



# ArabGlossBERT

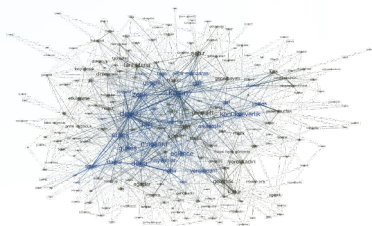
## Fine-Tuning BERT on Context-Gloss Pairs for Word Sense Disambiguation

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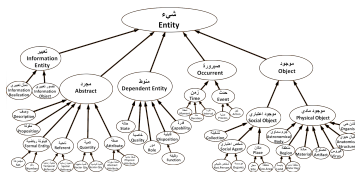
# Lexical Resources at Birzeit University

## Lexicographic Database



150 lexicons  
Largest Arabic-multilingual  
database

## Arabic Ontology



Formal Arabic Wordnet  
with ontologically clean  
content

## Dialect Corpora



Annotated corpora  
each word is annotated  
with many morph features

## Big Linguistic Data Graph

<https://ontology.birzeit.edu>

## ArabGlossBERT: Fine-Tuning BERT on Context-Gloss Pairs for WSD

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### Abstract

Using pre-trained transformer models such as BERT has proven to be effective in many NLP tasks. This paper presents our work to fine-tune BERT models for Arabic Word Sense Disambiguation (WSD). We treated the WSD task as a sentence-pair binary classification task. First, we constructed a dataset of labeled Arabic context-gloss pairs (~167k pairs) we extracted from the Arabic Ontology and the large lexicographic database available at Birzeit University. Each pair was labeled as *True* or *False* and target words in each context were identified and annotated. Second, we used this dataset for fine-tuning three pre-trained Arabic BERT models. Third, we experimented the use of different supervised signals used to emphasize target words in context. Our experiments achieved promising results (accuracy of 84%) although we used a large set of senses in the experiment.

### 1 Introduction

Word Sense Disambiguation (WSD) aims to determine which sense (*i.e.* meaning) a word may denote in a given context. This is a challenging

social media mining, Named-Entity Recognition, word sense disambiguation, topic classification and summarization, among others.

A gloss is a short dictionary definition describing one sense of a lemma or lexical entry (Jarrar, 2006, 2005). A context is an example sentence in which the lemma or one of its inflections (*i.e.* the target word) appears. In this paper, we aim to fine-tune Arabic models for Arabic WSD. Given a target word in a context and a set of glosses, we will fine-tune BERT models to decide which gloss is the correct sense of the target word. To do that, we converted the WSD task into a BERT sentence-pair binary classification task similar to (Huang et al., 2019; Yap et al., 2020; Blevins and Zettlemoyer, 2020). Thus, BERT is fine-tuned on a set of context-gloss pairs, where each pair is labeled as *True* or *False* to specify whether or not the gloss is the sense of the target word. In this way, the WSD task is converted into a sentence-pair classification task.

One of the main challenges for fine-tuning BERT for Arabic WSD is that Arabic is a low-resourced language and that there are no proper labeled context-gloss datasets available.

To overcome this challenge, we collected a rela-

## The Word Sense Disambiguation (WSD) Task

Given a word in a context, which sense (i.e. meaning) this word denotes?

قصيدة من عيون الشعر

### Set of senses

1. عُضُو الإِبْصَارِ فِي الْإِنْسَانِ وَالْحَيَوَانَ: لَهُ عَيْنَانِ كَعَيْنَيْ الصَّقْرِ - أَلَا إِنَّمَا الْعَيْنَانِ لِلْقَلْبِ رَائِدٌ ...
2. جَاسُوسٌ، "كَانَ عَيْنًا لِدَوْلَةٍ أجنبيَّةٍ . بَثَّ الْعَيُونَ : تَجَسَّسٌ، رَاقِبٌ - فَلَانٌ عَيْنٌ عَلَى فَلَانٍ : نَاطِرٌ عَلَيْهِ
3. أَجُودٌ كُلُّ شَيْءٍ وَأَحْسَنُهُ وَنَفِيسُهُ: عَيُونُ الْفَنِّ.
4. حَارِسٌ: فَلَانٌ عَيْنٌ عَلَى الْمَكَانِ.
5. الْحَاضِرُ مِنْ كُلِّ شَيْءٍ أَصْبَحَ أَثْرًا بَعْدَ عَيْنٍ ...
6. عَيْنُ الْمَاءِ:- يَنْبُوعُهُ، تُحَلِّقُ الطَّيُورُ فَوْقَ عَيُونِ الْمَاءِ
7. عَيْنُ الشَّيْءِ:- نَفْسُهُ، ذَاتُهُ (تَسْتَعْمَلُ لِلتَّوَكِيدِ): جَاءَ الْقَوْمُ أَعْيُنَهُمْ - كُنَّا فِي الْمَكَانِ عَيْنَهُ.
8. عَيْنُ الْعَقْلِ:- قُدْرَةُ ذَهْنِيَّةٍ مَوْرُوثَةٌ عَلَى التَّخْيِيلِ وَتَذَكُّرِ الْأَحْدَاثِ.
9. ....

WSD has been a challenging task for many years but has gained recent attention due to the advances in contextualized word embedding models such as BERT.

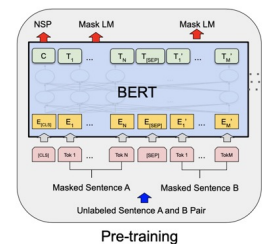
## ❖ Arabic context-gloss pairs Dataset (167k)

- Extracted from Birzeit University’s Lexicographic database
- Annotated target words in context;

| Gloss   | Context                         | Label |
|---|---------------------------------|-------|
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP]                         | [CLS] قصيدة من عيون الشعر [SEP] | True  |
| [SEP] ( تستعمل للتوكيد ) ذاته ، ذاته ( تستعمل للتوكيد ) [SEP] | [CLS] قصيدة من عيون الشعر [SEP] | False |
| [SEP] ( تستعمل للتوكيد ) ذاته ، ذاته ( تستعمل للتوكيد ) [SEP] | [CLS] جاء القوم أعينهم [SEP]    | True  |
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP]                         | [CLS] جاء القوم أعينهم [SEP]    | False |

## ❖ Three Fine-tuned BERT Models

- WSD into **binary sequence-pair classification task**
- **Accuracy 84%**
- 4 types of signals to emphasize target words in context



# Related Work

## BERT-based:

Used context-gloss pairs in fine-tuning (Huang et al., 2019; Yap et al., 2020; Blevins and Zettlemoyer, 2020). El-Razzaz et al. (2021) used a small dataset, but the experiment is not reliable.

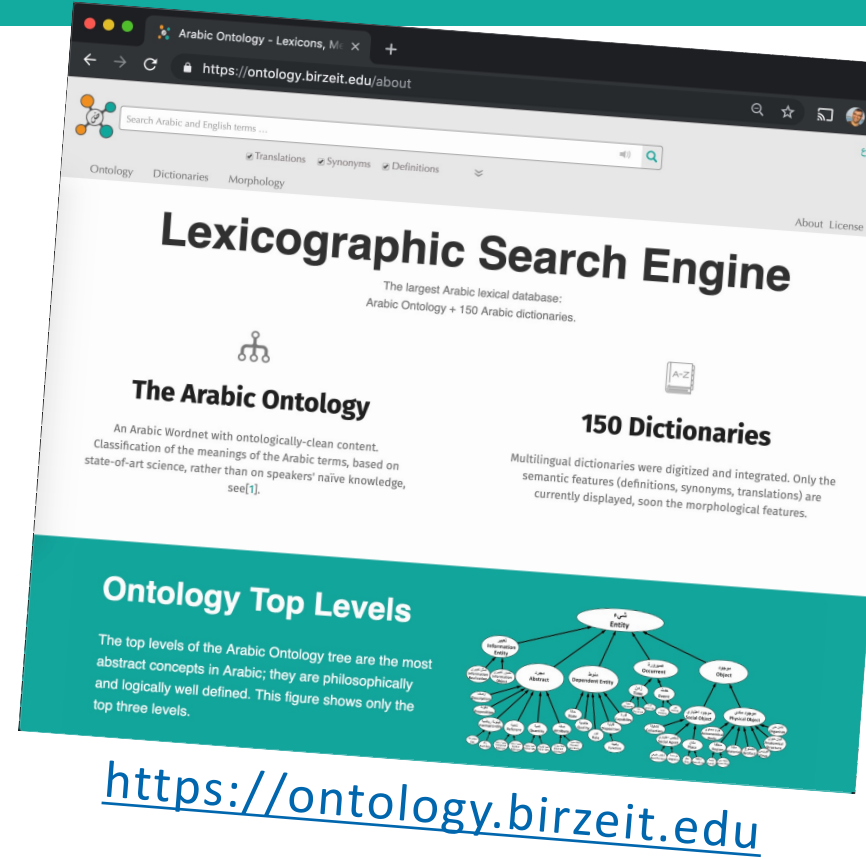
Used markers to emphasize target words in context-gloss Huang et al. (2019); Botha et al. (2020); Lei et al. (2017);

## Static Embeddings (related to Arabic WSD)

Sense vectors (Laatar et al., 2017), Stem2Vec and Sense2Vec (Alkhatlan et al., 2018), Lemma2Vec (Al-Hajj and Jarrar, 2021), Word Sense Induction (Alian and Awajan, 2020), or using fastText (Logacheva et al., 2020). Elayeb (2019) reviewed Arabic WSD approaches until 2018.

# Constructing a dataset of context-gloss pairs

- Extracted (60K pairs) from the **Arabic Ontology** and the **~400K dictionary definitions** available at Birzeit University. All are labeled as *True*.
- Generated 107k *False* pairs from the *True* pairs.
- Annotated target words in context



# Constructing a dataset of context-gloss pairs

## Statistics:

|                               | <b>count</b> |
|-------------------------------|--------------|
| Unique Lemmas (undiacritized) | 26169        |
| Avg glosses per Lemmas        | 1.25         |
| Unique Glosses                | 32839        |
| Unique Contexts               | 60272        |
| Avg context per gloss         | 1.83         |
| True context-gloss pairs      | 60323        |
| False context-gloss pairs     | 106884       |
| Total True and False pairs    | 167207       |

## Training and Test Datasets

- every context selected in the test set should not be selected in the training set;
- every gloss should be selected in both the training and the test sets.

| <b>Datasets</b> | <b>Pairs</b> | <b>Count</b>   | <b>Total</b> |
|-----------------|--------------|----------------|--------------|
| Training        | True pairs   | 55,585         | 152,035      |
|                 | False pairs  | 96,450         |              |
| Test            | True pairs   | 4,738          | 15,172       |
|                 | False pairs  | 10,434         |              |
|                 | <b>Total</b> | <b>167,207</b> |              |

Download: <https://ontology.birzeit.edu/downloads>



# Constructing a dataset of context-gloss pairs

- Annotating Target Words in Context

- Used four methods in parallel
  - Substring
  - Character-level cosine similarity
  - Levenshtein distance
  - Lemmatization
- All results were **validated manually**

|         |                            |
|---------|----------------------------|
| Lemma   | عين                        |
| Gloss   | أجود كل شيء وأحسنه ونفيسه  |
| Context | قصيدة من <u>عيون</u> الشعر |

- Used four variations of tagging target words

supervised signals used to emphasize target words in context during the BERT fine-tuning

|                                       |   |             |
|---------------------------------------|---|-------------|
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] | [CLS] قصيدة من عيون الشعر [SEP]   | Variation 1 |
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] | [CLS] قصيدة من 'عيون' الشعر [SEP] عيون: أجود كل شيء وأحسنه ونفيسه [SEP]                   | Variation 2 |
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] | [CLS] قصيدة من [UNUSED0] عيون [UNUSED0] الشعر [SEP] عيون: أجود كل شيء وأحسنه ونفيسه [SEP] | Variation 3 |
| [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] | [CLS] قصيدة من [UNUSED0] عيون [UNUSED1] الشعر [SEP] عيون: أجود كل شيء وأحسنه ونفيسه [SEP] | Variation 4 |

## ❖ Fine-tuned three Arabic pre-trained BERT models

The WSD task is converted into **binary sequence-pair classification task**

[CLS] قصيدة من عيون الشعر [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] True  
 [CLS] قصيدة من عيون الشعر [SEP] عين الشيء : نفسه ، ذاته ( تستعمل للتوكيد ) [SEP] False  
 [CLS] جاء القوم أعينهم [SEP] عين الشيء : نفسه ، ذاته ( تستعمل للتوكيد ) [SEP] True  
 [CLS] جاء القوم أعينهم [SEP] أجود كل شيء وأحسنه ونفيسه [SEP] False

## ❖ Results

| Model      |           | True | False | Accuracy |
|------------|-----------|------|-------|----------|
| AraBERTv02 | Precision | 81   | 85    | 84       |
|            | Recall    | 66   | 93    |          |
|            | F1-score  | 72   | 89    |          |
| CAMELBERT  | Precision | 77   | 83    | 82       |
|            | Recall    | 60   | 92    |          |
|            | F1-score  | 67   | 87    |          |
| QARiB      | Precision | 73   | 82    | 80       |
|            | Recall    | 58   | 90    |          |
|            | F1-score  | 65   | 86    |          |

## Tagging target words while fine-tuning

To emphasize target words during the BERT fine-tuning

Tested 4 variations of tags

### ❖ Results (with AraBERTv02)

| Variation                       |           | True | False | Accuracy |
|---------------------------------|-----------|------|-------|----------|
| <b>Variation 1</b><br>No signal | Precision | 80   | 85    | 83       |
|                                 | Recall    | 64   | 92    |          |
|                                 | F1-score  | 71   | 88    |          |
| <b>Variation 3</b><br>UNUSED0   | Precision | 81   | 85    | 84       |
|                                 | Recall    | 64   | 93    |          |
|                                 | F1-score  | 71   | 89    |          |
| <b>Variation 4</b><br>UNUSED0,1 | Precision | 81   | 85    | 84       |
|                                 | Recall    | 64   | 93    |          |
|                                 | F1-score  | 71   | 89    |          |

# Summary

Converted: **WSD** task into *binary sequence-pair classification*

Constructed Dataset: labeled Arabic context-gloss pairs (167K)

Annotated: Target words in context, to emphasize them

Fine-tuned: **BERT**: three models (84% accuracy)

# Thank You

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