

#### **AMIA 2012 Annual Symposium**

## Auditing Medical Records Accesses via Healthcare Interaction Networks

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(Joint work with Bradley Malin and Steve Nyemba)





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Jail Time for Malicious Accesses

undates without the email



## Patient information needs to be protected from insiders

- Traditional security practices (e.g., role-based access control) are insufficient to ensure EMR security
  - Common for >100 employees to access a patient's medical record during their visit
  - Often difficult to determine who the members of a care team are and who will need access to what information at which time



# EHRs have adopted collaborative capabilities to facilitate interaction between teammates and coordinate care

- We hypothesize that HCO departments will exhibit predictable interaction behavior
- Our goals:
  - 1. Investigate if such behavior exists
  - 2. If so, determine if it is stable
    - If stable interactions become unstable 

       associated patients will be anomalous

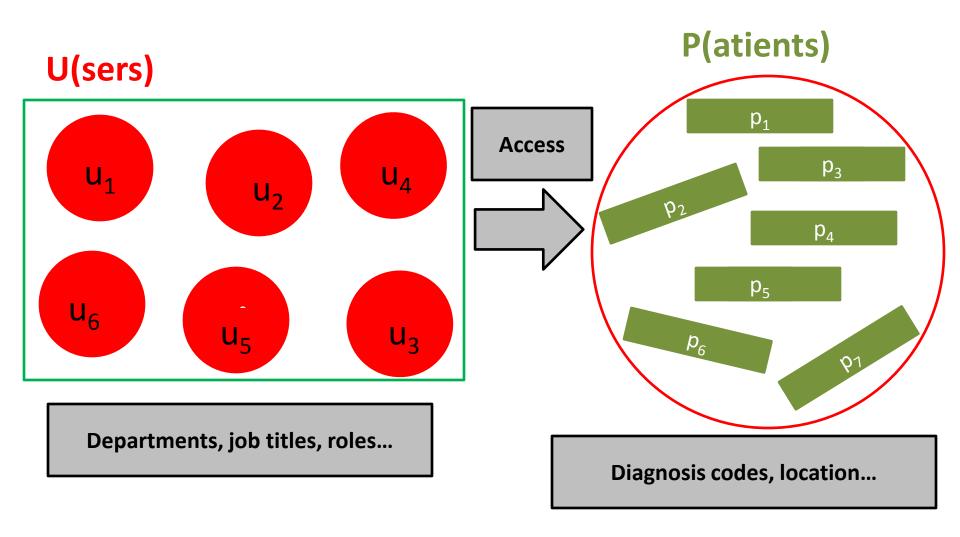


# The dependent relations between green departments and red departments are very low

Our goal is to retrieve the dependent relations of departments and determine whether the dependencies among departments touching that patient are expected?

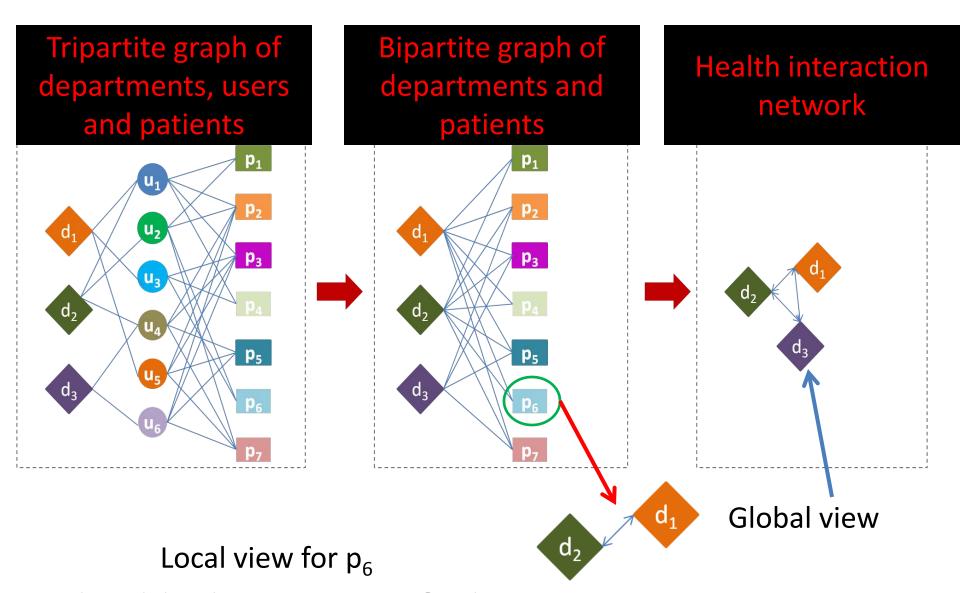


## Two general objects of EHR access logs





#### Healthcare Interaction Networks





#### Where are We Going?

#### **A Global Network of Departments**

Two metrics: certainty and reciprocity

Stable status in terms of the two metrics

#### **Local Network-for a specific patient**

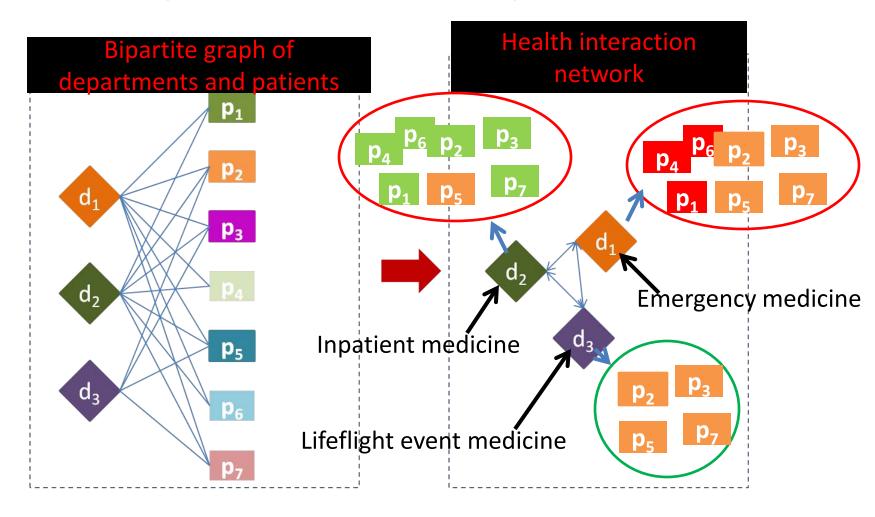
Two metrics: local network score and reciprocity

#### **Application of the Networks**

Detecting patients with anomalous medical records accesses



#### **Certainty to Model Relationship of Global Network**

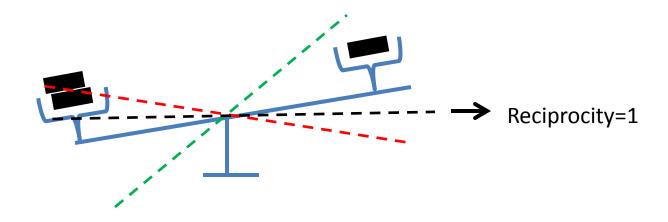


Cert(Lifeflight event medicine  $(d_3)$ ->Emergency medicine  $(d_1)$ ) = 4/4

Cert(Inpatient medicine( $\frac{d_2}{d_2}$ )->Inpatient medicine( $\frac{d_2}{d_2}$ )) = 6/7



## Using reciprocity to characterize the mutual interaction between all pairs of departments in the global network



Inpatient Admin -> VUH Admitting 0.75

VUH Admitting -> Inpatient Admin 0.12



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#### Dataset used for this study

- Vanderbilt University Medical Center "StarPanel"
- 3 months in 2010
- Arbitrary Week
  - ≈ 9,200 users
  - ≈ 99,000 patient records
  - $\approx$  400,000 accesses
  - $\approx$  450 departments



### Although the relations of the network are very unbalanced, the unbalance is stable over time

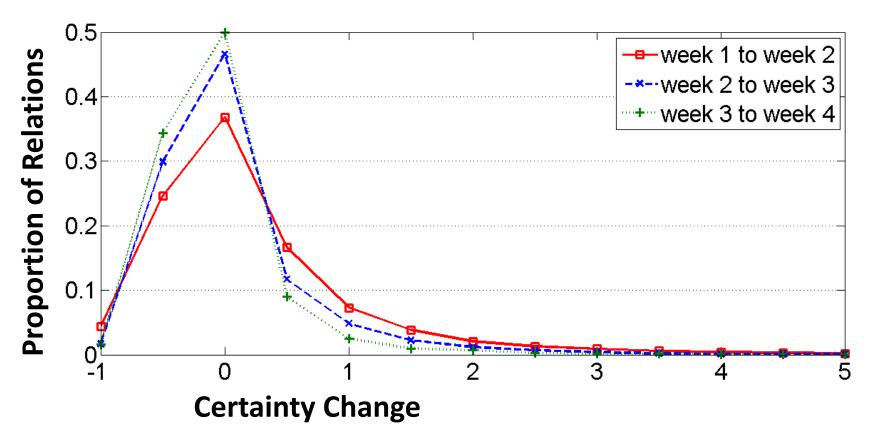
Time	Week 1	Week 2	Week 3	Week 4
Reciprocity	0.267	0.2814	0.2858	0.2871

$$(0.2814 - 0.267) / 0.267 = 0.05$$

Week 1 to week 2



### The changes become smaller over time (centralization: green > blue > red)



Degree of relations between departments changes little over time >82.5% of the change resides in [-0.25, 0.25]



## Strong relations between VUMC departments over a four week period

<b>Department</b> $(d_i)$	<b>Department</b> $(d_j)$	Min Certainty	Max Certainty			
Intradepartmental Relations						
4East OB/GYN	4East OB/GYN	0.74319	0.7669			
Adult Emergency Medicine	Adult Emergency Medicine	0.74024	0.78453			
Cancer Infusion Center	Cancer Infusion Center	0.73171	0.844			
8N Inpatient Medicine	8N Inpatient Medicine	0.7197	0.80909			
Newborn Nursery	Newborn Nursery	0.70406	0.72727			
Interdepartmental Relations						
DOT Radiology	Orthopaedics	0.99621	1			
Nursing Education and Development	Medical Information Services	0.95833	1			
Main OR - Trauma/Renal	Medical Information Services	0.94444	1			
Life Flight Event Medicine	Emergency Medicine	0.90805	1			
Emergency Medicine Admin	Adult Emergency Medicine	0.91489	0.94186			



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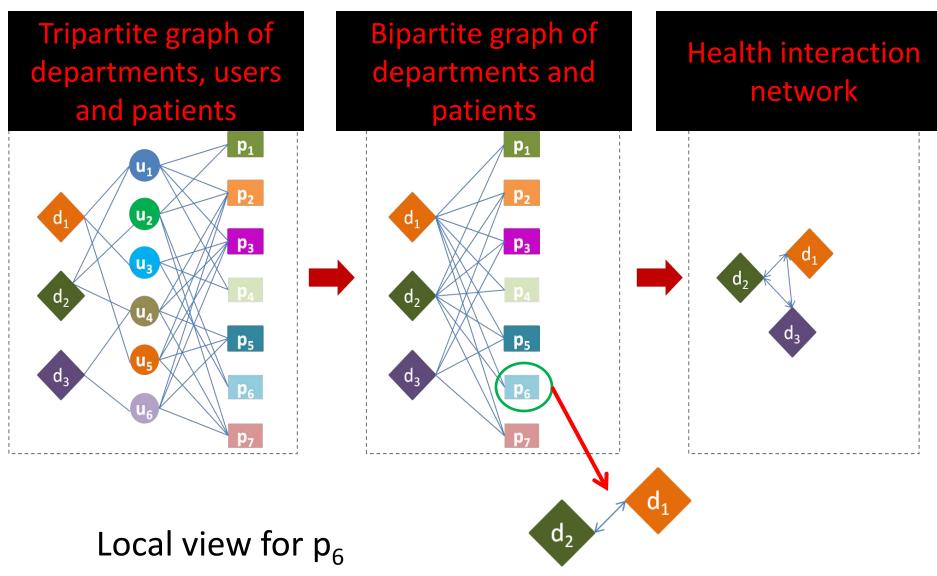
Two metrics: local network score and reciprocity

#### **Application of the Networks**

Detecting patients with anomalous medical records accesses

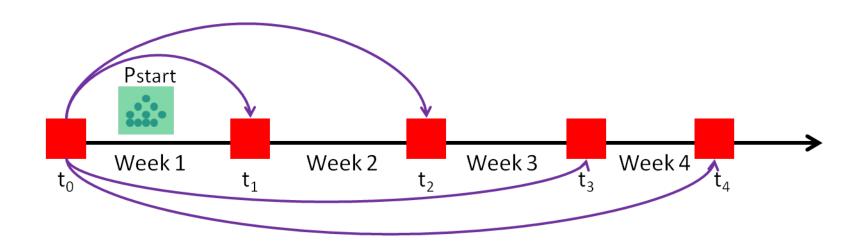


#### Healthcare Interaction Networks





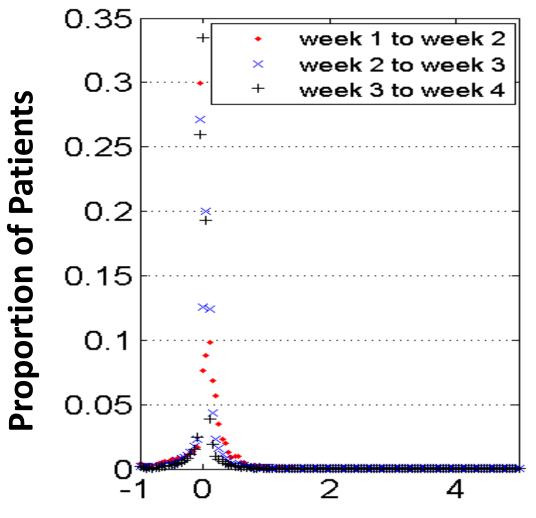
## **Evolution of Local Networks in Terms of Local Network Score and Local Network Reciprocity**



Each point in P<sub>start</sub> corresponds to a local network



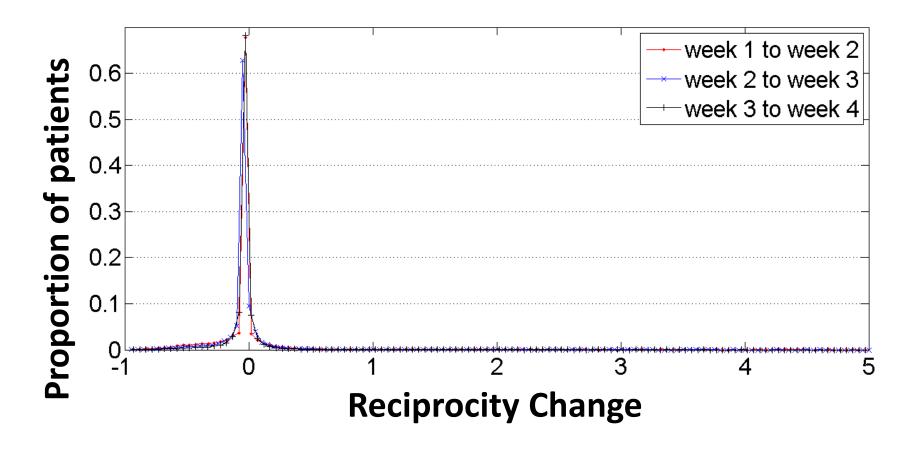
### Over 98% of patients are normal because they exhibit a score change < 0.05



**Local Network Score Change** 



## Approximately 99% of patients are normal because they have a change of reciprocity <0.1





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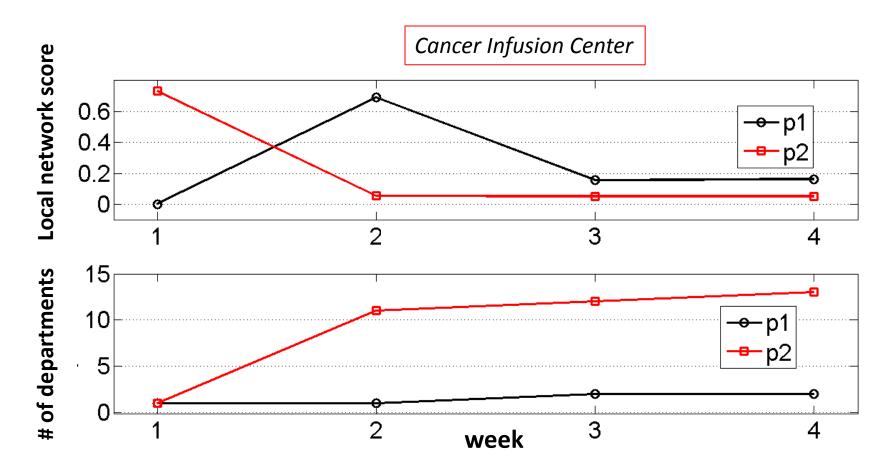
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### p2 has -0.93 change of local network score and -0.79 change of local reciprocity from the 1<sup>st</sup> to the 2<sup>nd</sup> week



Breast Center, [Anonymized Street Location], Care/Eskind Diab Acces, Disease Management Service, Eskind Diabetes - Adult, Free Stipends, Internal Medicine, VIM, VMG Physician Billing Services, Vanderbilt Home Care Primary

Auditing

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#### Conclusions

- We hypothesized an HCO would exhibit strong stability
  - confirmed by our experiments
- We can characterize how strange a patient's local network appears
  - Two groups of patients; those with small changes in local network score and reciprocity score and those with significant changes
  - The changes in the latter group do not justify the claim that the patient has been intruded upon, but may provide a reason for an investigation that incorporates more nuanced domain knowledge



#### Some Limitations

- Global and local networks appear to represent the business processes of HCO departments
  - however, such claims must be confirmed with employees knowledge about the working of the medical center and its affiliated clinics
- Need to specialize tool to account for semantics of patients
  - Patient: {Diagnosis, Procedure, Demographics, Residence, physical location in a hospital}
  - Incorporating semantics about the patient, p<sub>2</sub> in the last figure may have no intrusion; rather it is likely a complex cancer patient, which could be confirmed by inspection of clinical documents in the medical record



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#### Questions? Comments?

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Health Information Privacy Lab: http://www.hiplab.org/

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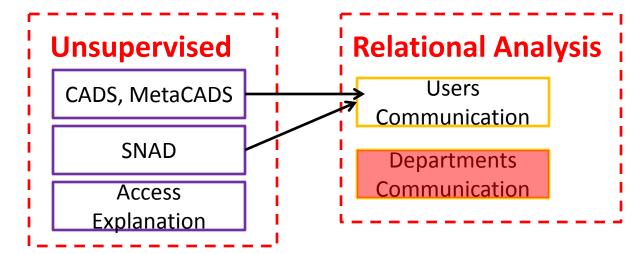


#### Auditing Models on EHRs systems

#### **Supervised**

Logic regression and SVM

require adjudication from privacy experts and pre-existing knowledge of what constitutes a suspicious access



Our works will investigate whether the relations of departments are stable or not and how the dynamic characteristics could be applied to assess if the set of specific

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