

#### Alexander Panchenko

FROM UNSUPERVISED INDUCTION OF LINGUISTIC STRUCTURES FROM TEXT TOWARDS APPLICATIONS IN DEEP LEARNING



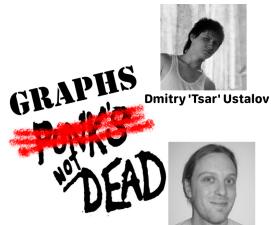
#### In close collaboration with ...



**Chris Biemann** 



Stefano Faralli



Simone P. Ponzetto



#### In collaboration with ...

- Andrei Kutuzov
- Eugen Ruppert
- Fide Marten
- Nikolay Arefyev
- Steffen Remus
- Martin Riedl
- Hubert Naets
- Maria Pelevina
- Anastasiya Lopukhina
- Konstantin Lopukhin







### Levels of Linguistic Analysis

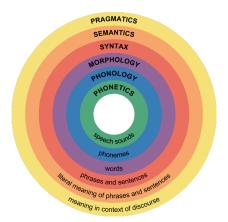


Image source: https://commons.wikimedia.org/wiki/File:Major\_levels\_of\_linguistic\_structure.svg



#### Motivation 0 • 0 0 0 0 0 0 0

### Levels of Linguistic Analysis

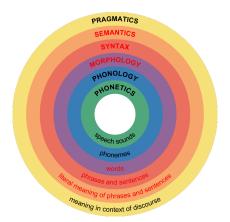
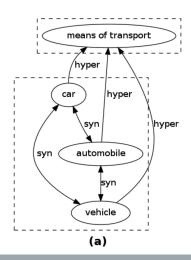


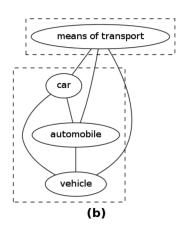
Image source: https://commons.wikimedia.org/wiki/File:Major\_levels\_of\_linguistic\_structure.svg

## Linguistic Structures and Graphs

- (Written) language is a symbolic system
- Semantic level: typed weighted graphs of concepts
  - Co-occurrence networks
  - Lexical databases, e.g. WordNet
  - Thesauri, e.g. NLM
  - Ontologies, e.g. DBPedia
  - Associative networks, e.g. Edinburgh Associative Thesaurus
  - ...

## **Semantic Graphs**

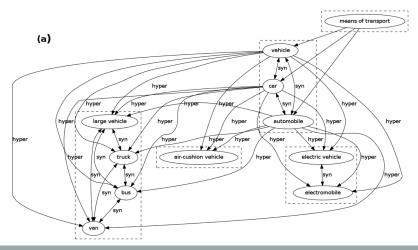






#### Motivation

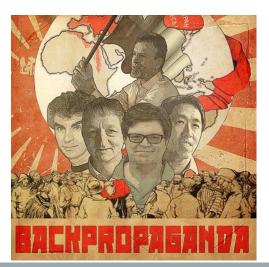
### **Semantic Graphs**







## The new brave world of Deep Learning



- "Anti-connectivism"
- End-to-end learning: symbolic representations aren't needed



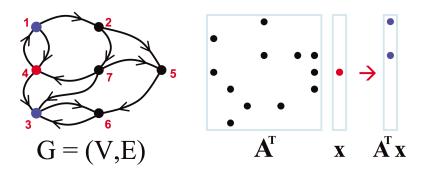


# The new brave world of Deep Learning

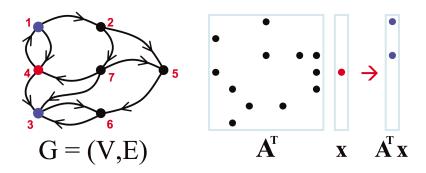


- "Anti-connectivism"
- End-to-end learning: symbolic representations aren't needed
- Word embeddings lookup (at most)

## **Graph Matrix Duality**

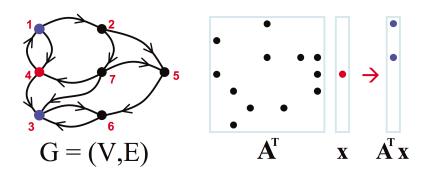


## **Graph Matrix Duality**



lacksquare Adjacency matrix f A is dual with the corresponding graph G.

## **Graph Matrix Duality**



- lacksquare Adjacency matrix **A** is dual with the corresponding graph G.
- Vector matrix multiply  $\mathbf{A}^T\mathbf{x}$  is dual with breadth-first search.



Motivation 0000000

### Goal: Linguistic Structures in DL

Learn interpretable symbolic structures from text in an unsupervised way, which are more complex than words.



Motivation 000000

## Goal: Linguistic Structures in DL

- Learn interpretable symbolic structures from text in an unsupervised way, which are more complex than words.
- **Represent the learned structures** in a vector space.



Motivation ○○○○○○

#### Goal: Linguistic Structures in DL

- Learn interpretable symbolic structures from text in an unsupervised way, which are more complex than words.
- Represent the learned structures in a vector space.
- Use the vector representations instead/in addition to word embedding the deep learning applications. Lookup of word senses, frames, etc.



Motivation ○○○○○○

#### Goal: Linguistic Structures in DL

- Learn interpretable symbolic structures from text in an unsupervised way, which are more complex than words.
- **Represent the learned structures** in a vector space.
- Use the vector representations instead/in addition to word embedding the deep learning applications. Lookup of word senses, frames, etc.
- 4 More complex structures could improve performance, but also provide better interpretability of the deep learning models.