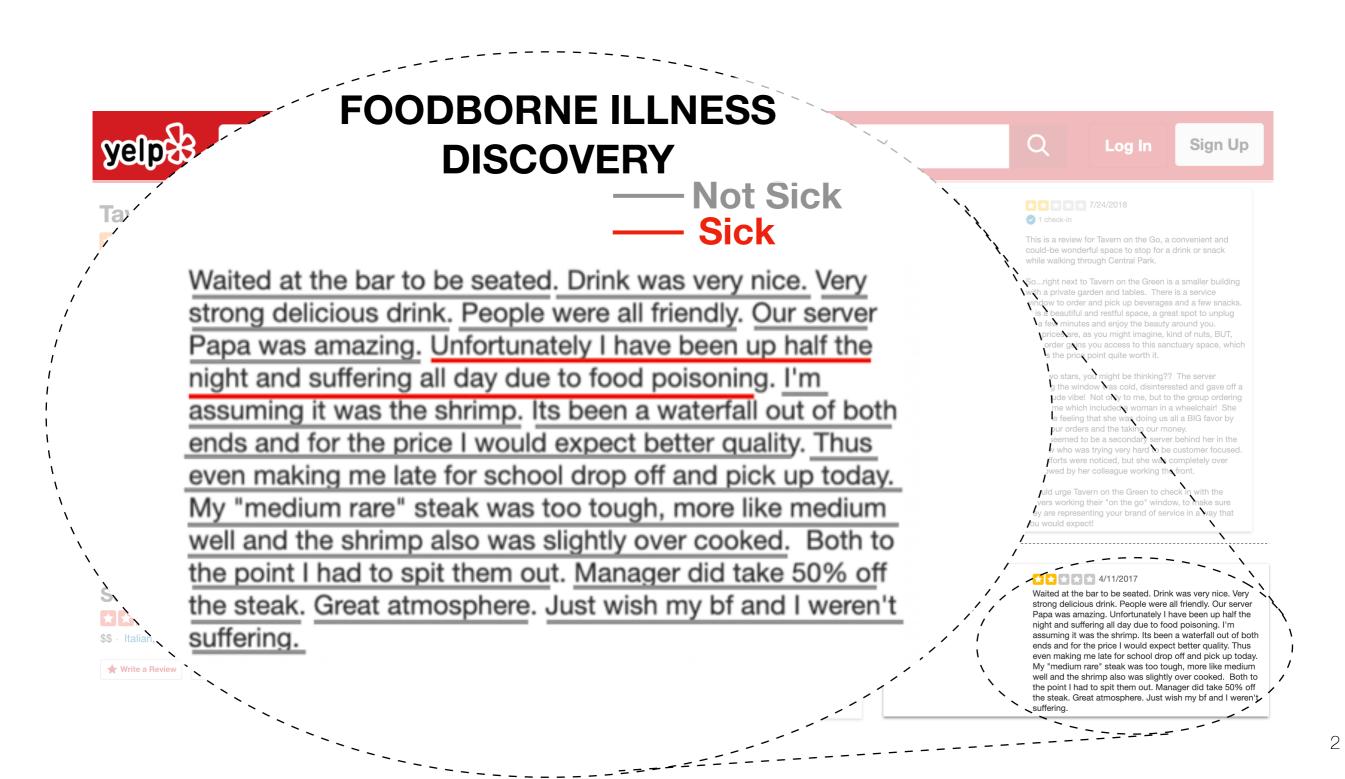
Detecting Foodborne Illness Complaints in Multiple Languages Using English Annotations Only

Ziyi Liu, Giannis Karamanolakis, Daniel Hsu, Luis Gravano Department of Computer Science, Columbia University zl8888@columbia.edu, {gkaraman, djhsu, gravano}@cs.columbia.edu



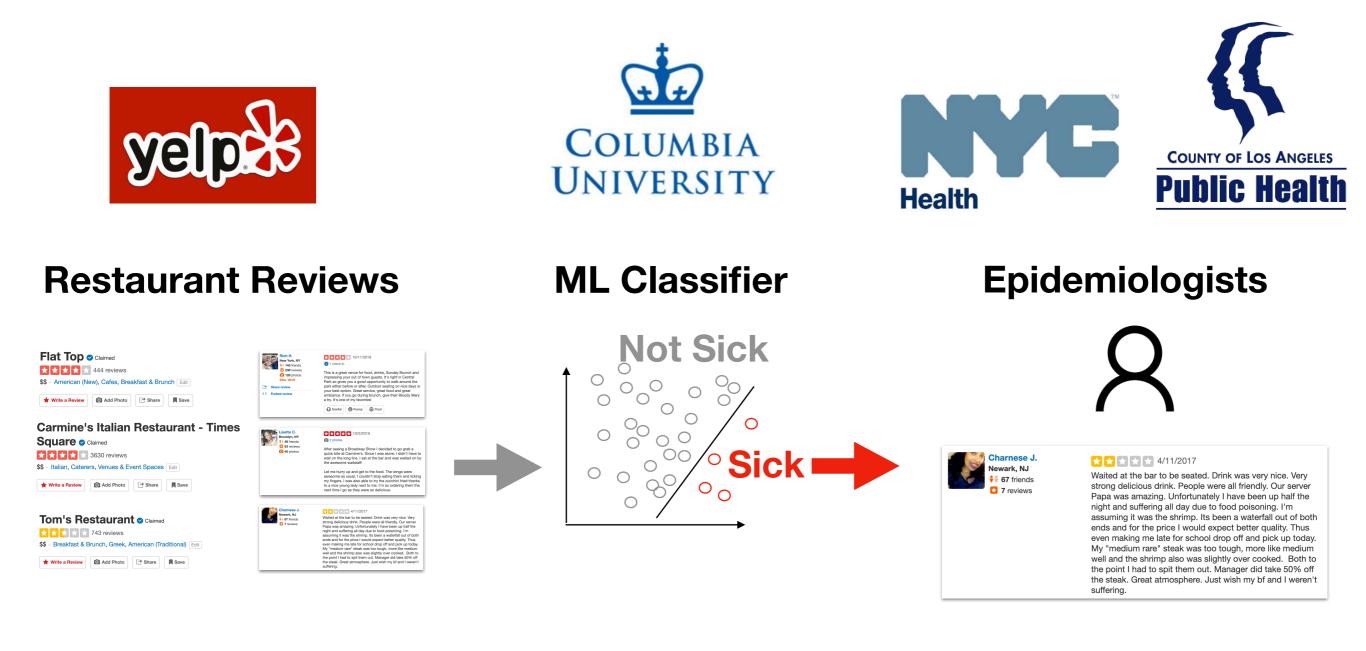
Social Media Analysis For Public Health



• Important for the early detection of foodborne illness outbreaks in restaurants!

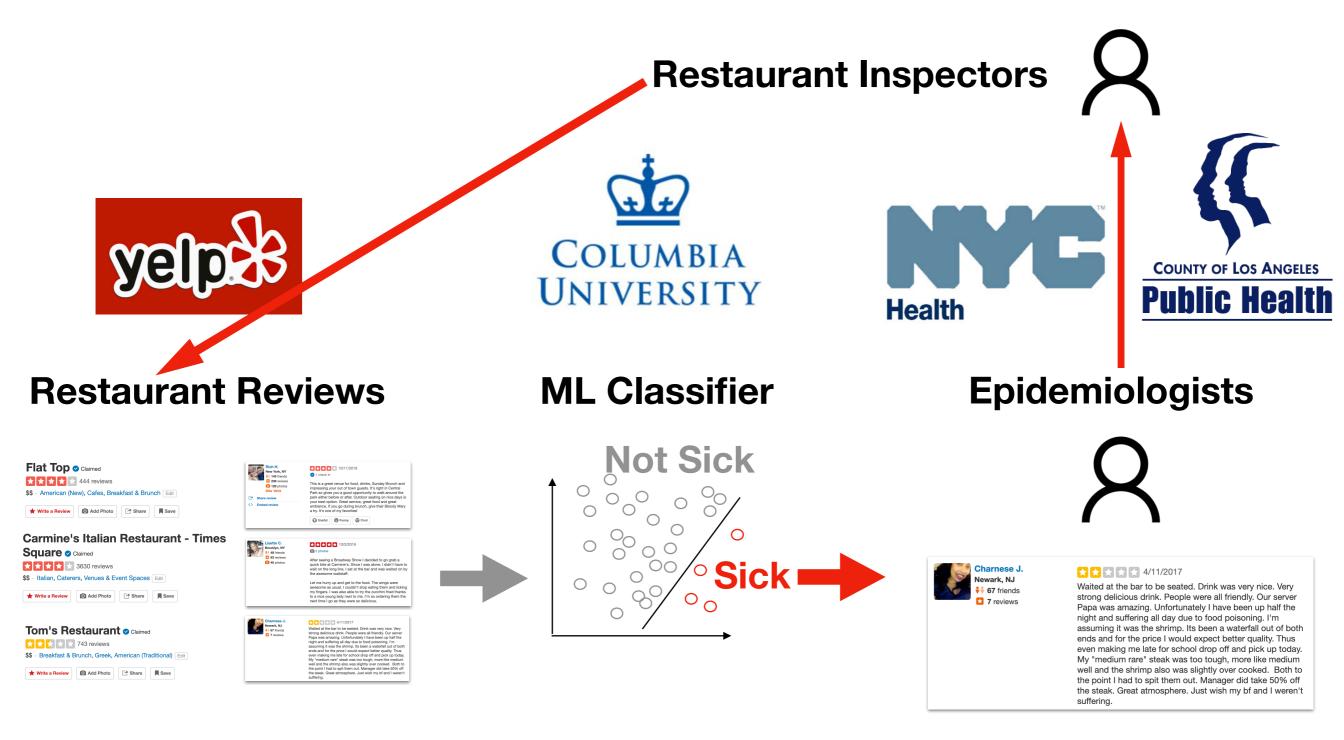
Detecting Foodborne Illness from Restaurant Reviews

Collaboration between Yelp, Columbia, NYC/LA health departments



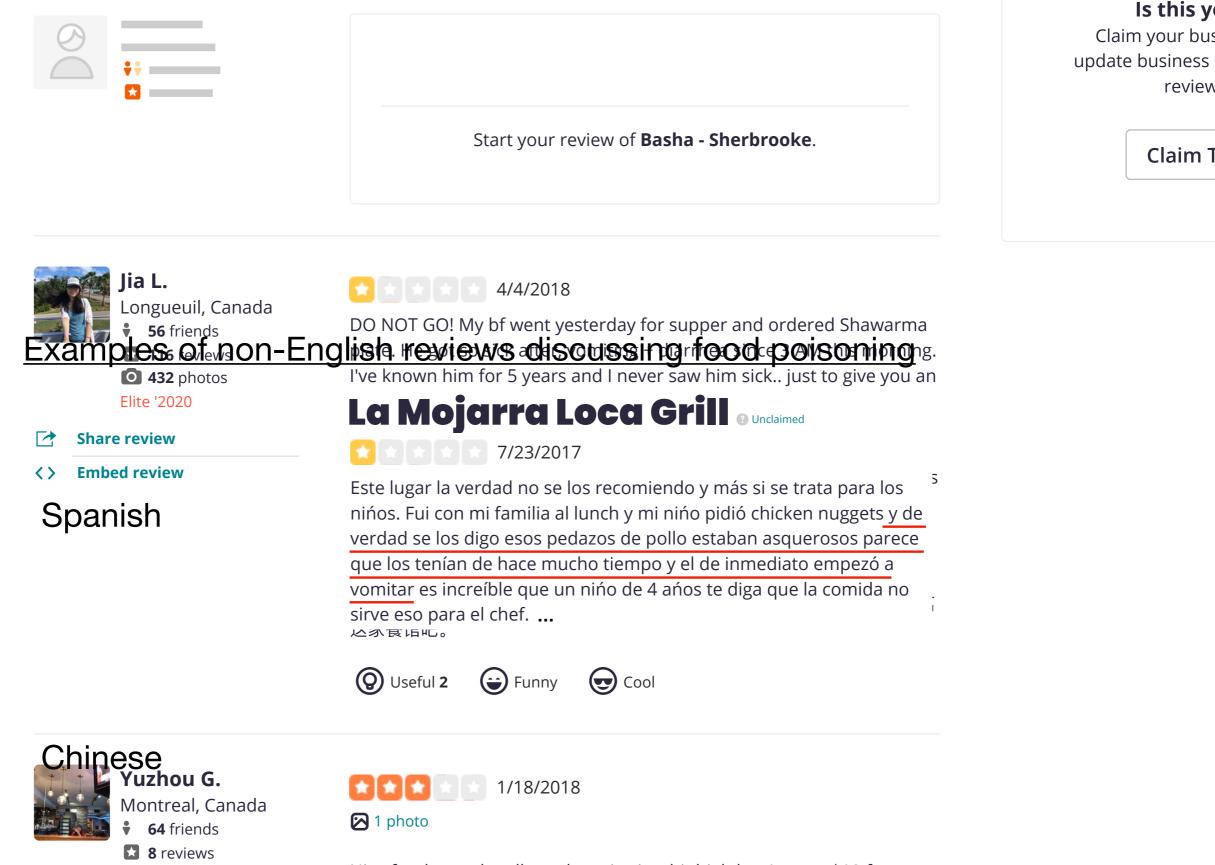
http://publichealth.cs.columbia.edu/

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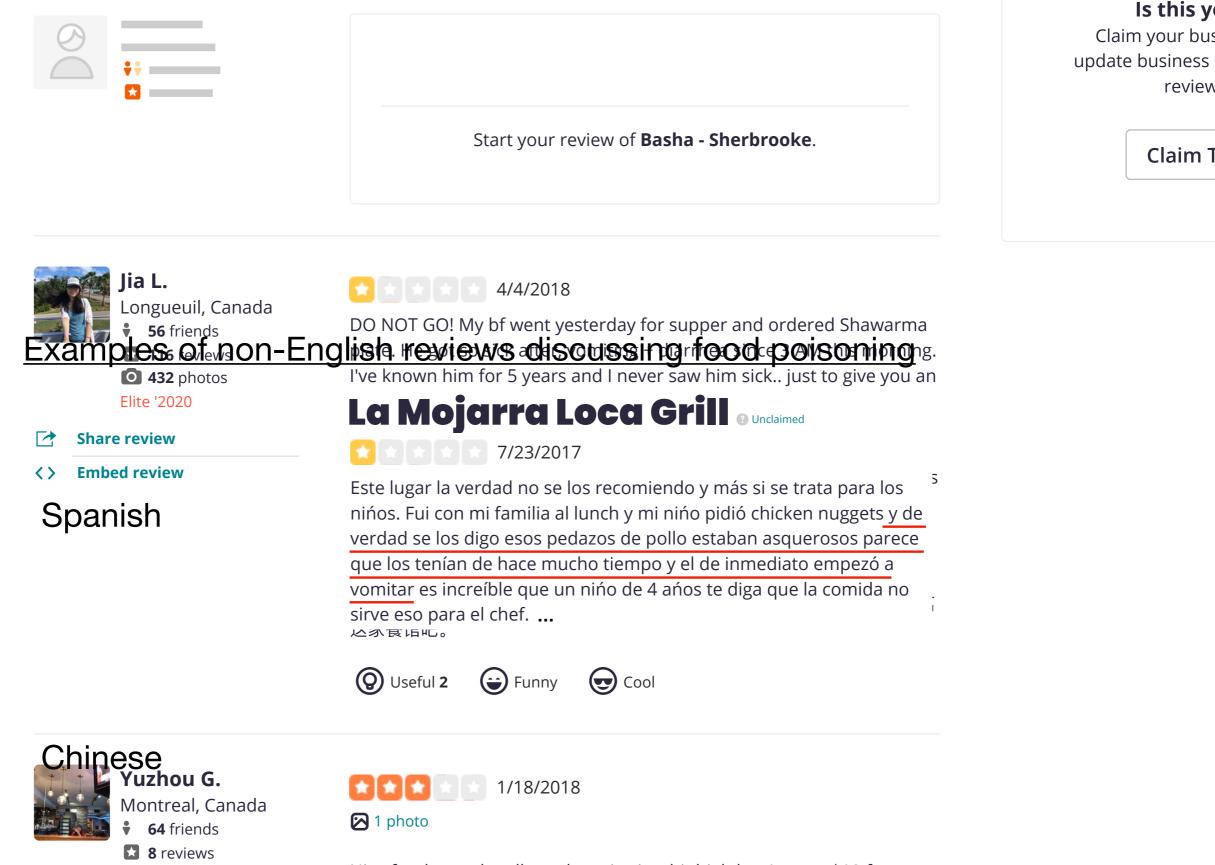
Extending Our System Beyond English



Nice food, people tell me the price is a bit high but I guess \$12 for a very big meal include rice, baked potato, chicken, beef, salad and pita

0 7 photos

Extending Our System Beyond English



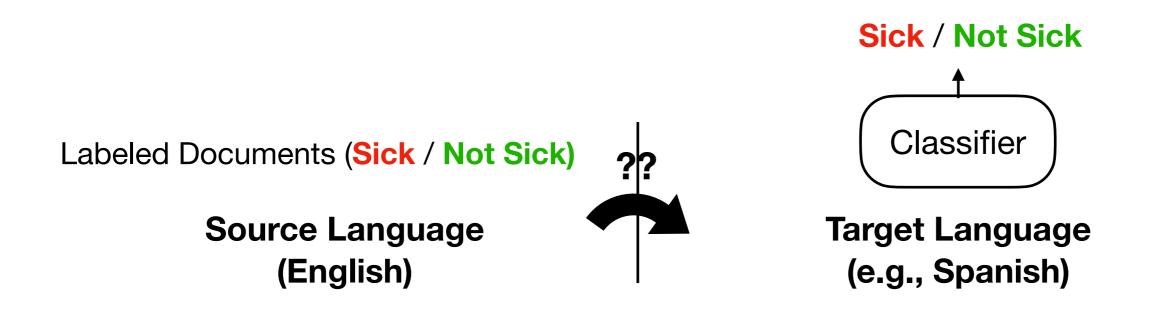
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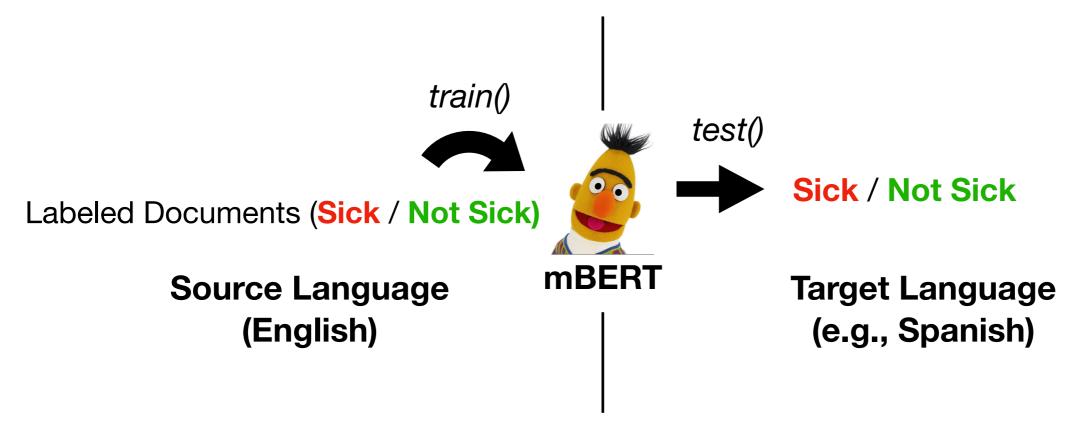
- Train a classifier for a target language...
- ... using labeled documents from a **source** language



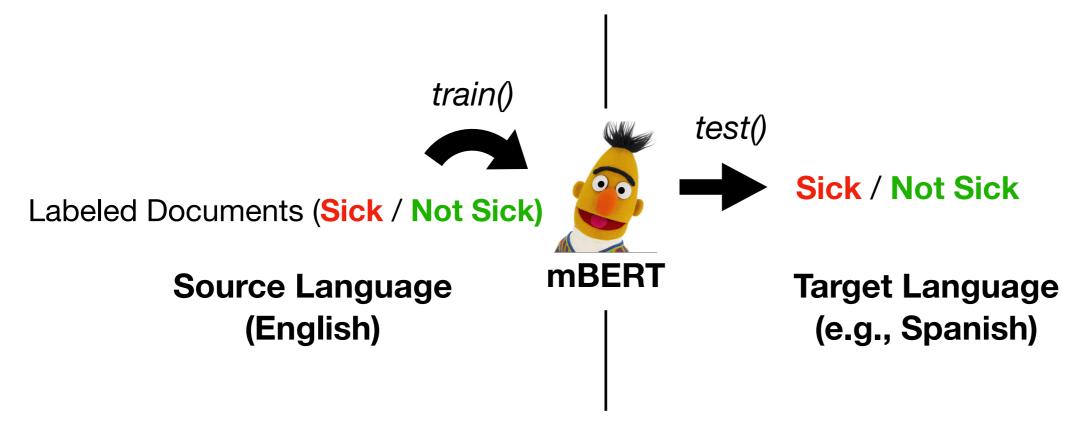
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- Train a classifier for a target language...
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- Recent approach: zero-shot classifiers using pre-trained multilingual models
 - Multilingual BERT (mBERT): train in source language, test in target language



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(-) not effective: pre-trained mBERT does not capture foodborne illness

This Work

1. We present a **cross-lingual learning approach** for foodborne illness detection in non-English social media documents.

(+) efficient: requires only English labeled data.

- 2. We improve the performance of mBERT for our rare classification task(+) effective: generates artificial labeled data in multiple languages
- 3. We evaluate our approach in **7 languages** and highlight its potential for successful deployment in health departments

Outline

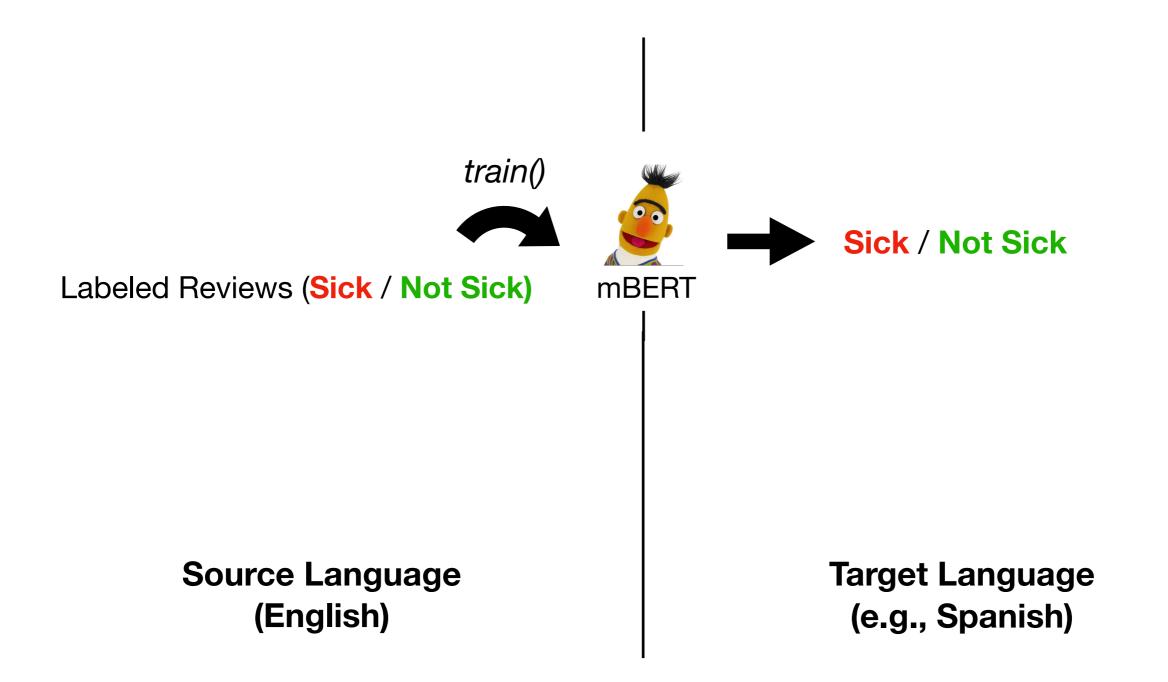
1. Intro: Multilingual Foodborne Illness Detection

2. Our Approach

- 3. Experiments in 7 Languages
- 4. Conclusions and Future Work

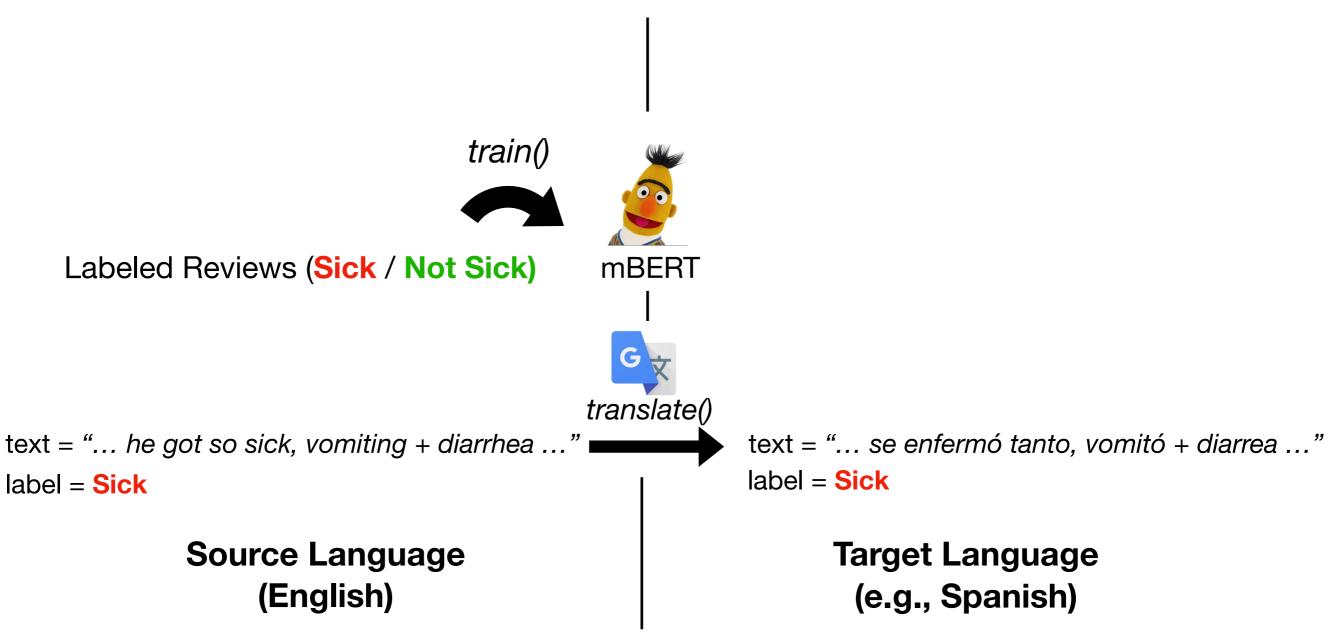
Encouraging Stronger Cross-Lingual Alignment

Pre-trained mBERT does not capture foodborne illness



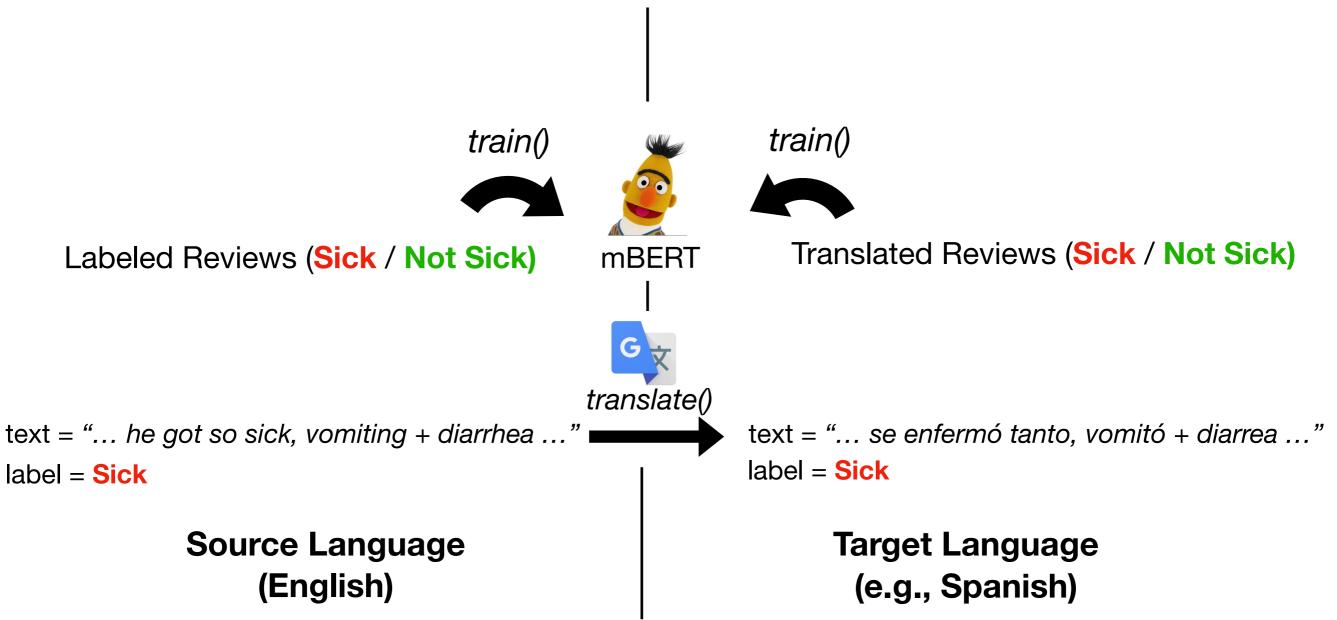
Encouraging Stronger Cross-Lingual Alignment

- Pre-trained mBERT does not capture foodborne illness
- We create artificial non-English training data using Machine Translation



Encouraging Stronger Cross-Lingual Alignment

- Pre-trained mBERT does not capture foodborne illness
- We create artificial non-English training data using Machine Translation
- Then, we fine-tune mBERT in both source and target languages



Training mBERT on Multiple Source Languages

• We generate training datasets in multiple languages

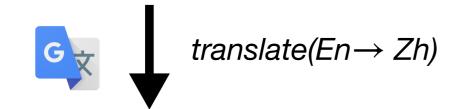
Translated Spanish Reviews (Sick / Not Sick)

"... se enfermó tanto, vomitó + diarrea ..."



Labeled English Reviews (Sick / Not Sick)

"... he got so sick, vomiting + diarrhea ..."



Translated Chinese Reviews (Sick / Not Sick)

"他病得很厉害,呕吐和腹泻"

Training mBERT on Multiple Source Languages

- We generate training datasets in multiple languages
- We fine-tune mBERT for all languages in parallel
 - (+) May prevent overfitting to a single source language
 - (+) Final model can be applied on any target language (supported by mBERT)



Outline

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Experiments: Foodborne Illness Detection in Multiple Languages

Datasets:

1. Labeled restaurant reviews in English

- 25K reviews (20K train, 5K test) annotated by epidemiologists

2. Unlabeled restaurant reviews

- Sources: Yelp NYC, Yelp Los Angeles, Yelp Challenge Dataset
- Languages:
 - 1. English (En)
 - 2. Spanish (Es)
 - 3. Chinese (Zh)
 - 4. French (Fr)
 - 5. German (De)
 - 6. Japanese (Ja)
 - 7. Italian (It)

Experiments: Foodborne Illness Detection in Multiple Languages

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- Evaluation Procedure:
 - Train mBERT on source language(s)
 - Test mBERT on each target language (~5K test reviews translated from English)

Model	Train Language	Average F1 (across 7 languages)	
mBERT	English (zero-shot)	74.6	
	Target	88.2 (+18.2%)	
	English+Target	89.2 (+19.6%)	
	ALL (7 languages)	89.3 (+19.7%)	

Target > English (zero-shot)

 In-language training documents are important to learn language-specific features!

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English + Target > Target

- Training mBERT jointly on English and target language is more effective than training on each language separately
- Considering both languages encourages stronger cross-lingual alignment for foodborne illness aspect

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Training mBERT for ALL languages > Target > zero-shot

Training signals from multiple languages lead to better cross-lingual representations

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Training mBERT for ALL languages > Target > zero-shot

Training signals from multiple languages lead to better cross-lingual representations

Comparable performance to English+Target

ALL > English+Target for target \in {English, French, German, Italian} ALL < English+Target for target \in {Chinese, Japanese}

Considering ALL languages degrades performance on non-Romance languages

Model	Train Language	Average F1 (across 7 languages)	
mBERT	English (zero-shot)	74.6	1 model
	Target	88.2 (+18.2%)	7 models
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Training mBERT for ALL languages > Target > zero-shot

- Training signals from multiple languages lead to better cross-lingual representations
- Comparable to English+Target
- Easier deployment: single model for all languages without language detectors

We Detect Foodborne Illness Complaints In Real-World Yelp Reviews

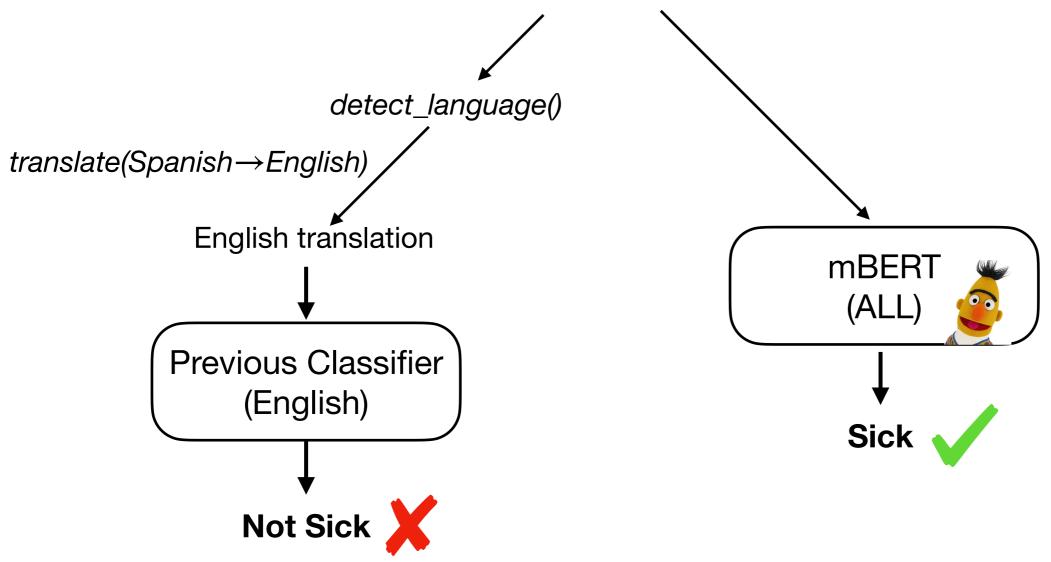
Yelp Review (Spanish)

Definitivamente mi peor experiencia, me intoxique con un ostra mala, llevo 4 días en muy malas condiciones, por favor tengan cuidado, los ostiones y mariscos no se pueden comer en cualquier lugar, yo aprendi por las malas, espero que mi experiencia le sirva a alguien

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See our paper for more results and examples!

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1. Intro: Multilingual Foodborne Illness Detection

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Foodborne Illness Detection in Multiple Languages

- We presented a **cross-lingual learning approach** for foodborne illness detection beyond English that only requires English labeled reviews
- We showed promising improvements over zero-shot **mBERT** by creating training datasets in **multiple** languages through machine translation
- We evaluated our approach in **7 languages** and demonstrated its potential for successful deployment in health departments

Current And Future Work

- Creating human-labeled evaluation datasets in multiple languages
 - Previous test sets are translated from English and thus may express foodborne illness differently than native-language reviews
 - -We have been creating better evaluation datasets via crowdsourcing
- Detecting foodborne illness in languages with limited resources
 - -Our approach requires mBERT + Google Translate, which are available for 103 out of about 4,000 written languages
 - -We will apply our recent cross-lingual transfer approach [1] that **does not require** machine translation or pre-trained multilingual models
- Applying for other (rare) text classification problems related to public health

[1] "Cross-lingual text classification with minimal resources by transferring a sparse teacher" Giannis Karamanolakis, Daniel Hsu, Luis Gravano, Findings of EMNLP '20

Thank you!

Our project: http://publichealth.cs.columbia.edu/

Contact gkaraman@cs.columbia.edu https://gkaramanolakis.github.io

