



INFORMATICS PROFESSIONALS. LEADING THE WAY.

Detecting Tweets Mentioning Drugs with a Deep Neural Network Ensemble

Deep Learning

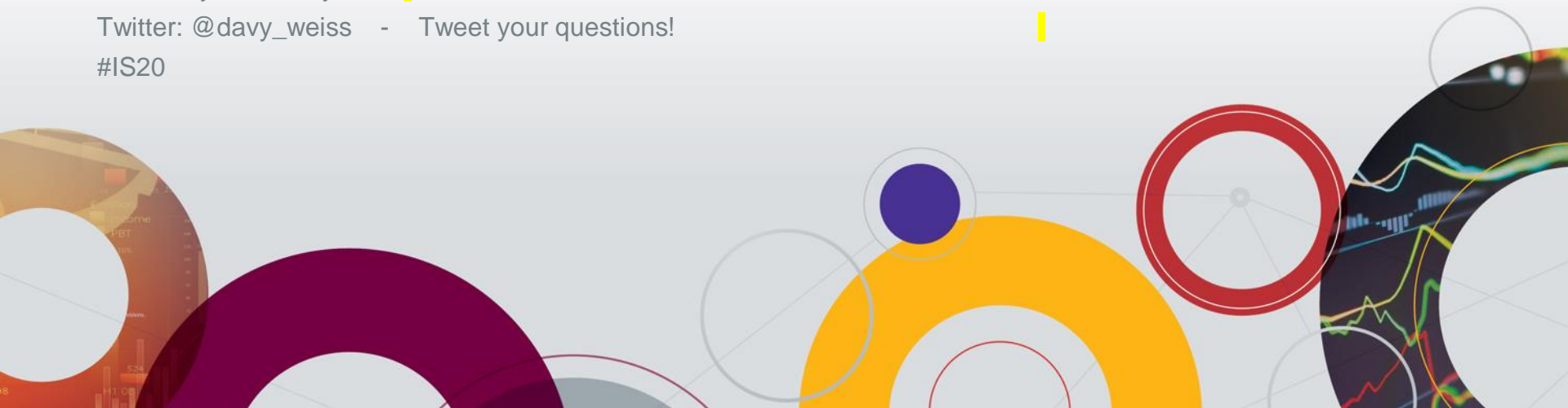
VS07

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Twitter: @davy_weiss - Tweet your questions!

#IS20



Disclosure

I and my wife have NO relevant relationship with commercial interest to disclose.

Learning Objectives

After participating in this session the learner should be able to:

- Define the problem of drug names detection in tweets
- Explain the limitation of the lexical search method
- List the main methods for detecting drug names
- Explain the best approach based on deep neural networks
- Report on the current performance of drug names detection when applied on 112 home timelines

Why extracting tweets mentioning drugs?

- **Tweets mentioning drugs** useful for:
 - Syndromic surveillance
 - Pharmacovigilance
 - Monitoring drug abuse
- Typically, collections use drug names as keywords – inherent bias
- In a ‘natural’ timeline, these tweets **are rare**, only 0.26% in our corpus
- Lexical searches fail:
 - Misspellings
 - Ambiguity
- **Goal: Extract drug mentions from user timelines (an extremely sparse corpus) using Deep Neural Networks**

Detecting Tweets Mentioning Drugs

- Drug definition
 - Drug product as defined by the FDA (prescription & over the counter)
 - Dietary supplements
- Resolving a **binary classification** task
 - Lola may has a sty, or pink eye. Doc recommends warm compress => 0 (not mentioning a drug)
 - Meet Mr and Ms Lexapro... Guaranteed fidelity => 1 (mentioning a drug)

➡ Kusuri, 薬, demo:
<https://hlp.ibi.upenn.edu/kusuri/home/>

➡ The competition is now open @ #SMM4H, task 1:
<https://healthlanguageprocessing.org/smm4h-sharedtask-2020/>

UPennHLP Timeline Corpus

- **Corpus:** 112 timelines (98,959 tweets) of women twitting during their pregnancy
- **Annotation:** 282.5 hours of annotation, IAA $k=0.88$
- **Corpus split:**
 - 70% for training (181 positive, 69,091 negative examples)
 - 30% for evaluation (77 positive, 29,610 negative examples)
- Only 258 tweets mentioned a drug, *i.e.* 0.26% of the tweets

A DNN for Drug Detection

Attention layer helps

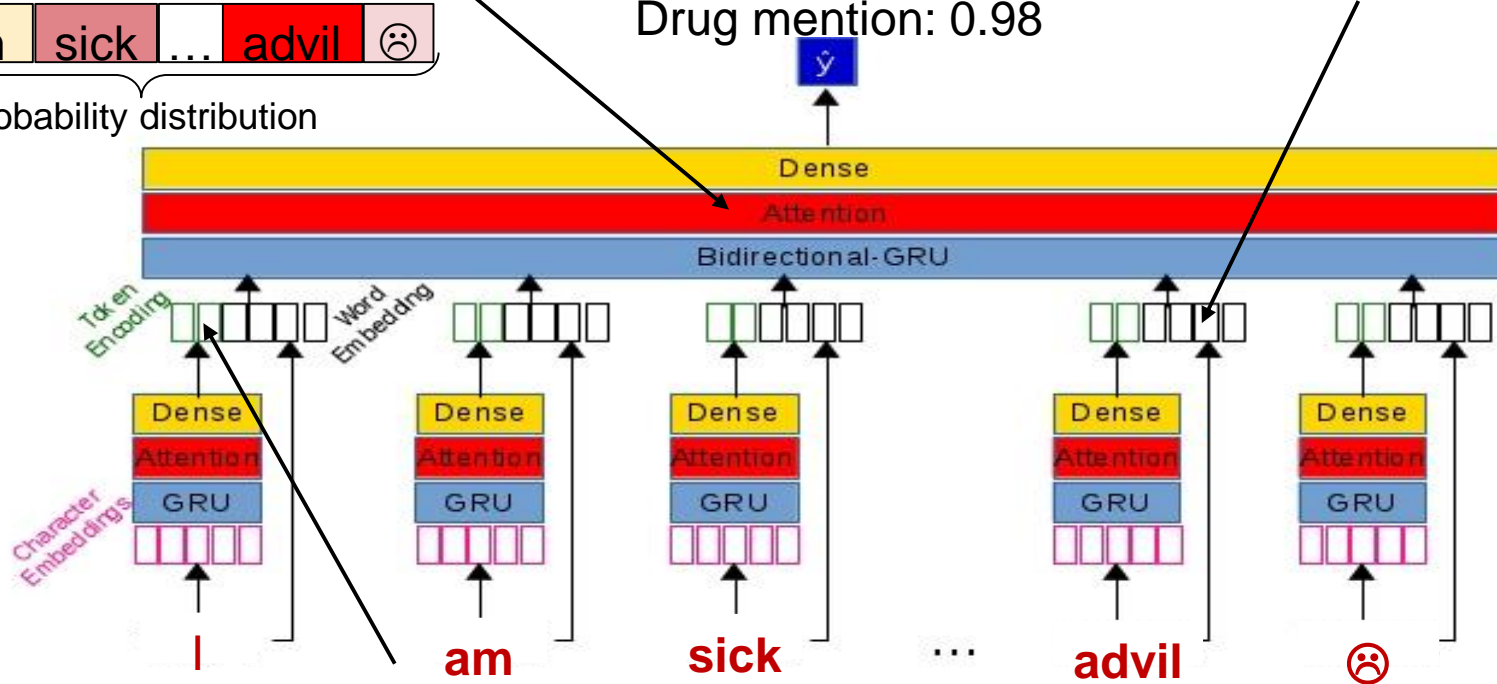
To focus on important words

Word Embeddings

Encode semantic of words

Drug mention: 0.98

Probability distribution



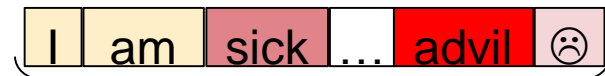
Character Embeddings
Encode morphology of words

A DNN for Drug Detection

Attention layer helps

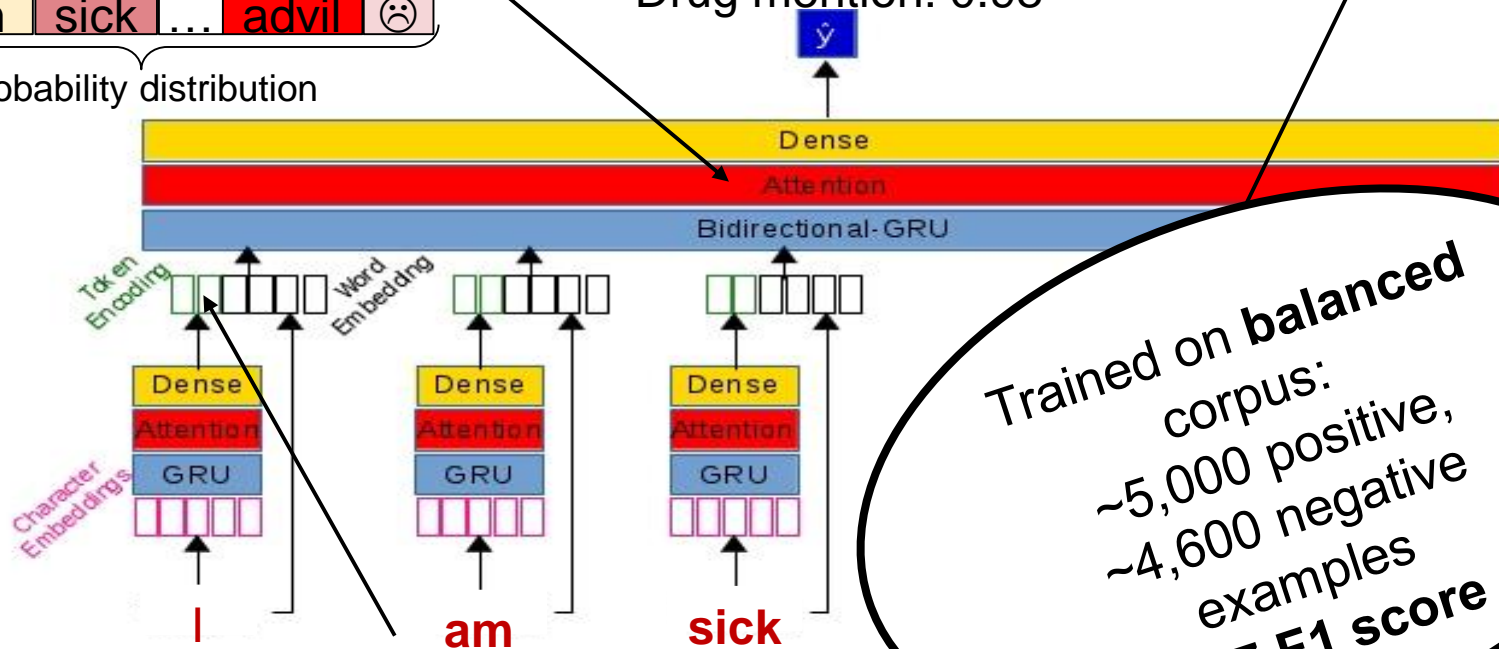
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Probability distribution

Drug mention: 0.98



Character Embeddings
Encode morphology of words

Trained on **balanced**
corpus:
~5,000 positive,
~4,600 negative
examples
93.7 F1 score

On unbalanced data, require pre-filtering

112 Timelines



Selecting tweets
potentially mentioning
drugs

Lexicon
Filter

RxNorm

Select tweets mentioning drug names in the lexicon

T1: I love **Lyrice**'s new song!!!

T2: **Lyrice** experience? I was on **Gabapentin**.

Variant
Filter

RxNorm

Select tweets mentioning generated variants of drug names

T3: My child likes her **tylonol**...

T4: im Allergic to all **Tynenol**, Aspirin, n ibuprofen.

T5: **Aderal** always motivates me

Pattern
Filter

Regular
Expressions

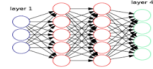
Select tweets containing phrases manually selected

T6: my OB **doctor prescribed me** promethazine

T7: I'm **on** Gabapentin which doesn't help much.

T8: I **took a** nap and I'm still sleepy

Deep Neural
Network
Filter



Select tweets recognized by a DNN

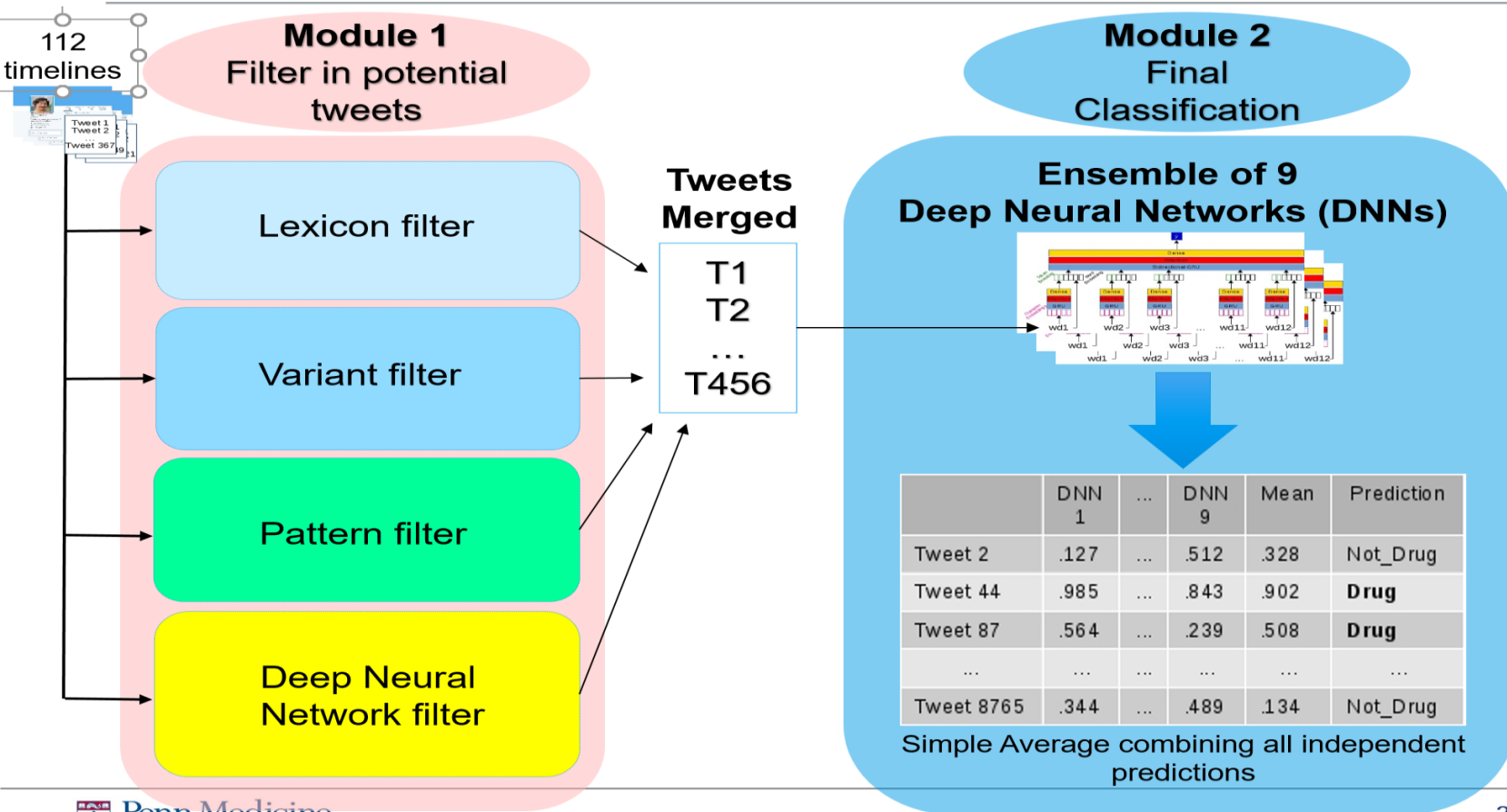
(DNN trained on tweets mentioning unambiguous drugs)

T9: I took 600 mg of **lyrica** am I going to be ok?

T10: **snickers bars** are my drug of choice

T11: Nasal allergies flare up during monsoon

Kusuri overview



Performance on Natural Corpus

Systems	Precision	Recall	F1
Lexicon+Variant classifier	55.0	71.4	62.1
Ensemble Bidirectional-GRUs (trained on unbalanced corpus)	87.5	63.6	73.7
Kusuri	94.55	67.5	78.8

Conclusion and Future Work

- On balanced corpus, performance is close to human annotators'
- On natural corpus, our system outperforms existing performances in well-established challenges [Limsopatham and Collier, 2016]
- Extreme imbalanced data is the main challenge:
 - Replacing the 4 filters by active learning
 - Exploiting unlabeled data with Generative Adversarial Networks

[Weissenbacher et al., Deep neural networks ensemble for detecting medication mentions in tweets. 2019. JAMIA.

Thank you!

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