# A High Coverage Method for Automatic False Friends Detection for Spanish and Portuguese

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## Introduction

**Objective**: classify between false friends or cognates for Spanish-Portuguese

**False friends**: pair of words from different languages that are written or pronounced in a similar way, but have different meanings.

#### **Example False Friends**

obligado — obrigado no - noaceite — aceite borracha — borracha cadera — cadeira desenvolver – desenvolver propina — propina

# Motivation

False friends make harder to **learn a language** or to **communicate**, especially when it's similar to the mother tongue.

• Between Spanish and Portuguese, the amount of cognates reaches the 85% of the total vocabulary (Ulsh, 1971).

Frunza, 2006: supervised machine learning using orthographic distances as features to classify between cognates, false friends or unrelated.

**Mitkov et al., 2007:** used a combination of distributional and taxonomy-based approaches. Worked with English-French, English-German and English-Spanish.

They use WordNet taxonomy similarities to classify, and if a word is missing they fall back to a distributional method.

#### Mitkov et al., 2007

For the distributional method they build vectors based on word windows, computing the co-occurrence probability. Then, they compared the N closest words of each word in the pair, translate one of them and count occurrences in the other one. They defined a threshold based on Dice coefficient.

**Ljubešić et al., 2013**: based on (Mitkov et al., 2007), experiment with several ways to build the vector space (e.g. tf-idf) and measure vector distances (e.g. cosine distance). They also proposed to use PMI.

They worked with closely related languages: Slovene and Croatian.

**Sepúlveda and Aluísio, 2011**: false friends resolution for Spanish-Portuguese, highly based on (Frunza, 2006).

They added an experiment with a new feature whose value is the **likelihood of translation**, from a probabilistic dictionary (generated taking a large sentence-aligned bilingual corpus).

# Word Vector Representations

Related work crafted their own word vector representations. We propose to use the skip-gram-based **word2vec** model (Mikolov et al, 2013a).

## Transform between Vector Spaces

**Mikolov et al, 2013b:** propose a method to correspond two word2vec vector spaces via a **linear transformation**.

Used to build dictionaries and phrase tables.

#### **Transform between Vector Spaces**



# Our Method

Build word2vec vector spaces, find a linear transformation and measure vector distances.

Note that we don't cope with related/unrelated, we just focus on cognate/false friends



## Our Method

We used the **Wikipedia**'s for the vector spaces.

**Open Multilingual WordNet** (Bond and Paik, 2012) was used as a bilingual lexicon to fit the linear transformation: we iterate over synsets and took lexical units from each language. Then we employed Least Squares.

# Our Method

We take one of the word vectors, transform it to the other space and compute:

- The cosine **distance** between T(source\_vector) and target\_vector.
- The number of word vectors in the target vector space closer to target\_vector than T(source\_vector).
- The sum of the distances between target\_vector and T(source\_vector\_i) for the top 5 word vectors source\_vector\_i nearest to source\_vector.

## Experiments

We used (Sepúlveda and Aluísio, 2011) dataset, which is composed by 710 pairs (338 cognates and 372 false friends).

# Experiments

Method	Accuracy	Coverage
WN Baseline	68.18	55.38
Sepúlveda 2	63.52	100.00
Sepúlveda 3.2	76.37	59.44
Apertium	77.75	66.01
Our method	77.28	97.91
+ frequencies	79.42	97.91

## Experiments: different configurations

Method configuration	Accuracy
es-400-100	77.28
es-800-100	76.99
es-100-100	76.98
es-200-100	76.84
es-200-200	76.55
pt-200-200	76.13
es-200-800	75.99
pt-400-100	75.99
pt-100-100	75.84
es-100-200	75.83
es-100-100-2	74.98

# **Experiments: bilingual lexicon**



# Conclusions

- We have provided a new approach to classify false friends with high accuracy and coverage.
- We studied it for Spanish-Portuguese and provided state-of-the-art results for the pair.
- The method doesn't require rich bilingual datasets.
  - It could be easily applied to other language pairs.

# Future Work

- Experiment with other word vector representations and state-of-the-art vector space linear transformation.
- Work on fine-grained classifications.
  - E.g., partial false friends.



**Questions?** 

Code and slides available at: **github.com/pln-fing-udelar/false-friends**