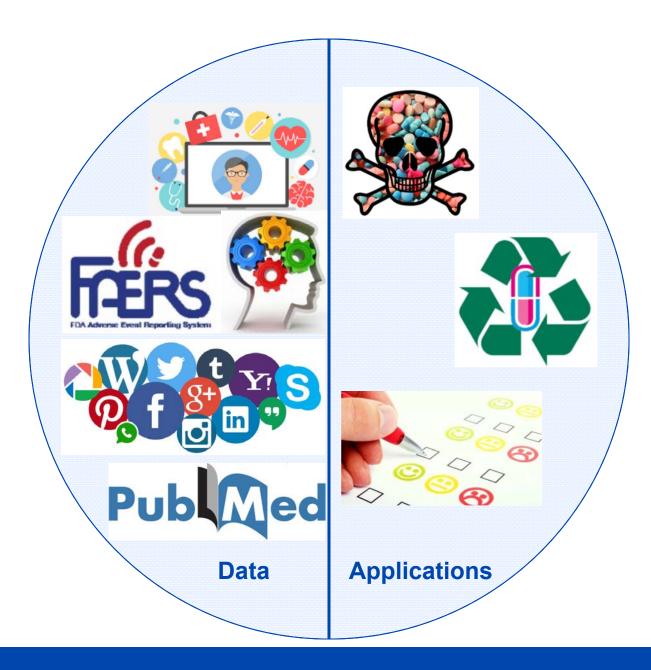
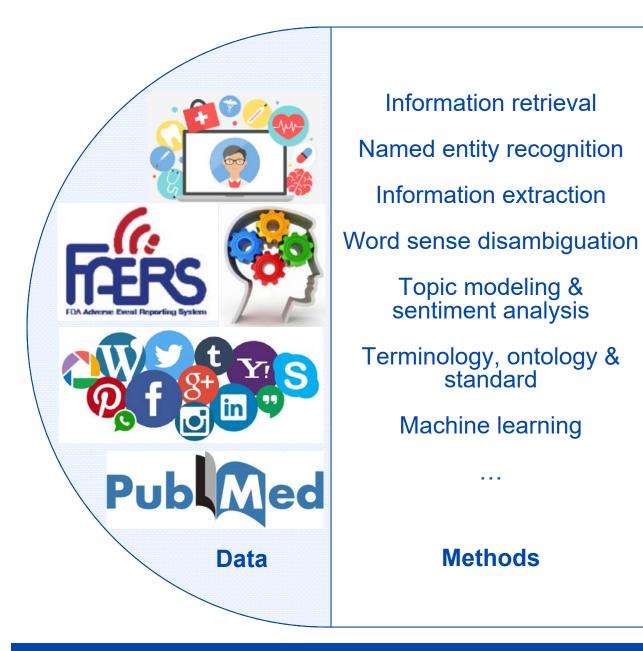
MAYO CLINIC

A Mayo Solution for Drug Safety, Repositioning and Other Patient-Reported Medication Outcomes

Lixia Yao, Hongfang Liu Division of Biomedical Statistics and Informatics Department of Health Sciences Research June 15, 2017







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Detecting Drug Safety Signals



Data Categories in the Mayo Clinic Unified Data Platform									
Patient	Provider	Location	Diagnosis (billing)						
Procedure (billing)	Flowsheet	Lab (Microbiology)	Lab (Pathology)						
Lab (specimen)	Admit, Discharge, Transfer	Radiology	Diagnosis						
Surgical Procedures	Orders	Orders Medication Administration							
Allergy	Immunization	Clinical Documents	Patient Provided Information						
Clincal Deci	sion Support	•	rastructure & mance)						

9.2 million patients at various sites of Mayo Clinic

Query via diagnostic codes or specific i2b2 inclusion & exclusion criteria

Free text and non-discrete field search supported by natural language processing with machine-learning techniques







Dr. Leonard T. Kurland helped launch **the Rochester Epidemiology Project (REP)** in 1966.

"The REP allows the study of health and disease across the entire community, from birth to death, and from primary to specialty care."

- Barbara Yawn, MD, REP Co-Principal Investigator 2006-2015



FDA Drug Safety Communication: Updated FDA review concludes that use of type 2 diabetes medicine pioglitazone may be linked to an increased risk of bladder cancer

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[12-12-2016]

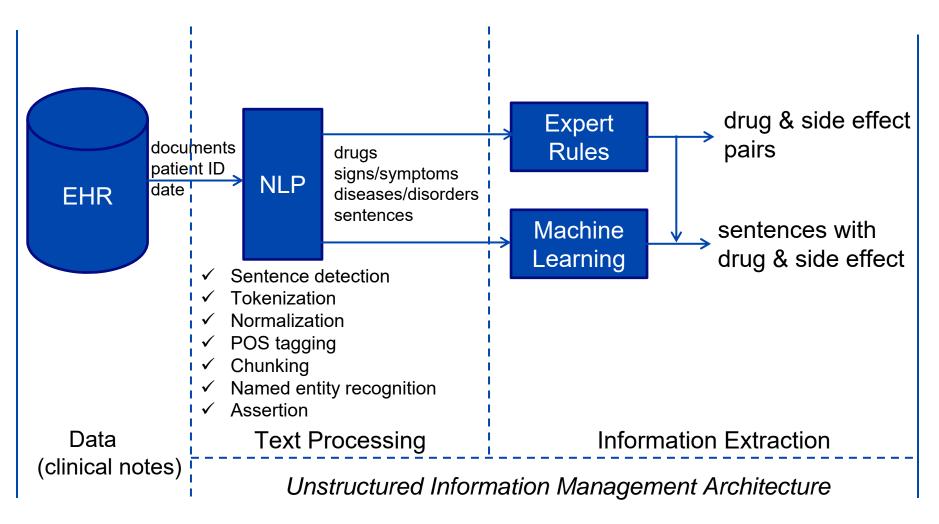
This information is an update to the **FDA Drug Safety Communication: Updated drug labels for pioglitazone-containing medicines** issued on August 4, 2011.

A really **quick** query in 10 minutes for 1/1/2010 – 12/31/2016

	Bladder cancer	No Bladder cancer
Piolitazone	54	3,860
No Piolitazone	4,235	1,041,899

Odds Ratio = 3.44 (95% CI: 2.60, 4.54), p value < 0.001





A generic NLP process for drug side effect extraction



Sohn, Sunghwan, and Hongfang Liu. "Analysis of medication and indication occurrences in clinical notes." AMIA Annual Symposium Proceedings. Vol. 2014. American Medical Informatics Association, 2014.

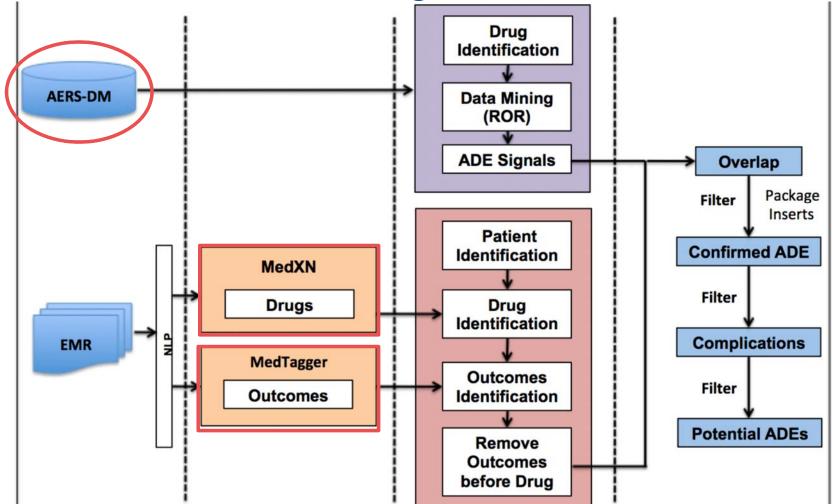
Side effect extraction on independent test set for psychiatry & psychology drugs

	Including allerg	y section	Excluding allergy section			
	Rules	Rules+C4.5	Rules	Rules+C4.5		
Precision	0.862	0.640	0.750	0.423		
Recall	0.875	0.891	0.778	0.815		
Fscore	0.868	0.745	0.764	0.557		
SE_recall*	0.867	0.893	0.806	0.861		
90% CI of SE_recall†	0.788 to 0.920	0.820 to 0.940	0.675 to 0.893	0.738 to 0.934		

*Number of side effect and drug pairs in retrieved side effect sentences/total number of side effect and drug pairs. †CI was obtained by using a modified Wald method

Sohn, Sunghwan, et al. "Drug side effect extraction from clinical narratives of psychiatry and psychology patients." Journal of the American Medical Informatics Association 18.Supplement 1 (2011): i144-i149.

Combining FDA spontaneous reports with EMR for adverse drug event detection



Wang, Liwei, etc. Discovering adverse drug events combining spontaneous reports with electronic medical records: a case study of MAYO CLINIC conventional DMARDs and biologics for rheumatoid arthritis. AMIA Clinical Research Informatics, 2017 Wang, Liwei, etc. Standardizing adverse drug event reporting data. Journal of biomedical semantics 5, no. 1 (2014): 36.

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Adverse Events for Disease-Modifying Antirheumatic Drugs (DMARDs)

Adverse Event	Number and Percentage of Cases in EMR	Report number from AERS-DM	ROR	Number and Percentage of Control in EMR
Endometrial cancer	128 (25.8%)	30	2.87	124 (12.8%)
Rhinorrhea	111 (22.3%)	355	2.81	119 (12.3%)
Productive cough	100 (20.1%)	346	2.83	102 (10.5%)
Bladder neoplasm	71 (14.3%)	17	2.60	0
Gastroduodenal ulcer	54 (10.9%)	7	3.07	0
Amyotrophic lateral sclerosis (ALS)	44 (8.9%)	26	2.71	107 (11.0%)
Sinus congestion	40 (8.0%)	237	4.87	34 (3.5%)
Sjogren's syndrome	36 (7.2%)	72	5.54	0
Respiratory tract congestion	36 (7.2%)	230	6.59	0
Bunion	34 (6.8%)	79	6.41	10 (9.7%)



Drug repositioning

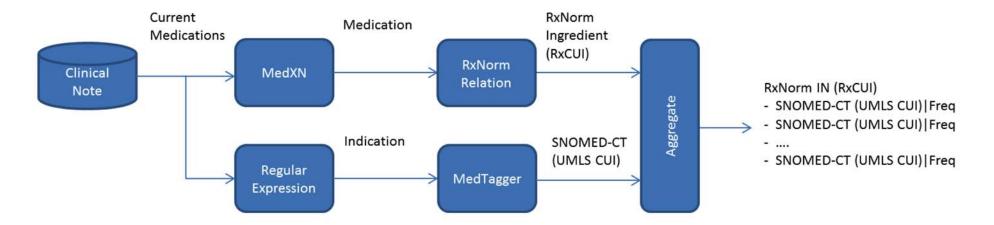


- Drug usage in reality is messy No consensus on clinically meaningful drug-indication relationships
- Literature Mining
- Social media Mining
- Electronic Health Record Validation



- Aspirin 325-mg tablet enteric-coated 1 tablet by mouth one-time daily. Indication: stroke prevention.
- Bactrim-DS 160-800 mg tablet 1 tablet by mouth one-time daily.
- Ativan 1-mg tablet 1 tablet by mouth once-a-month.
- Indication, Site, Instruction: anxiety. take 1-mg onehour every month.
- Azopt 1% drop suspension 1 drop ophthalmic two times a day. Site: Both eyes. Indication: glaucoma.

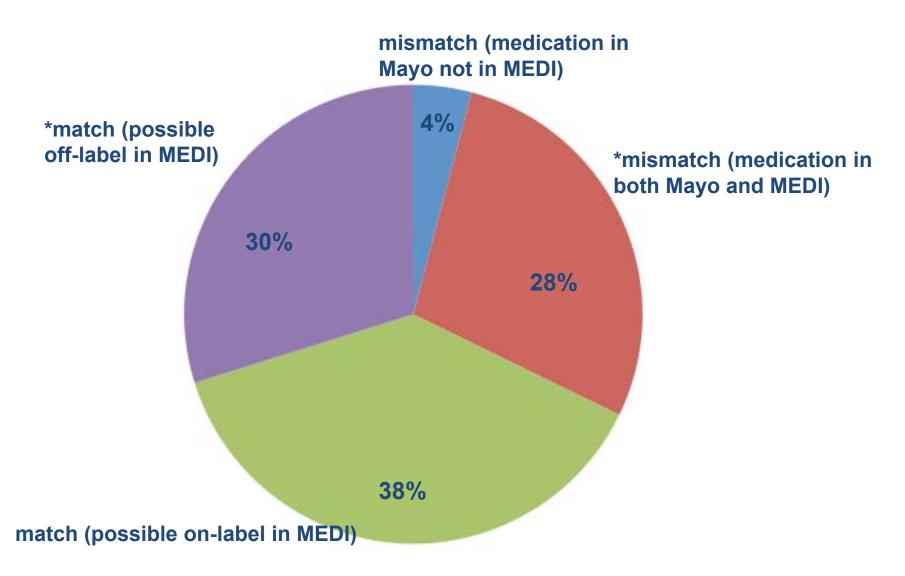




A workflow to extract medication and indication pairs from clinical notes



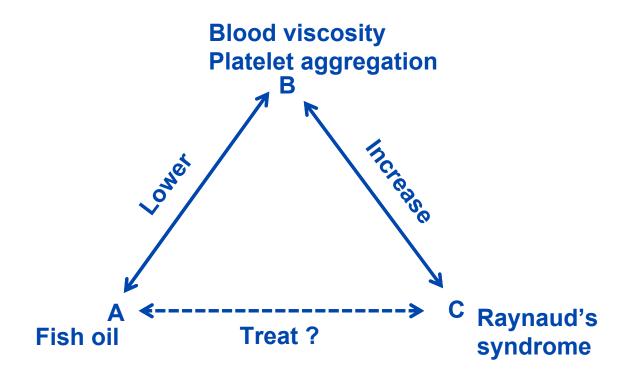
Sohn, Sunghwan, and Hongfang Liu. "Analysis of medication and indication occurrences in clinical notes." AMIA Annual Symposium Proceedings. Vol. 2014. American Medical Informatics Association, 2014.



* Possible off-label uses (log₂(# of patients)>=5, Flexible match)



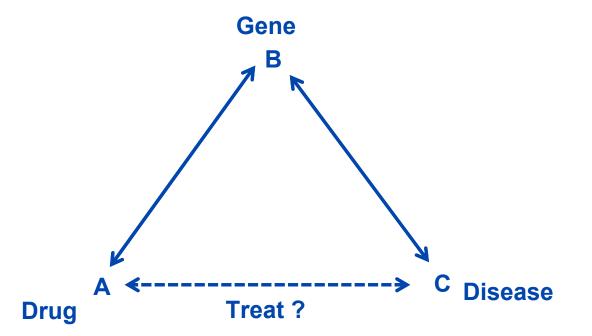
Mining literature for drug repositioning



Swanson ABC Model



Mining literature for drug repositioning

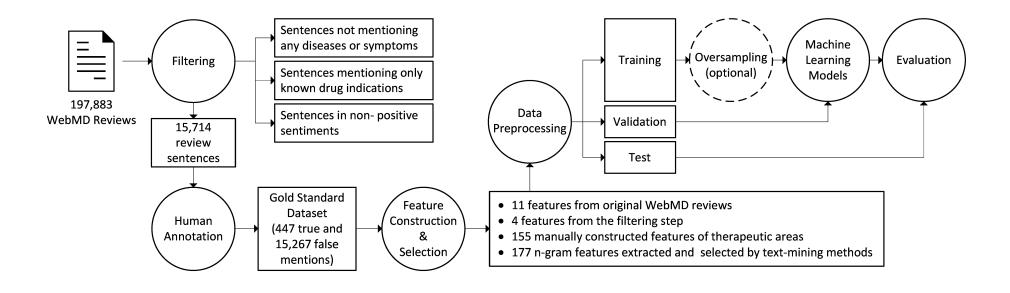


Andronis, Christos, et al. "Literature mining, ontologies and information visualization for drug repurposing." Briefings in bioinformatics 12.4 (2011): 357-368.

Li, Jiao, et al. "A survey of current trends in computational drug repositioning." Briefings in bioinformatics 17.1 (2016): 2-12. Rastegar-Mojarad, Majid, et al. "A new method for prioritizing drug repositioning candidates extracted by literature-based discovery." Bioinformatics and Biomedicine (BIBM), 2015 IEEE International Conference on. IEEE, 2015.



Mining social media for drug repositioning





Ru B, Warner-Hillard C, Ge Y, **Yao L**, "Identifying serendipitous drug usages in patient forum data – a feasibility study", The 10th International Conference on Health Informatics (HEALTHINF), Porto, Portugal (2017)

Model Performance

Madal		Test data	set	10-fold cross validation			
Model	AUC	Precision	Recall	AUC	Precision	Recall	
SVM	0.900	0.758	0.397	0.926	0.817	0.539	
SVM - Oversampling	0.893	0.474	0.429	0.932	0.470	0.620	
Random Forest	0.926	0.857	0.381	0.935	0.840	0.506	
Random Forest - Oversampling	0.915	0.781	0.397	0.944	0.866	0.530	
AdaBoost.M1	0.937	0.811	0.476	0.949	0.791	0.575	
AdaBoost.M1 - Oversampling	0.934	0.800	0.444	0.950	0.769	0.559	



True Positive Predictions

Drug	Known indications	Serendipitous usage	Example	# of references
Metformin Metformin Syndrome, etc.		Obesity	I feel AWFUL most of the day, but the weight loss is great.	>10
Tramadol	Pain	Depression , anxiety	It also has helped with my depression and anxiety.	1
Acetaminophen & oxycodone	Pain	Depression	While taking for pain I have also found it relieves my major depression and actually gives me the energy and a clear mind to do things.	1
Bupropion	Depression, attention deficit & hyperactivity disorder	Obesity	I had energy and experienced needed weight loss and was very pleased, as I did not do well on SSRI or SNRIs.	4
Ondansetron	Vomiting	Irritable bowel syndrome with diarrhea	A lot of people have trouble with the constipation that comes with it, but since I have IBS-D (irritable bowel syndrome with diarrhea), it has actually regulated me .	1
Desvenlafaxine	Depression	Lack of energy	I have had a very positive mood and energy change, while also experiencing much less anxiety.	

Validating drug repositioning signal using EHR

Data Summary	Vanderbuilt	Mayo Clinic
Unique patient count (May 2013)	2.2 M	7.4 M
Cancer patients (excluding skin cancer and age <18) diagnosed between 1995 and 2010, based on Tumor Registry	44,257	102,546
Cancer patients from above, but excluding congestive heart failure and chronic kidney disease before tumor diagnosis date	42,165	96,169
With Type 2 Diabetes	5,796	8,939
Taking Metformin	2,218	3,029
Taking other oral diabetes drug	903	1,629
Taking insulin monotherapy	377	1,462
Ineligible & Excluded	2,298	2,819
Without Type 2 Diabetes	28,917	73,138

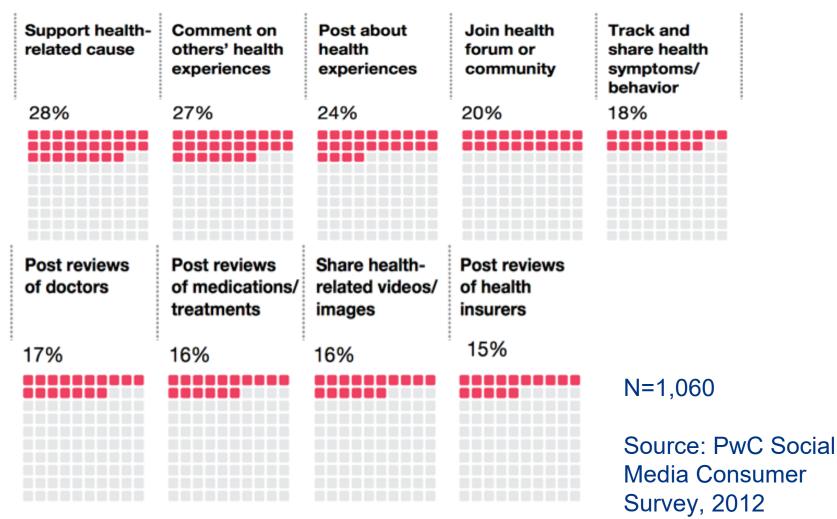


Xu, Hua, et al. "Validating drug repurposing signals using electronic health records: a case study of metformin associated with reduced cancer mortality." Journal of the American Medical Informatics Association 22, no. 1 (2015): 179-191.

Using stratified Cox proportional hazard models, we found that cancer patients using metformin was associated with decreased mortality, compared to cancer patients (diabetic and non-diabetic) not on metformin.



Mining patient-reported medication outcomes





Social Media and Metrics Diseases and Drugs		WebMD			PatientsLikeMe					YouTube	Twitter	
		Effectiveness	Ease of Use	Satisfaction	No. of Reviews	Perceived Effectiveness	Side effects	Adherence*	Financial Burden*	No. of Reviews	No. of Posts	No. of Posts
	Albuterol	3.87	4.39	3.55	112	4.06	2.08	4.01	4.29	137	2859	11381
Asthma	Ipratropium	4.17	4.17	3.75	12	4.13	1.33	4	3.95	8	140	271
	Prednisone	4.00	3.92	3.32	367	4.19	2.89	4.68	4.17	48	8569	971
Cystic	Azithromycin	I	-	-	-	2.91	1.73	4.55	4.65	11	31	90
Fibrosis	Ivacaftor	-	-	-	_	-	-	-	-	-	818	5060
	Meloxicam	3.39	4.18	3.11	202	3.20	1.89	4.47	4.6	15	558	67
Rheumatoid Arthritis	Prednisone	4.11	4.31	3.61	229	4.03	3.20	4.68	4.17	63	10064	670
	Sulfasalazine	3.19	3.28	3.17	65	2.44	3.03	4.45	3.98	77	581	53
	Bromocriptine	2.23	3.08	2.15	13	-	-	-	-	1 - 1	29	214
Type 2 Diabetes	Insulin	3.50	4.22	3.35	265	4.21	1.60	4.73	4.2	106	11401	16308
Diabetes	Metformin	3.29	3.91	2.93	1302	3.69	2.30	4.41	4.17	331	7504	4042
Weighte	d Average:	3.52	4.00	3.14	-	3.74	2.32	4.42	4.18		-	-

* The average drug adherence and burden (cost) ratings on PatientsLikeMe may not be disease specific. In case that a drug has multiple indications, the adherence and burden ratings are consolidated across all indications.

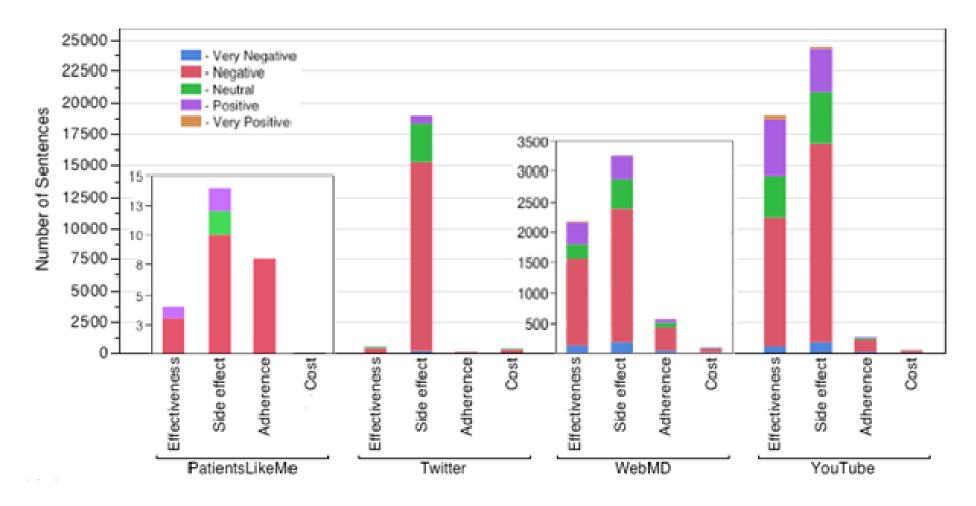


Ru, Boshu, Kimberly Harris, and Lixia Yao. "A content analysis of patient-reported medication outcomes on social media." Data Mining Workshop (ICDMW), 2015 IEEE International Conference on. IEEE, 2015. (best paper award)

Content Type	Example	Source				
	day 001 - not so bad so far, but really only 18 hours in . Asthma acting up a little. Taking Advair. Need to refill Albuterol script.	Twitter				
Effectiveness	Only been on it for 8 days but I noticed <i>relief from pain</i> right away. However, yesterday I was more short of breath than usual and my <i>blood pressure was high</i> , tightness in my chest so I will stop taking it to see if that stops.					
	make sure it is plain claritin. not claritin D or anything else. <i>Plain claritin has loratadine</i> . that is fine but anything else can kill it. <i>the regular claritin has no side effects.</i>	YouTube				
Side effects	Bubble gut to me is extreme gas <i>Very extreme gas, be careful for the first few weeks of use</i> . Don't put yourself in situations like elevators or long car rides before your body adjust to the med.					
	It is so very <i>easy to use</i> and the needle is tiny and does not hurt at all.	WebMD				
Adherence	They divide pills down to <i>easier to take</i> portions or fine tune the dose to better fit the patient. Have you ever tried to use a breakfast drink with the pills? Sometimes it helps pills slide down MOBETTA!	YouTube				
Cost	Private <i>insurance pays</i> for young Shan 's `miracle drug'; Kelsa can't <i>afford</i> it.	Twitter				
COST	Being so close to Mexico allowed me to get my inhalers <i>cheaper</i> and quicker.	WebMD				
Disease comorbidity &	Doctor prescribed this after I stopped taking Plaquenel due to stomach upset. In addition to RA I have a history of IBS, sensitive stomach and I have tolerated this medication well. <i>It has greatly improved my IBS while moderately improving my RA pain</i> . Ony side effect is feeling full, thirsty and ocassional gut pain.	WebMD				
repositioning	Personal side effect: Very sensitive to sun. <i>This clears up my bronchial spasms so quickly and as a bonus, it clears up my eczema!</i> I have asked my doctor to prescribe it regularly for my skin condition and he says there are too many side effects. too bad. It is a wonder drug. I was able to breathe very well.					



Sentiment Analysis





Learned lessons

- Individual data source has inherent bias
 - Findings can be inconsistent and hard to get buy-in from multiple stakeholders
- Domain experts needed to be engaged for clinical meaningfulness
- A need to move from counting co-occurrence to capturing the context and synthesizing the information
 - What + Who, When, Where, Why, How



Acknowledgment







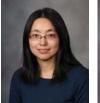


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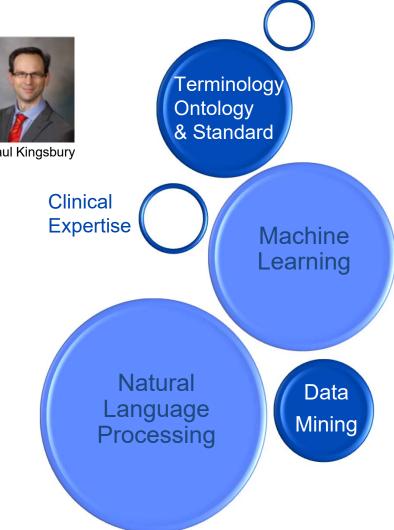
Riea Moon



David Chen



Na Hong





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Questions?

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