NLP Applications for Enhancing Clinical Decision Making



April 24, 2012

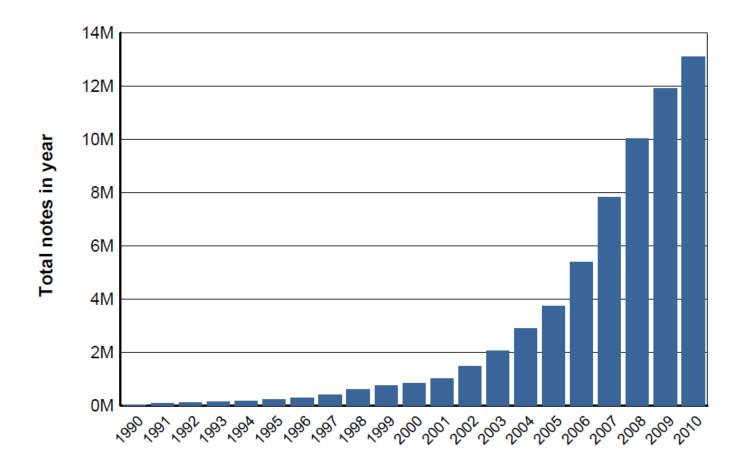
Clinical Informatics R&D, Partners Healthcare Division of General Internal Medicine and Primary Care Brigham and Women's Hospital Harvard Medical School





Why Do We Need NLP in CDS?

NLP in CDS



Much information that could support CDS is textual and therefore cannot be leveraged by a CDS system without NLP





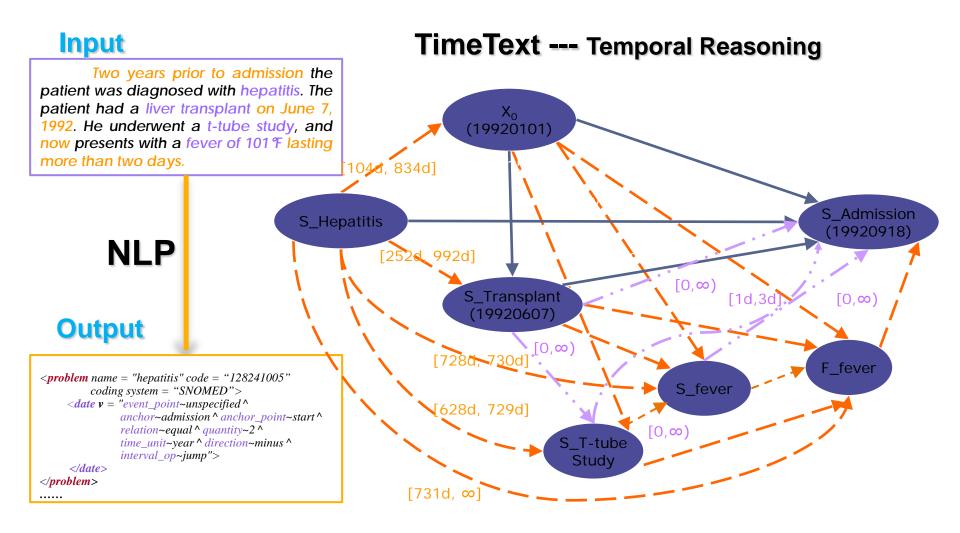
NLP ----> CDS

My journey in and perspectives of the two areas



Natural Language Processing

NLP in CDS



Zhou L, Friedman C, Parsons S, Hripcsak G. System architecture for temporal information extraction, representation and reasoning in clinical narrative reports. AMIA Annu Symp Proc :869-73.



Clinical Decision Support

- CDS Interventions: reminders, alerts, infobutton, order sets, etc.
- Inpatient/outpatient EHRs: CPOE/eprescribing, eMAR, etc.
- Structured knowledge representation and knowledge sharing
- Service-oriented architecture



- Improving patient safety
 - Improving medication management
- Enhancing EHR functions
 - Facilitating computerized provider order entry
- Reducing health care cost
 - Identifying high-risk, high-cost patients prospectively



Patient Safety & Medication List

- Medication errors can cause injuries, are common, and are very costly
 - Adverse drug events and medication errors are estimated to cost the US health care system \$177 billion annually
- Medication lists within patients' EHRs are often *outdated*, *incomplete* or *inaccurate*, which is a major cause of medication errors
 - Active medications are often not added in a timely manner to the structured medication list
 - Wagner and Hogan found discrepancies between the number of medications that patients reported taking (5.7) and those listed in their EHRs (4.7)
 - Outdated medications are frequently not deleted
 - 67.4% of medications were still active one calendar day after their inactive status was documented in the clinic notes



Medication Reconciliation

NLP in CDS

		LM	t OMA90 MEDICATIONS - Microsoft Internet Explorer provided by Partners	HealthCare Sy	stem			_1
м	eds fro		٩	A			JLS9	1
ource	Medicator		Adapt Depiter In a star in the Original	Oustan	Dent	-		IOSPITALISTS
UH	Acetylsa →		Select Desktop Pt Chart: Medications Oncolog	y Custom	Repo	ons Ad	min Sign Results ? Resourc	e Popup
		Discharge Medication Reconciliation Allergies: IV Contrast - HIVES, / QUETIAPINE - HYPEROLYCEMA, Discontinue Print						
IR	Acetyl unit stre							
	Atorvas	1000000			and the second se		- Martin	
IH	 Atorva Lipitor 				-	<	Acetylsalicylic acid	
ĸ	unit stre	l	81 MG (81 MG TABLET take 1) PO QD	09/04/08	Î	Verity	81 MG PO QD	
	take: 1					+	Insulin glargine	0
R	Ciprofic Ciprof	ا ۳	35 UNITS SCIQHS	09/04/08	Î	Modify	40 UNITS SC QPM	
	MG Q12H					+	Metformin	0
н	☐ Citalepr → ① Cital					Add	1500 MG PO QD	
	MG QD		Mettormin xr	0	-	<		
R	Citalor MG QD	ļ	1500 MG (500MG TAB.SR 24H take 3) PO QD	09/04/08	Î	Verity		
	Folic Ar				-	< Nerth		
H	-> 🚯 Folio	ļ	45 MG (45MG TABLET take 1) PO QD	10/02/08	Î	Verity		
R	E Folic J		Taper (5MG TABLET) PO x 5 days	09/15/08	î	< Venty	Prednisone Taper PO Give 30 MG q 24 H X 1 D0	ORE than Give 25 MG
н	E Furosi		Taper (Smo TABLET) POX3 days	03/13/00			Taper PO One Stimo q 24 HAT DO	SE tien one 25 mo
R	E Furosi		Alprazolam	0	-	+	Alprazolam	• Gen
	MAy go Insulin	ļ	1MO TABLET take 1 Tablet(s) PO TID x 30 days, DR. URIZAR'S DEA	09/04/08	Î	Modify	0.5 MG PO TID PRN Anxiety	
R	(insulin		C [®] <u>Mirtazapine</u>	0	-	+	Mirtazapine	0
	U QAM at prefiled	l	30MG TABLET take 2 PO QHS x 30 days	09/04/08	Î	Modify	30 MG PO QHS	
	Nicotine					+	Acetaminophen	
н	Nicotir For pati					Add	650 MG PO Q6H	
	Dancreli		Reconcile in Full Re	econcile in Full	& Sign		Reconcile in Part Reconci	le in Part & Sign <u>Ç</u> ar
н	Pancr 201 PO 4	e)			021035			A Local intranet

Bobnipperl, DiesignDanelloppplemberftattooloviibininapplicentiotromid ansolicialteelcosed/toefsatiditatipportedicteridisciplinary needicationecofterilladispritelfolistshargen. ihtely hat 2011; all 84(3) record devine record terribations. JAMIA. 2006; 13(6):581-92

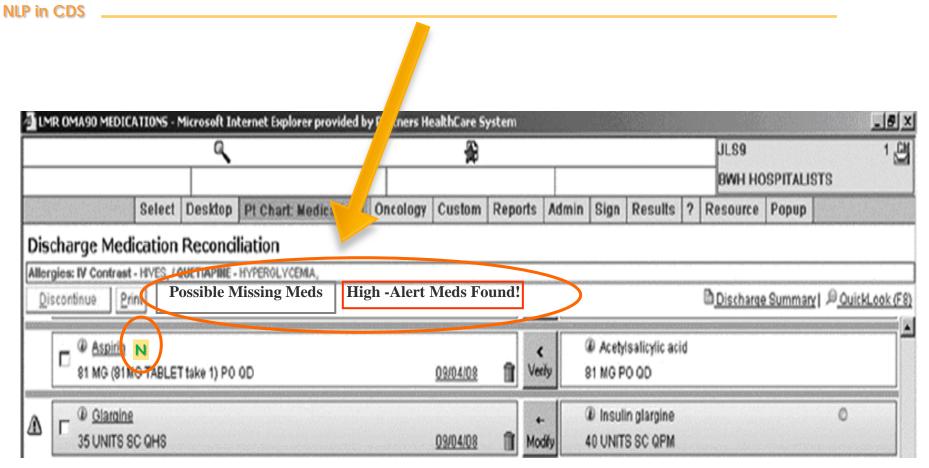


Medications in Clinical notes

- 30% of active medications mentioned in notes were missing from patients' medication list
 e.g., "trazodone and paxil prescribed by psychiatrist".
- Clinicians often need detailed or additional information beyond the medication list in order to make judgments, changes and other decisions
 - e.g., history and progress of the disease, consultation notes from medical specialists

Medications from NLP Output

Stroke



Zhou et al. Improving Medication Reconciliation Using Natural Language Processing. The 2011 AHRQ Annual Conference, September 2011, Bethesda, MD



NLP in CDS

Data interoperability / terminology standard

- Medication list may be coded using an institutional or commercial terminology; most existing NLP systems encode clinical text using standard terminologies
- Code information using multiple medical terminologies
- Conduct dynamic or static mapping as needed
- System / data integration
 - Integrate NLP system with other EHR applications
 - Conduct data integration, aggregation and summarization
 - Conduct reasoning (inference)

* Zhou, et al. Using Medical Text Extraction, Reasoning and Mapping System (MTERMS) to Process Medication Information in Outpatient Clinical Notes. AMIA 2011 Annu Symp Proc: 1639-48.



NLP in CDS

Improving patient safety Improving medication management

Enhancing EHR functions

Facilitating computerized provider order entry

Reducing health care cost

Identifying high-risk, high-cost patients prospectively



Free-text Order Entry

- **7.1%** of medication order entries are free-text
- 9.3% (2,412 out of 26,001) hypoglycemic medication orders were entered using free-text for 2,091 patients
- 75.2% of these free-text entries have an exact name match in the medication dictionary
 - the remaining 24.8% of the free-text entries could be coded if specific formulation information was also provided
- 17.4% of free-text hypoglycemic medication order entries included misspellings
 - e.g., "Glimepiride", "Glipizide", "Glyburide", "Humalog", "Humulin"

Zhou et al. How Many Medications are Entered through Free-text in EHR? – a Study on Hypoglycemic Agents. AMIA Annu Symp. 2012.



Free-text Order Entry - CDS

- CDS is not triggered when a medication order is entered as free-text
 - 92 drug-drug interaction (DDI) alerts were not triggered due to free-text entries, affecting 84 different patients
 - 196 patients who had a free-text hypoglycemic order entry also had the same exact drug entered as a structured and coded order during the study period
 - 10% had identical drug entries active in their medication list at the same time
- Only 25.9% of these patients had diabetes recorded in their problem list
- These CDS aspects are critical to patient safety
 - If unintentionally bypassed due to free-text medication order entries may result in potential harm to the patient



NLP ---> CPOE/CDS

NLP in CDS

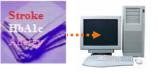
- Advanced search function
 - It should not be limited to only detecting exactly spelled medication name
 - It should provide a relevant and smart list (not just a long list to sort through)
- Spelling error detection / correction
 - e.g., suggest a correction, autocorrect
- Efficient user-interface to address workflow issues
 - Auto-fill features
 - "Favorite" list or pick-list (where free-text entries should be monitored)
 - Avoid navigating through multiple screens
- Speech recognition



Improving patient safety
 Improving medication management

NLP in CDS

- Enhancing EHR functions
 Facilitating computerized provider order entry
- Reducing health care cost
 - Identifying high-risk, high-cost patients prospectively



NLP → Risk-identification & Stratification

NLP in CDS

- There is a great need to minimize the cost of care delivered while still meeting quality incentives
 - 5% of patients generally account for up to 50% of costs
 - e.g., prevent readmissions by identifying high-risk, high-cost patients prospectively

How will NLP help?

- Most current risk assessment methods use claims and structured data
- Clinical narrative reports contain rich information, but are untapped
- Combine structured data with data extracted from free-text using NLP to identify target patients
- Employ machine learning methods (classification techniques and other probabilistic models) to stratify patients
- Provide CDS and make recommendations



Summary and Discussion

- Many opportunities to apply NLP to enhance CDS
 - Little things can make a big difference
 - Usability matters a lot
 - Make it easy for clinicians to do the right thing
- Many challenges to tackle
 - Speed is everything (subsecond "screen flips")
 - If it takes too long to work, it will be useless
 - Optimize system performance, particularly speed
 - Anticipate needs and deliver in real time
 - Bring information from free-text to clinicians at the time they need it
 - Fit into the user's workflow
 - e.g., develop an efficient user interface to present NLP output

* Bates & Middleton, et al Ten Commandments for Effective Clinical Decision Support: Making the Practice of Evidence-based Medicine a Reality. JAMIA. 2003.



Summary and Discussion

- Many challenges to tackle (cont.)
 - Standards and system interoperability
 - Service oriented architecture
 - Others: organizational issues; diverse clinical domains; users with diverse roles, background and needs
- Other requirements
 - Simple interventions work best
 - Simplify and condense NLP output and make it useful
 - Continuous monitor impact, receive feedback and make improvement

* Bates & Middleton, et al Ten Commandments for Effective Clinical Decision Support: Making the Practice of Evidence-based Medicine a Reality. JAMIA. 2003.

Thanks!

Izhou2@partners.org

