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People on Drugs : Credibility of User Statements in Health Forums

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Motivation: Internet as a healthcare resource

59% of US population use internet for health information

[Pew Research Center Report, 2013]

Half of US physicians rely on online resources

[IMS Health Report, 2014]

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Credibility of user-generated online health information



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Posts from Healthboards.com

“My girlfriend always gets a bad **dry skin, rash** on her upper arm, cheeks, and shoulders when she is on **[Depo]**. . . .”

“I have had no side effects from **[Depo]** (except ...), but otherwise **no rashes**. She should see her gyno. She may be allergic to something”



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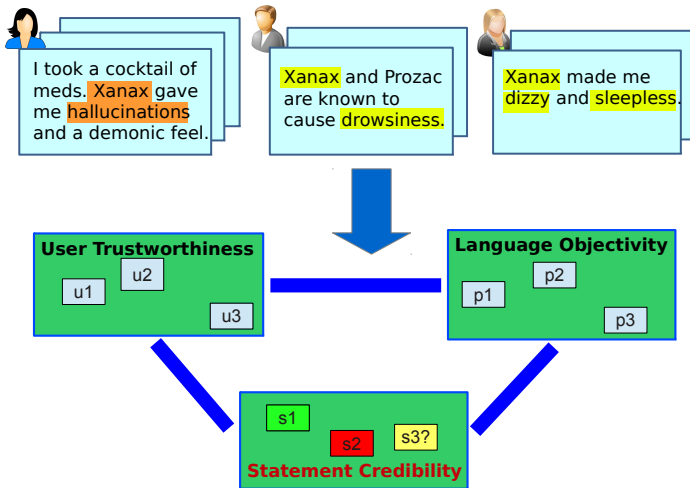
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Our Intuition

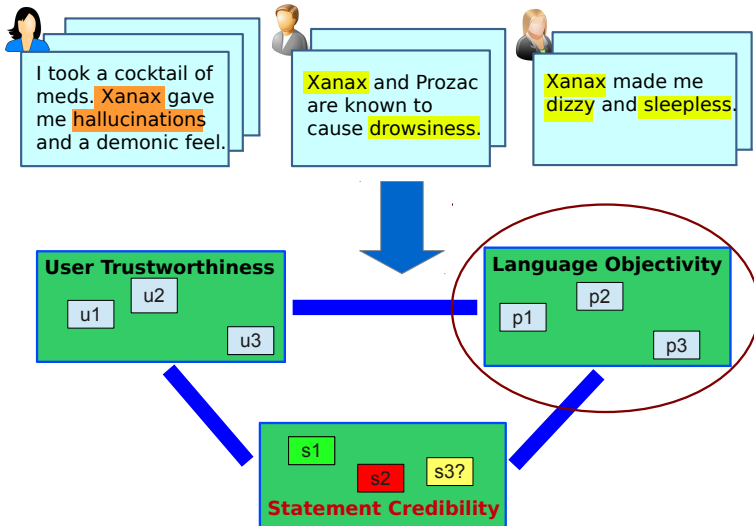
Users, language and credibility influence each other



Trustworthy users write credible posts
Agree with each other on credible statements



Our Intuition



Language: Stylistic Features

“I heard Xanax **can** have pretty bad side-effects. You **may** have peeling of skin, and apparently some friend of mine told me you **can** develop ulcers in the lips also. If you take this medicine for a long time then you **would** probably develop a lot of other physical problems. Which of these did you experience ?”

Usage of **modals**, indefinite determiner, conditional, probabilistic adverb, question particle, *etc.*



Language: Stylistic Features

“I heard Xanax can have pretty bad side-effects. You may have peeling of skin, and apparently **some** friend of mine told me you can develop ulcers in the lips also. If you take this medicine for a long time then you would probably develop a lot of other physical problems. Which of these did you experience ?”

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Language: Stylistic Features

“Depo is very dangerous as a birth control and has too many long term side-effects like reducing bone density. **Hence**, I will never recommend anyone using this as a birth control. Some women tolerate it well but those are the minority. Most women have horrible long lasting side-effects from it.”

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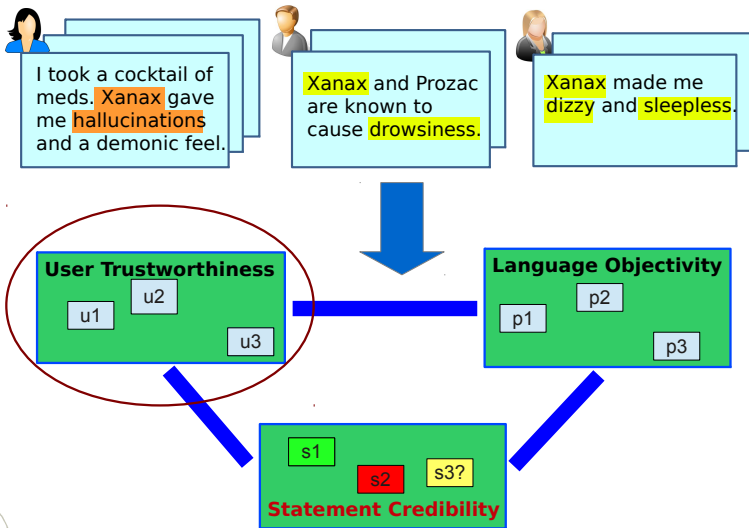


Language: Objectivity

“I started Cymbalta, but now I’m having a **panic attack** or an allergic reaction. I have a hardcore burning sensation in my chest and warm sensations all over. It’s like my body can’t decide whether it wants to be cold or hot. I feel if I close my eyes I’ll lose control, go **crazy** and pass out.”



Our Intuition

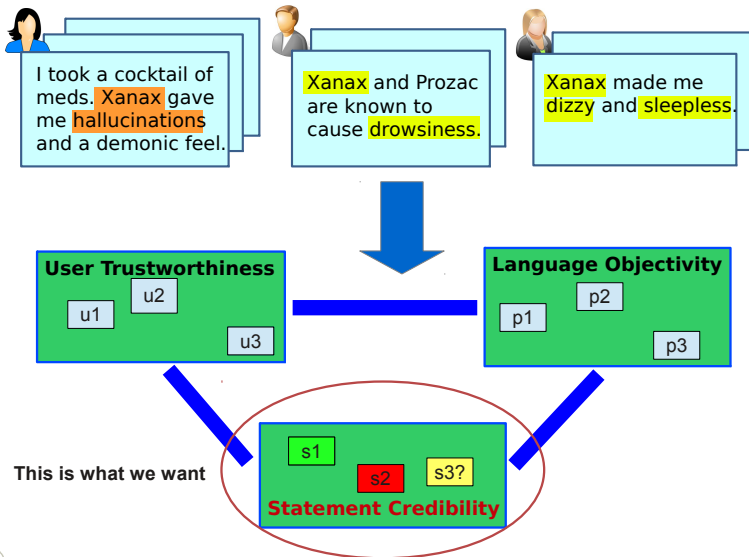


User Features

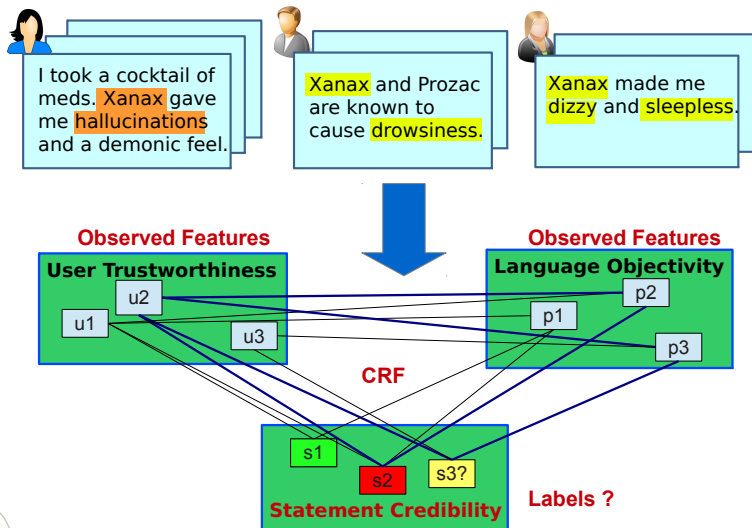
- ▶ User demographic features like *age, gender, location*
- ▶ Engagement features like *number of posts, questions, answers, thanks*
- ▶ User post properties like *avg. post length*



Objective



Probabilistic Inference: CRF

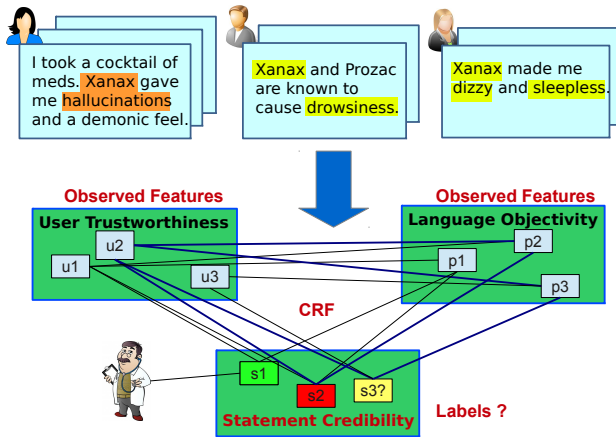


Predict the *most likely* label assignment of statements



Semi Supervised Learning

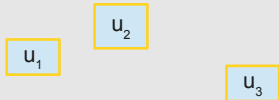
Protects against users conveying misinformation using confident and objective language



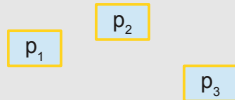
Expert stated side-effects of drugs from MayoClinic portal

Semi-Supervised CRF (Sketch)

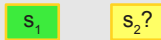
User Trustworthiness



Language Objectivity



Statement Credibility

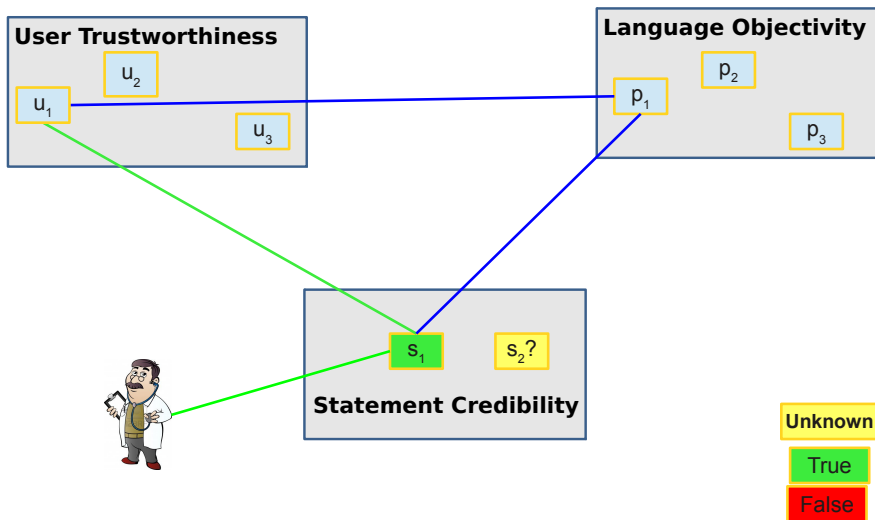


Unknown

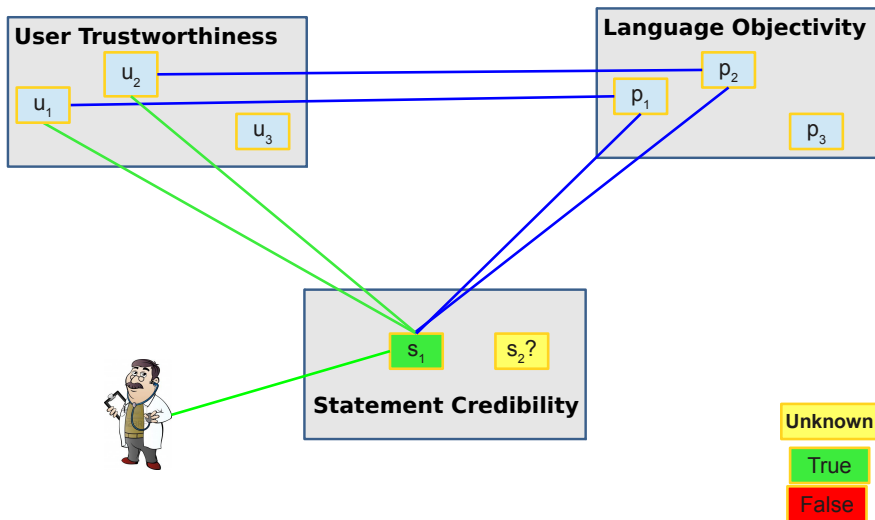
True

False

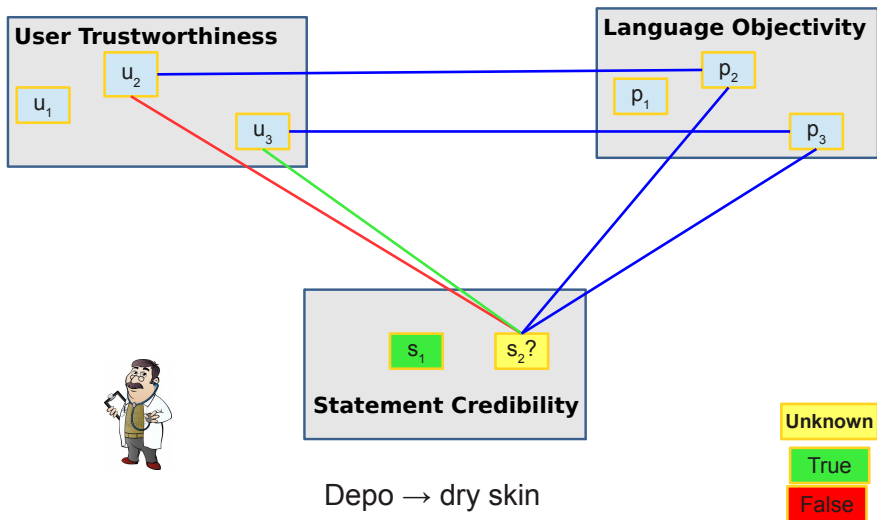
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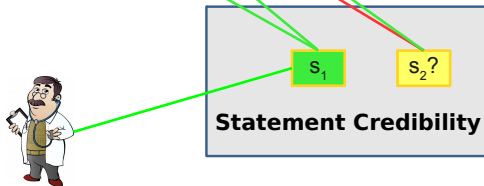
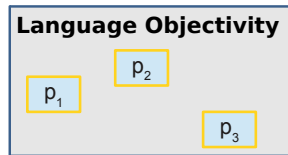
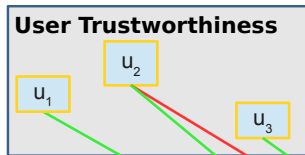
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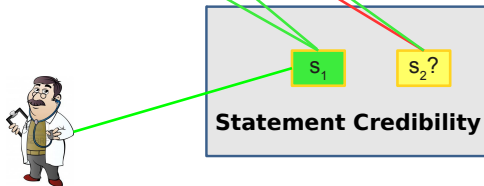
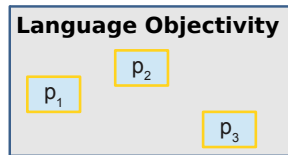
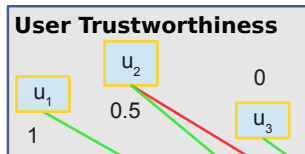
Semi-Supervised CRF (Sketch)



1. Estimate user trustworthiness : $t_k = \frac{\sum_i \mathbb{1}_{S_{i,k}=\text{True}}}{|S_k|}$

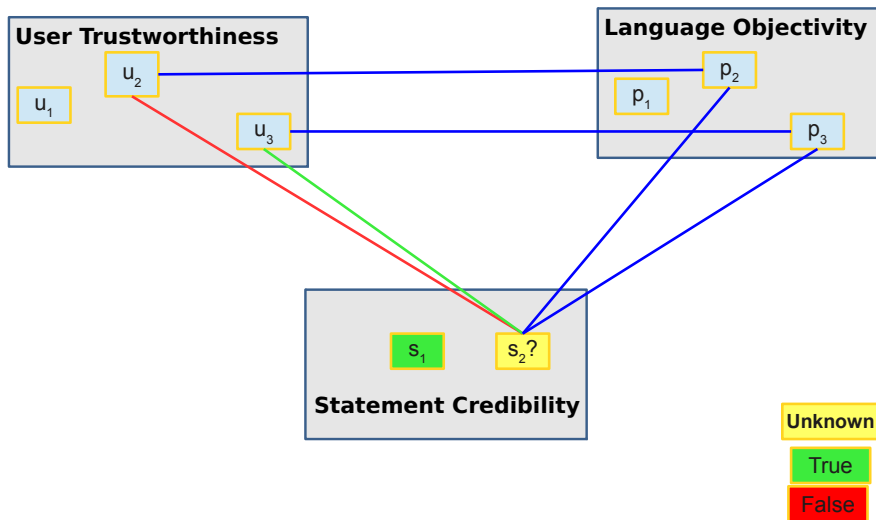


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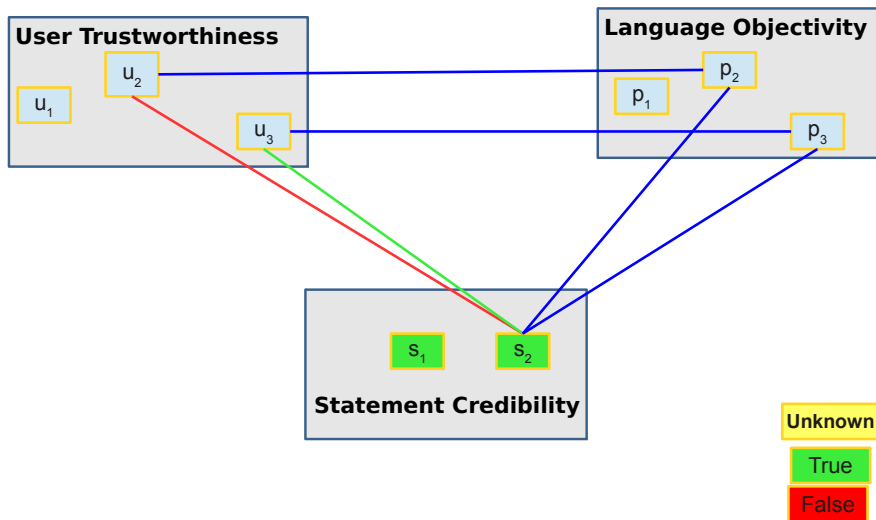
2. E-Step : Estimate label of unknown statements by Gibbs' sampling :

$$Pr(S_i^U | P, U, S^L; W) \propto \prod_{\nu \in C} t_k \times \phi_{\nu}(S_{\nu}^*, p_j, u_k; W)$$



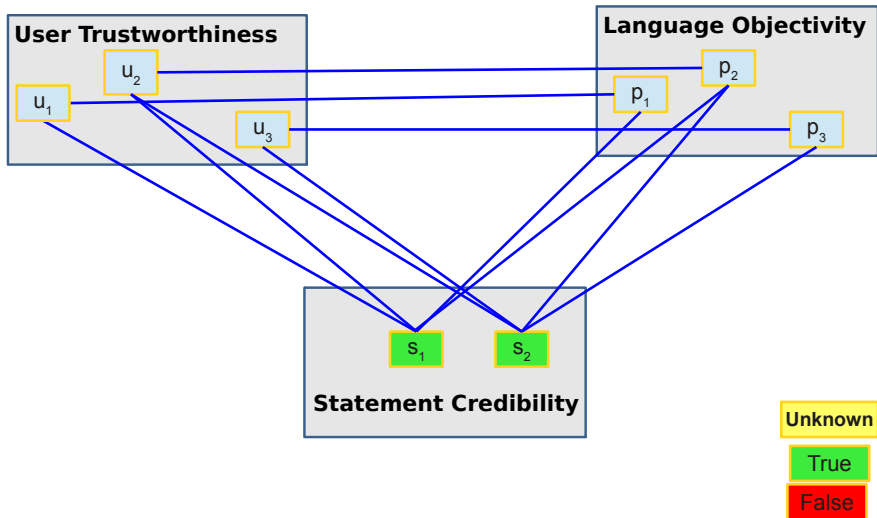
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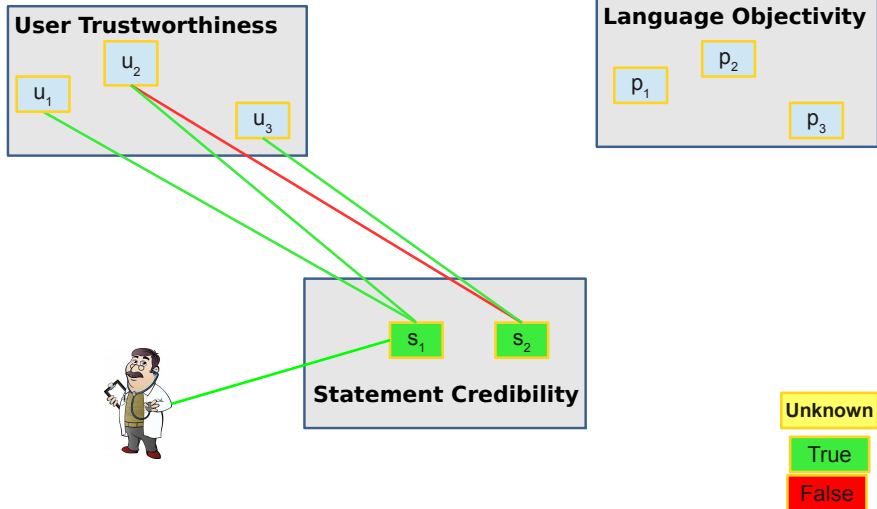


3. M-Step : Maximize log-likelihood to estimate feature weights using Trust Region Newton :

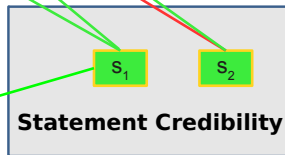
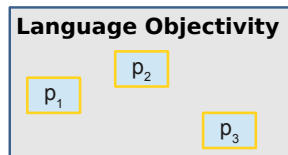
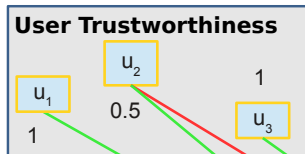
$$W^{(\nu+1)} = \underset{W'}{\operatorname{argmax}} \sum_{S^U} q(S^U) \log \Pr(S^L, S^U | P, U; W')$$



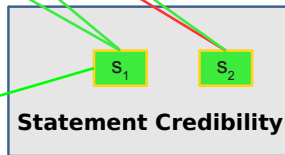
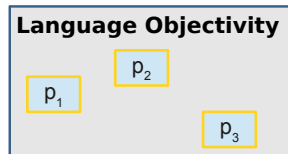
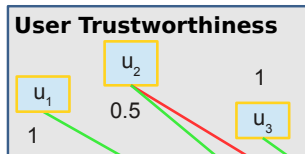
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5. Apply *E-Step* and *M-Step* until convergence

Dataset

Healthboards.com community (www.healthboards.com) with 850,000 registered users and 4.5 million messages

- ▶ We sampled 15,000 users with 2.8 million messages

Expert labels about drugs from MayoClinic portal

- ▶ 2172 drugs categorized in 837 drug families
- ▶ 6 widely used drugs used for experimentation



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Drug Statistics^a

^aData available at : <http://www.mpi-inf.mpg.de/impact/peopleondrugs/>

Drugs	Treatment For	# Users	# Posts
alprazolam	anxiety, depression, panic disorder	2.8K	21K
ibuprofen	pain, symptoms of arthritis	5.7K	15K
omeprazole	acidity in stomach and ulcers	1K	4K
metformin	high blood sugar, diabetes	.8K	3.6K
levothyroxine	hypothyroidism	.4K	2.4K
metronidazole	bacterial infection	.5K	1.6K

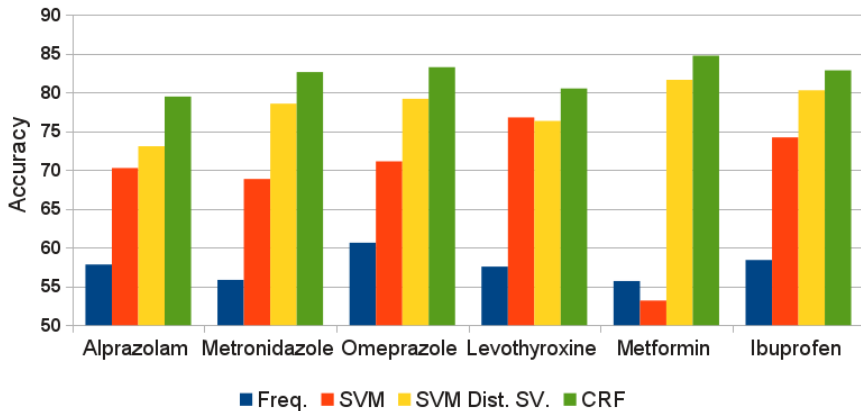


Baselines

- ▶ Frequency of statements
- ▶ SVM Classification
 - ▶ Feature vector for each statement using all our features
- ▶ SVM Classification with Distant Supervision
 - ▶ Each user, post and statement instance constitutes a feature vector
 - ▶ Aggregate labels of all such instances for a statement by majority voting



Accuracy Comparison



Use-Case: Following Trustworthy Users

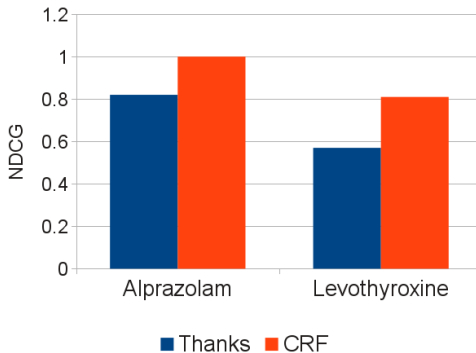
What users should I follow to get information on drug X ?

Baseline: Rank users based on *#thanks* from community



Use-Case: Following Trustworthy Users

Compare with human annotations



Conclusions

Proposed a probabilistic graphical model to jointly learn *user trustworthiness, statement credibility and language use*

- ▶ To extract side-effects of drugs from communities
- ▶ Identify expert users

Provides a framework to incorporate richer linguistic (e.g., bias, discourse) and user (e.g., perspective, expertise) features



Thank you

