girl of marriageable age, virgin



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# Lemma Embeddings: Sentiment Lexicon Induction



«Sentiment analysis and opinion mining is the field of study that analyzes people's opinions, sentiments, evaluations, attitudes, and emotions from written language.»

*Bing Liu, "Sentiment Analysis and Opinion Mining" Morgan & Claypool Publishers, 2012*

- ▶ Development of sentiment lexicons for Latin = list of words associated to scores expressing their prior polarity:
  - essential resource for both machine learning and lexicon-based sentiment analysis systems
  - set of lexicons created manually or automatically

# Lemma Embeddings: Sentiment Lexicon Induction

- ▶ Automatic induction from word embeddings starting from a list of seed terms with known sentiment score
  - **seed terms:** 200 most frequent adjs and nouns from the LASLA corpus
  - **embeddings:** pre-trained with word2vec on LASLA corpus with a LEMMA\_PoS representation, e.g. *rosa\_noun*, *amo\_verb*
  - **algorithm:** <https://github.com/WladimirSidorenko/SentiLex>
- ▶ **Output:** lexicon of 1,030 lemmas with three-value scores

Lemma	PoS	Sentiment
<i>miseria</i> ‘misery’	noun	negative
<i>cruciatus</i> ‘torture’	noun	negative
<i>optabilis</i> ‘desiderable’	adj	positive
<i>benevolentia</i> ‘good-will’	noun	positive
<i>aerumna</i> ‘trouble’	noun	negative

Sprugnoli, R., Passarotti, M., Corbetta, D., & Peverelli, A. (2020, May). Odi et Amo. Creating, Evaluating and Extending Sentiment Lexicons for Latin. In Proceedings of the 12th Language Resources and Evaluation Conference (pp. 3078-3086).

# Lemma Embeddings: Sentiment Analysis



Evaluation with respect to a manually annotated gold standard made of 1,144 lemmas:

- ▶ calculation of the **accuracy**
- ▶ comparison with the results obtained by creating a sentiment lexicon with a translation method

	TRANSLATION	INDUCTION
ADJ	64.9%	86.7%
NOUN	66.8%	62.5%
MICRO-AVG	66.1%	74.4%

# Lemma Embeddings: Sentiment Analysis



Possibility to generate **time-specific** sentiment lexicons

- ▶ Induction using embeddings trained on the *Computational Historical Semantics* corpus
  - 904.400 lemmas
  - Latin documentary texts
  - between the 2nd and the 15th century AD
- ▶ 71% of the entries in the induced lexicon were different from the ones obtained on the LASLA corpus

LEMMA	PoS	SENTIMENT
<i>archangelus</i> 'archangel'	noun	positive
<i>peccatrix</i> 'female sinner'	noun	negative
<i>criminosus</i> 'guilty'	adj	negative
<i>poenalis</i> 'penal'	adj	negative

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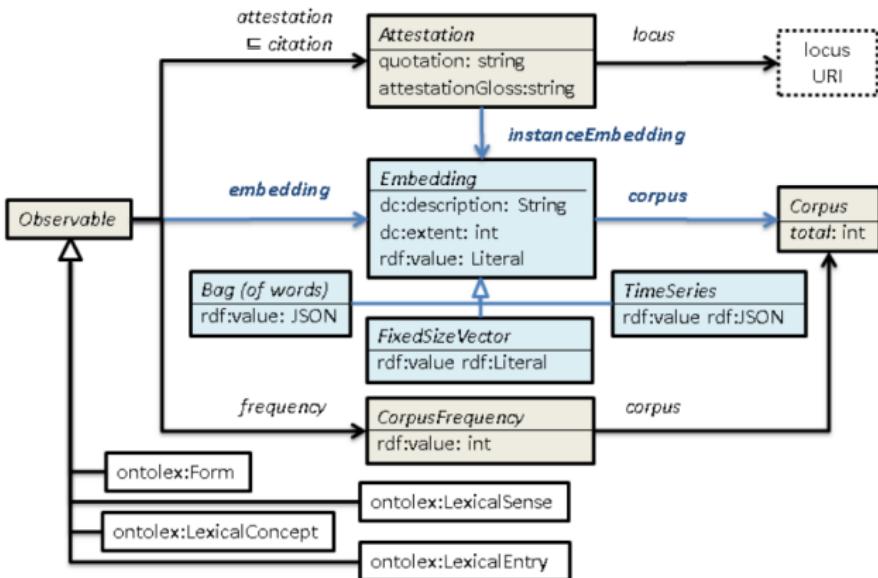
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# Frequency Attestation and Corpus (FrAC) module

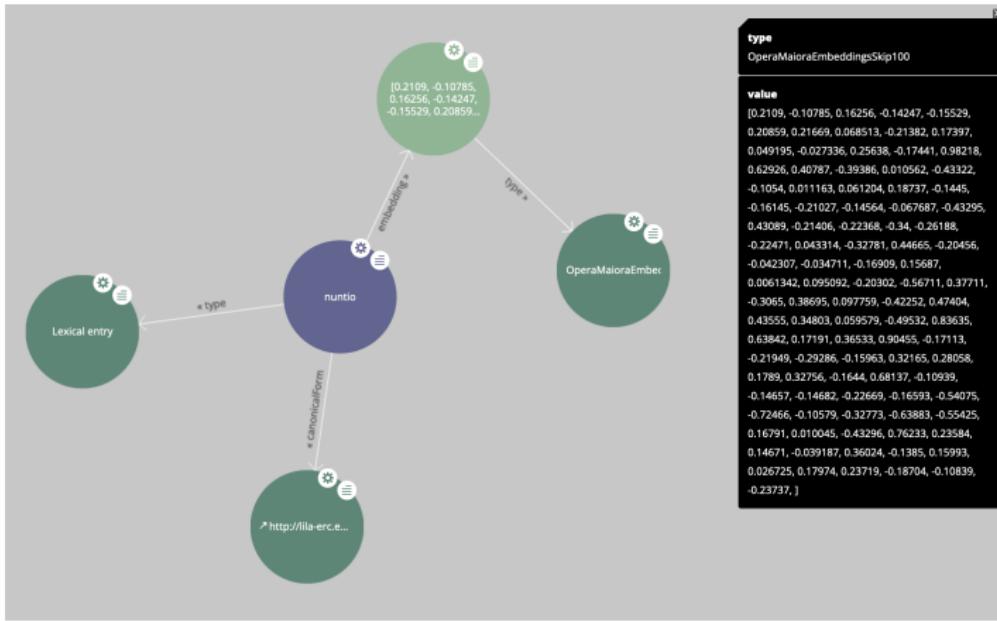


Chiarcos, C., Declerck, T., & Ionov, M. (2021). Embeddings for the Lexicon: Modelling and Representation. In Proceedings of the 6th Workshop on Semantic Deep Learning (SemDeep-6) (pp. 13-19).

# Frequency Attestation and Corpus (FrAC) module



- ▶ Example: lemma embeddings of *nuntio* ‘to announce’ pre-trained on *Opera Maiora*, skip-gram model, 100 dimensions



# Open Challenges

It is not so easy ...but we are doing our best



## To fully exploit digital texts

texts "in a form that both humans and machines can use,  
preferably in a way that leverages the unique capabilities of both"

(S. Huskey, SunoikisisDC SS 2021 - Session 12)

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- ▶ Models still missing for several types of (meta)data: e.g. for critical editions
- ▶ Community-based effort: persuading resource developers to adopt LOD practices and reaching consensus around shared vocabularies, ontologies, data categories etc.

# Thanks!

Get in touch



## LiLa: Linking Latin

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CIRCSE Research Centre

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🌐 <https://github.com/CIRCSE>

🌐 <https://lila-erc.eu>

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📍 Largo Gemelli 1, 20123 Milan, Italy



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