



VANDERBILT

# Creating Interpretable Collaborative Patterns for Auditing

You Chen

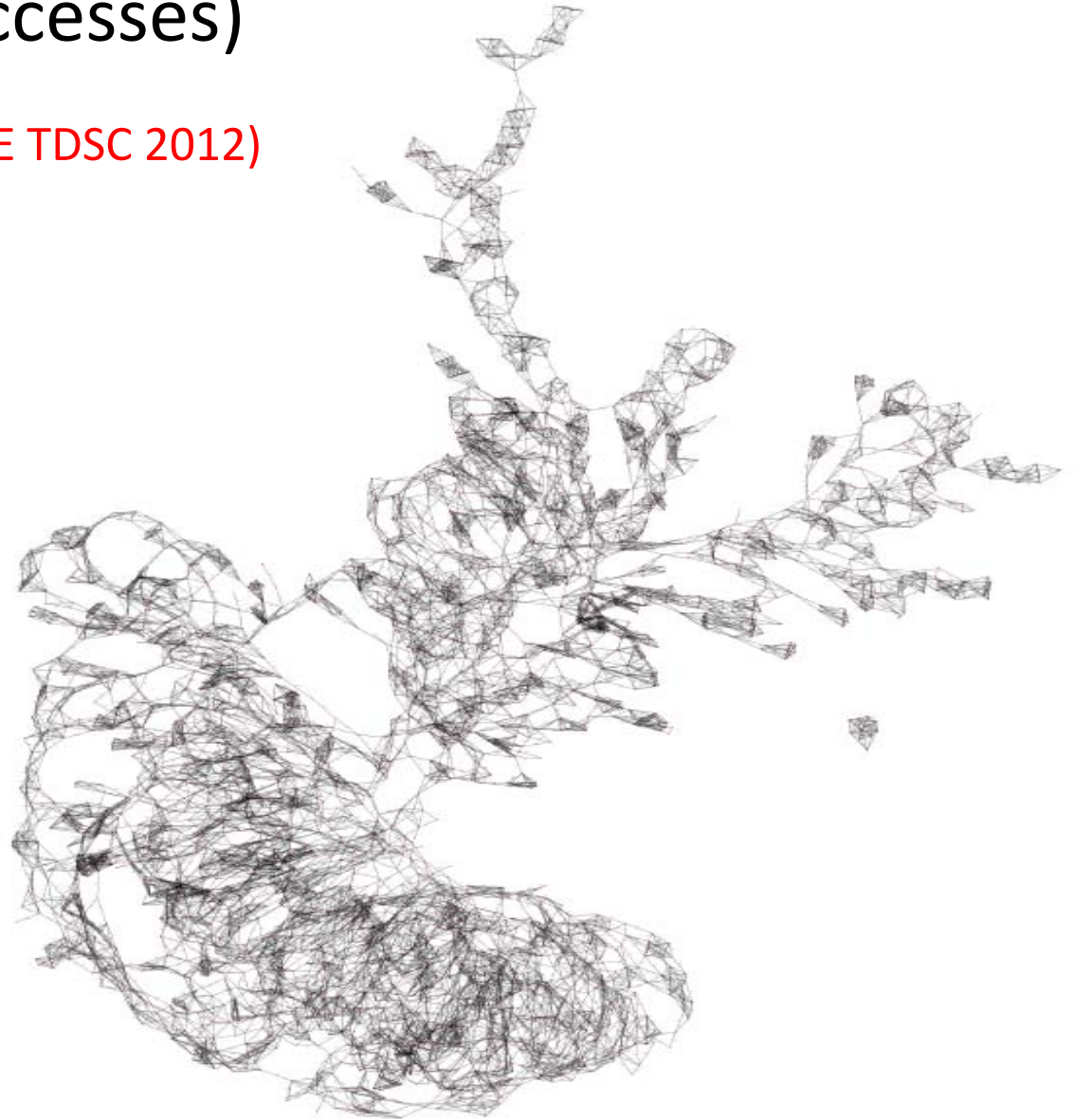
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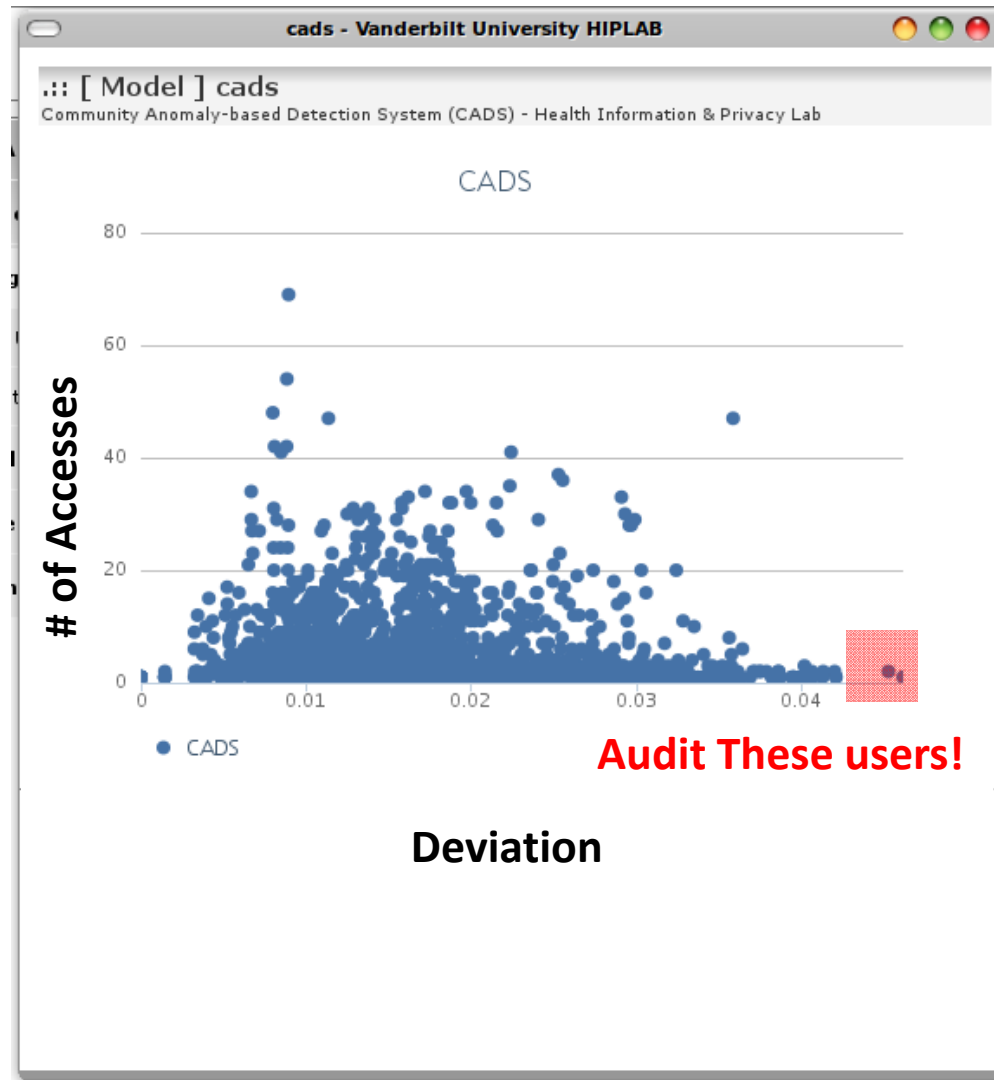
<http://hiplab.org/~ychen>

# 6-Nearest Neighbor Network-Vanderbilt Medical Center (1 day of accesses)

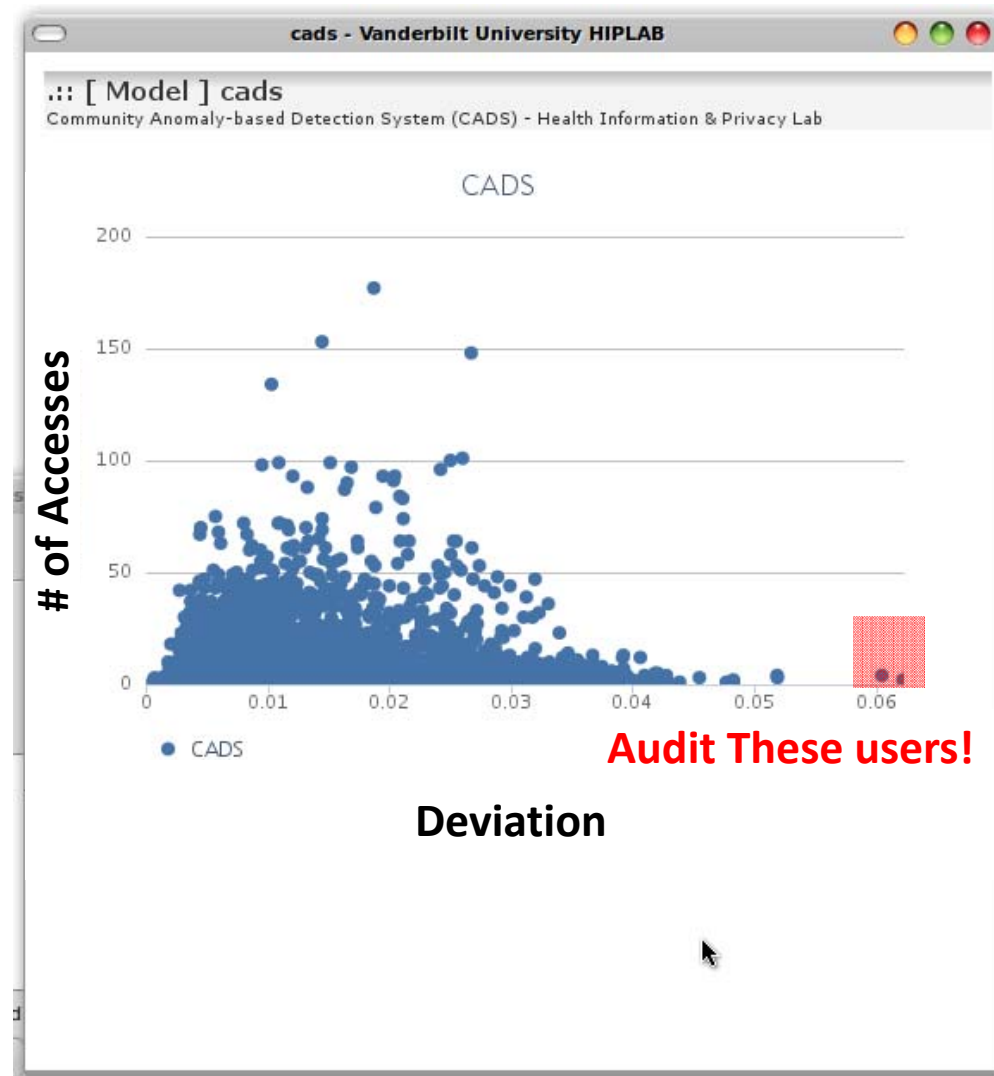
(Chen, Nyemba, & Malin – IEEE TDSC 2012)



## CADS on Vanderbilt Dataset

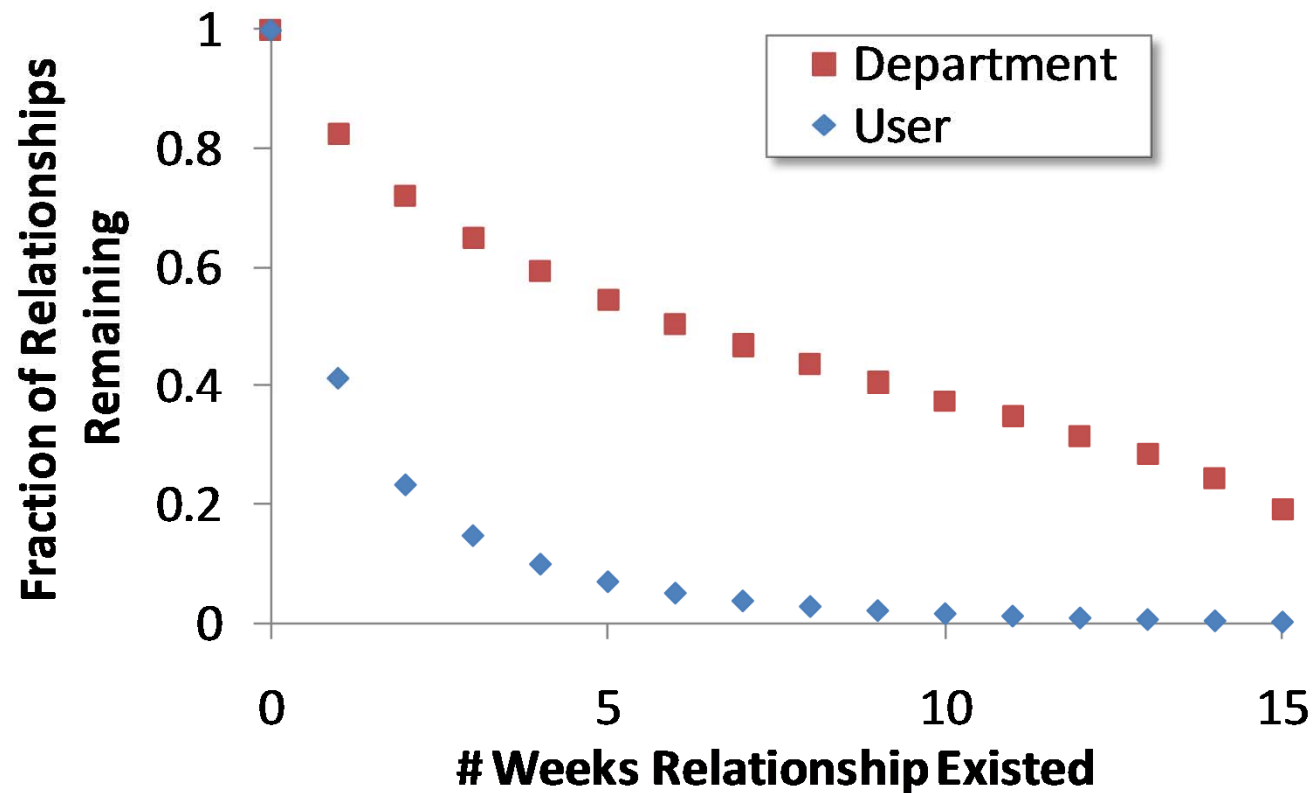


## CADS on Northwestern Dataset



# But Relationships Decay...

(Malin, Nyemba, & Paulett – JBI 2011)



◆ EMR <user, user> relationships

■ EMR <department, department> relationships

# Department Level

Auditing Medical Record Accesses  
through Department Interactions

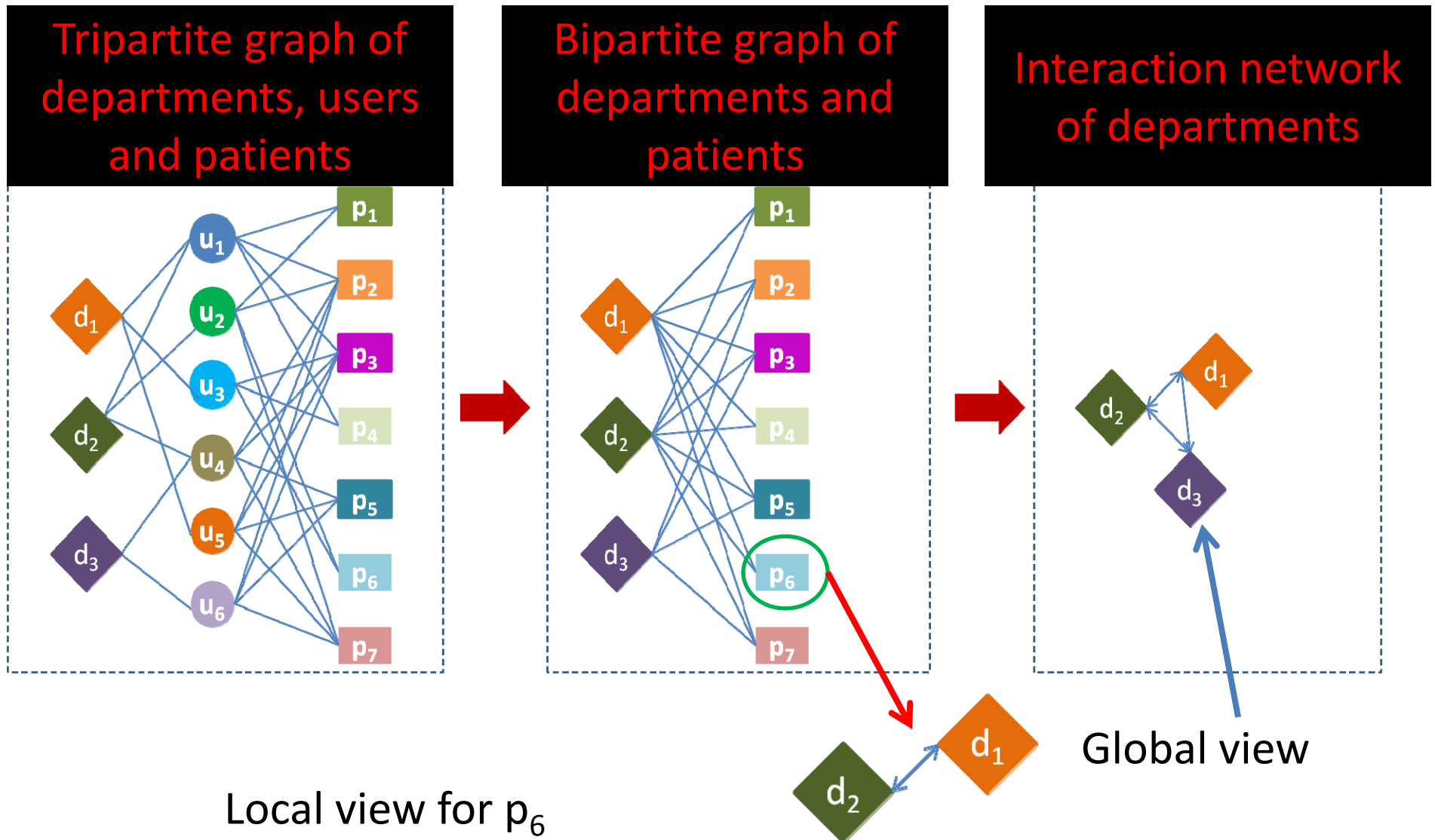
# Hospital Departmental Relations Can Be Inferred

(Chen, Nyemba, & Malin - AMIA 2012)

- Probability department  $d_i$  accesses a patient's record, given department  $d_j$  accessed the record.

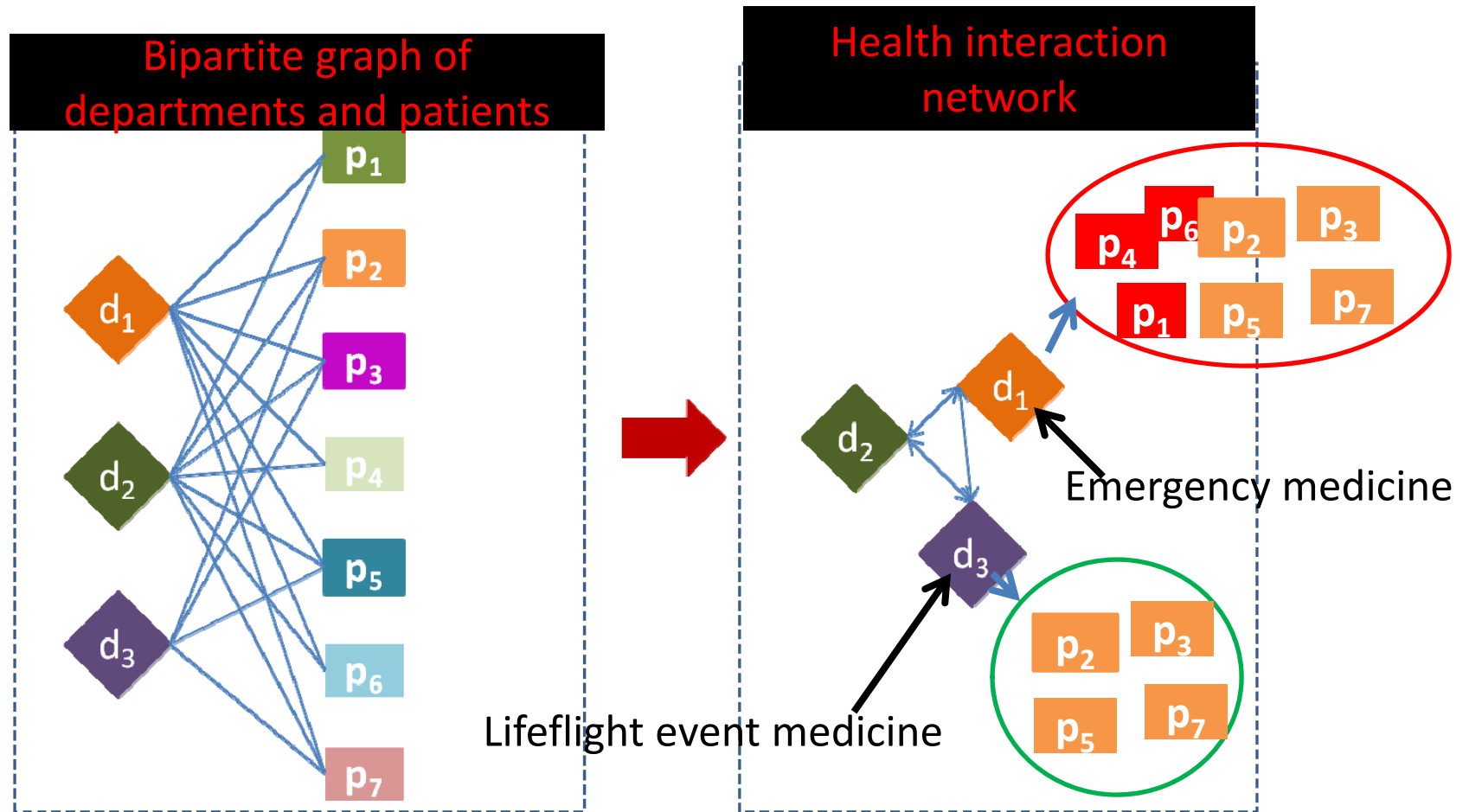
Department ( $d_i$ )	Department ( $d_j$ )	Min Certainty	Max Certainty
<i>Intradepartmental Relations</i>			
4East OB/GYN	4East OB/GYN	0.74319	0.7669
Adult Emergency Medicine	Adult Emergency Medicine	0.74024	0.78453
Cancer Infusion Center	Cancer Infusion Center	0.73171	0.844
8N Inpatient Medicine	8N Inpatient Medicine	0.7197	0.80909
Newborn Nursery	Newborn Nursery	0.70406	0.72727
<i>Interdepartmental Relations</i>			
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

# Organization Level-Department



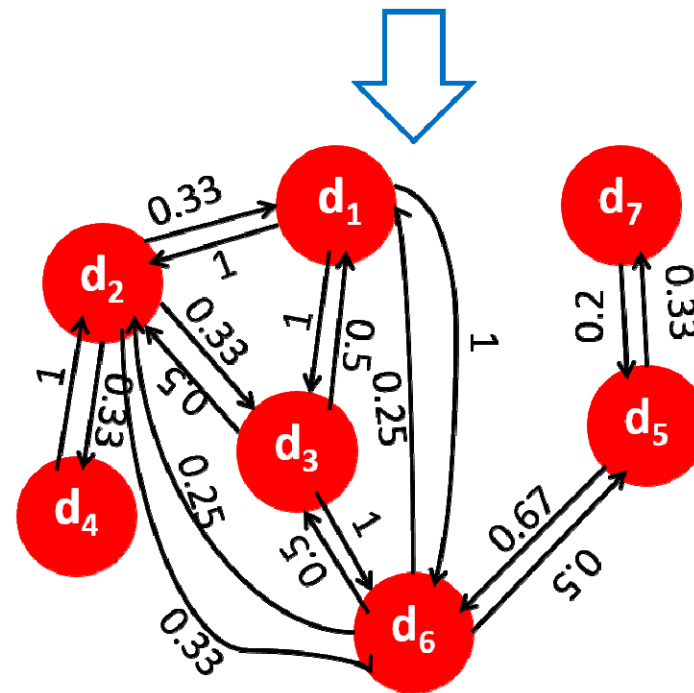
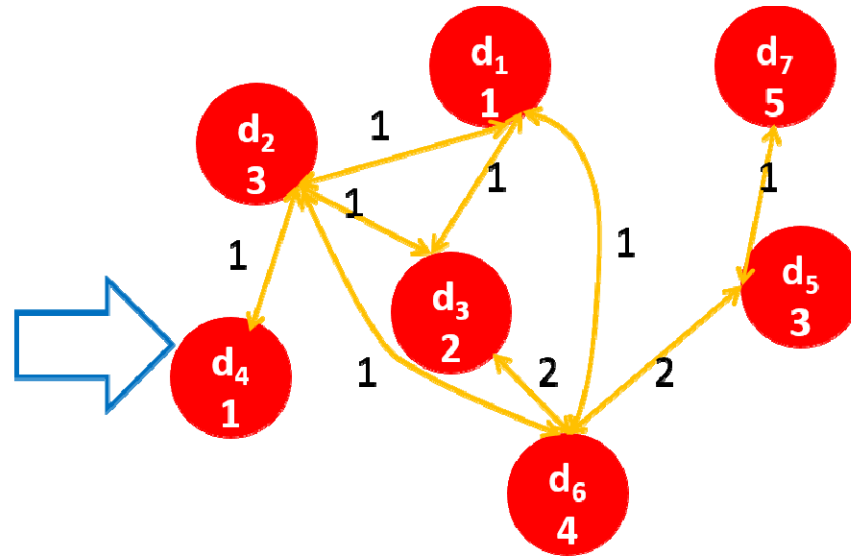
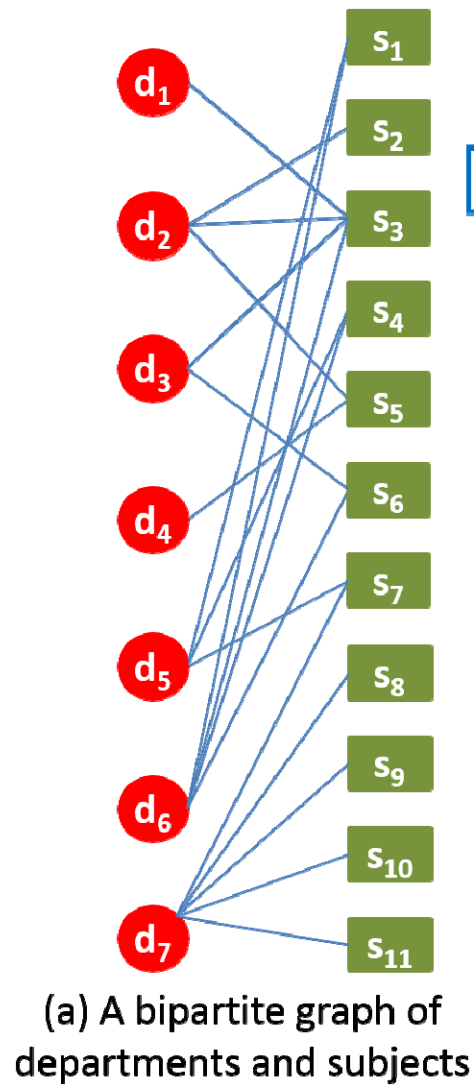


# Certainty to Model Relationship Among Departments

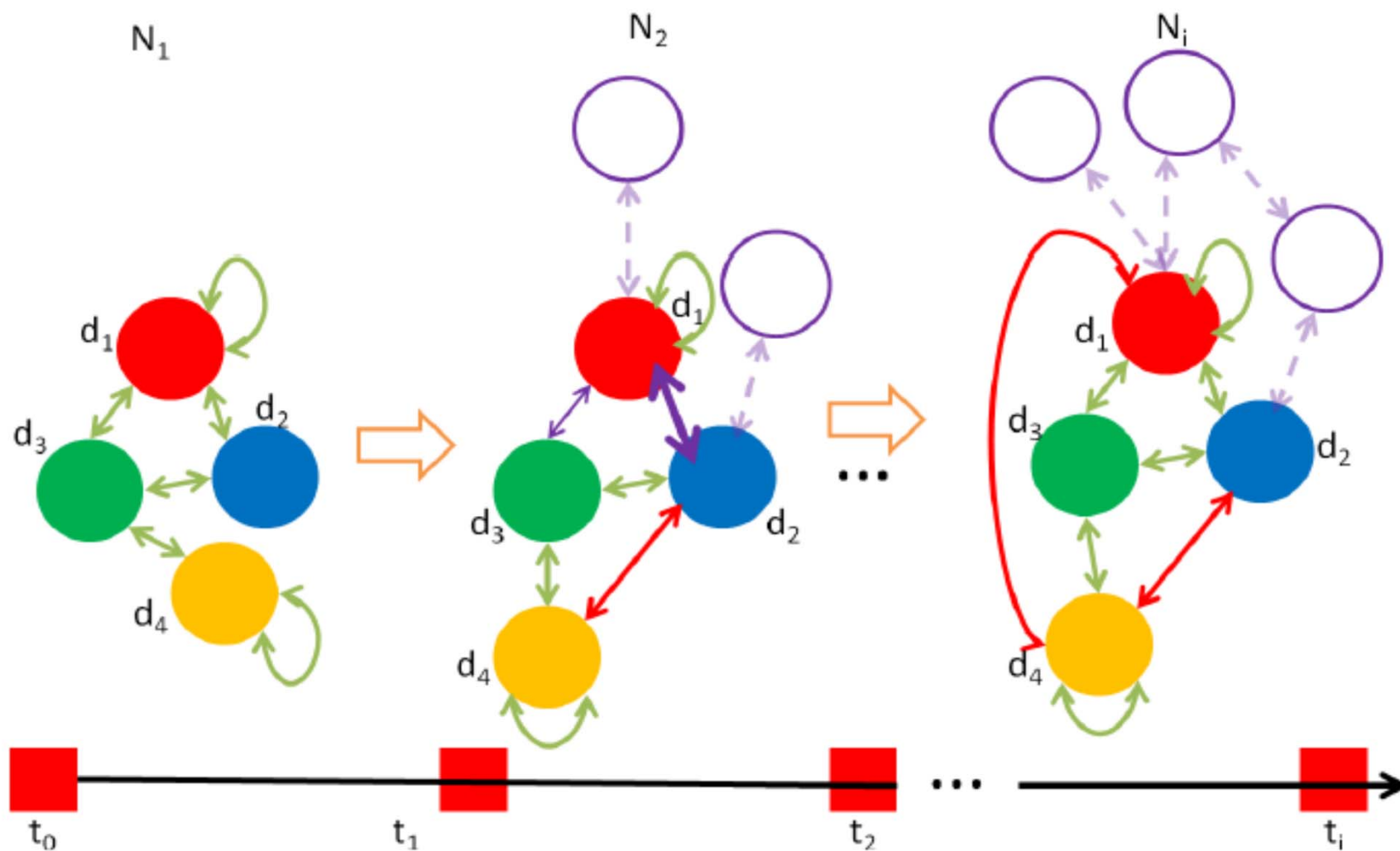


$\text{Cert}(\text{Emergency medicine } (d_1) \rightarrow \text{Lifelight event medicine } (d_3)) = 4/7$

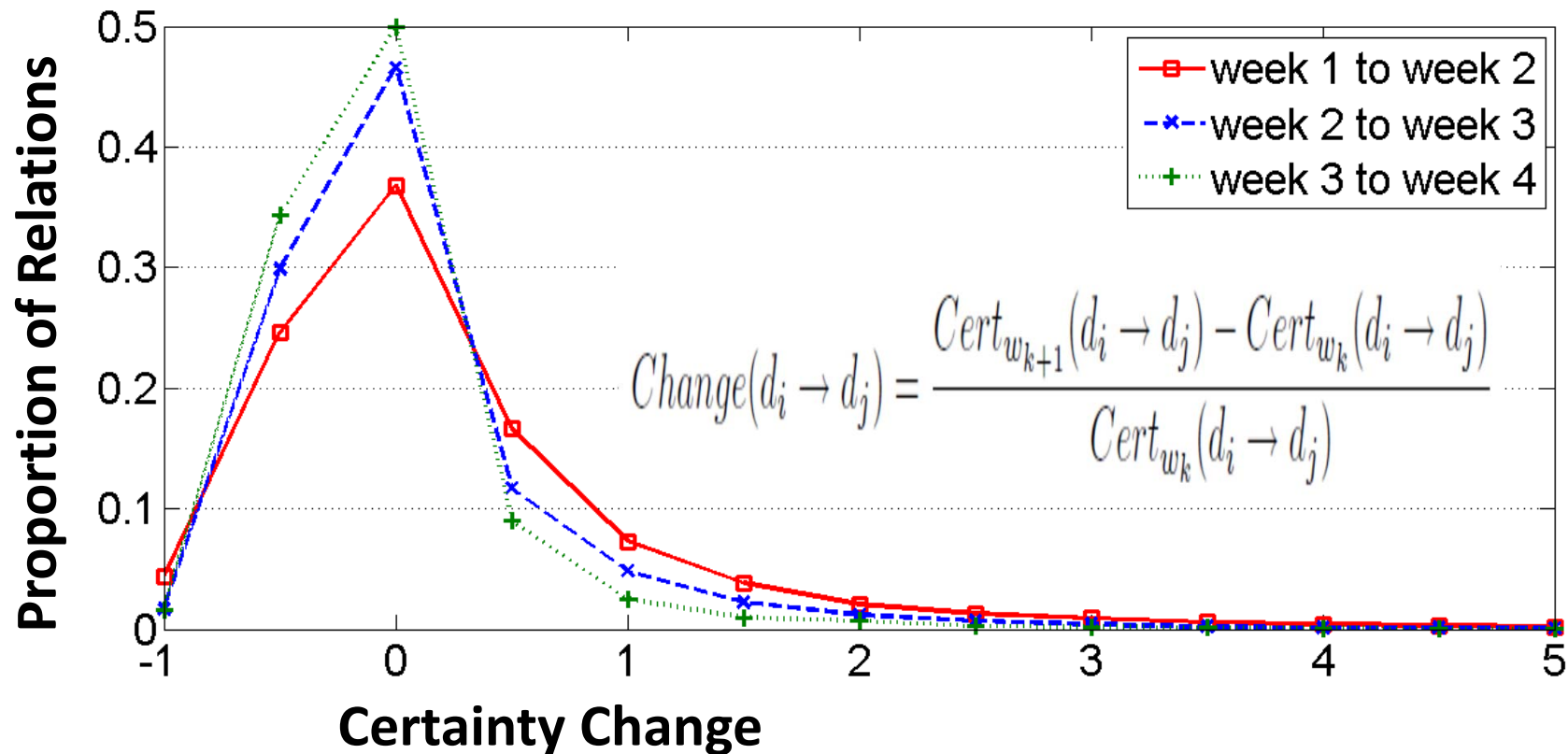
$\text{Lifelight event medicine } (d_3) \rightarrow \text{Cert}(\text{Emergency medicine } (d_1)) = 4/4$



# Evolution of A Global Network Over the Time

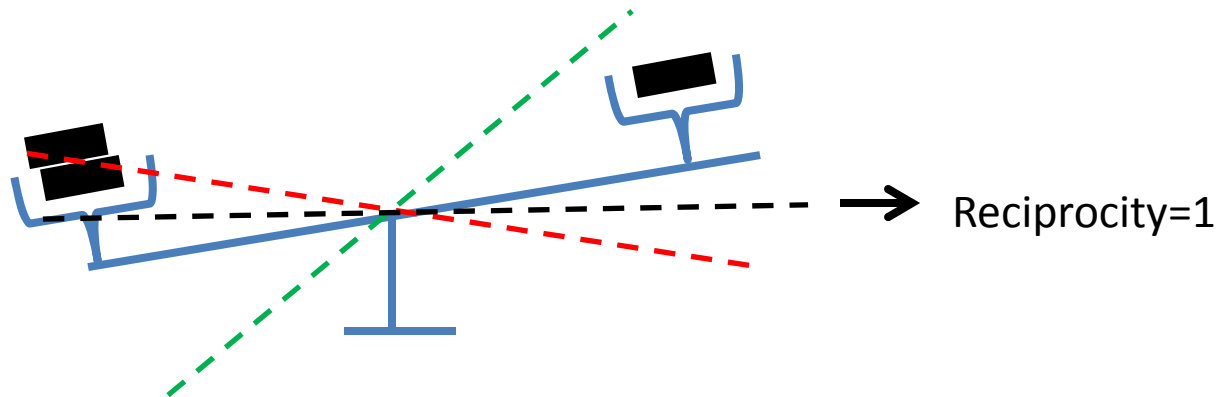


**The changes become smaller over time  
(centralization: green > blue > red)**



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

# 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

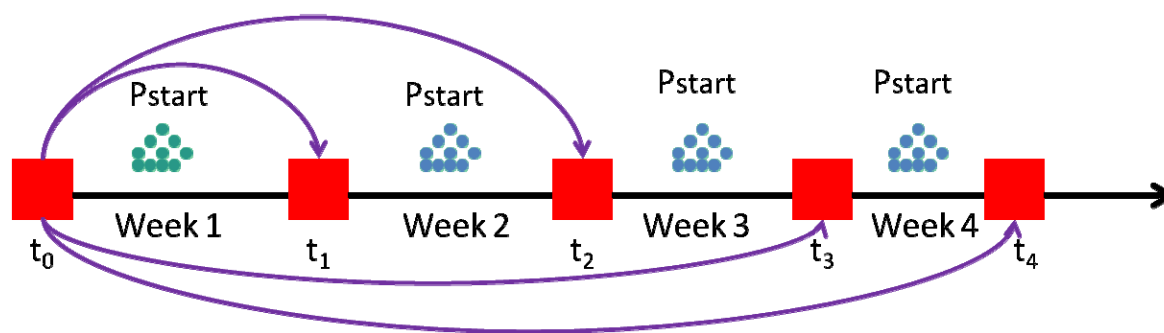
**Although the relations of the network are very unbalanced, the unbalance is stable over time**

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

$$Reciprocity = \frac{\sum_{\forall d_i, d_j \in D, i \neq j} |(Cert(d_i \rightarrow d_j) - a) \times (Cert(d_j \rightarrow d_i) - a)|}{\sum_{\forall d_i, d_j \in D, i \neq j} (Cert(d_i \rightarrow d_j) - a)^2}$$

$$a = \frac{\sum_{\forall d_i, d_j \in D, i \neq j} Cert(d_i \rightarrow d_j)}{2 \times |E|}$$

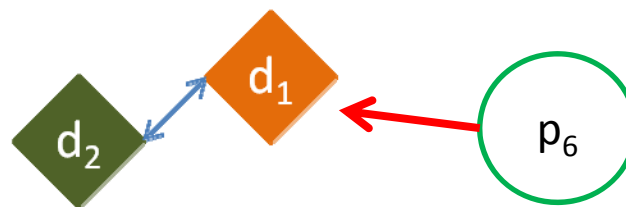
# Evolution of Local Network Relations Can be Used Detect “Strange” Behavior



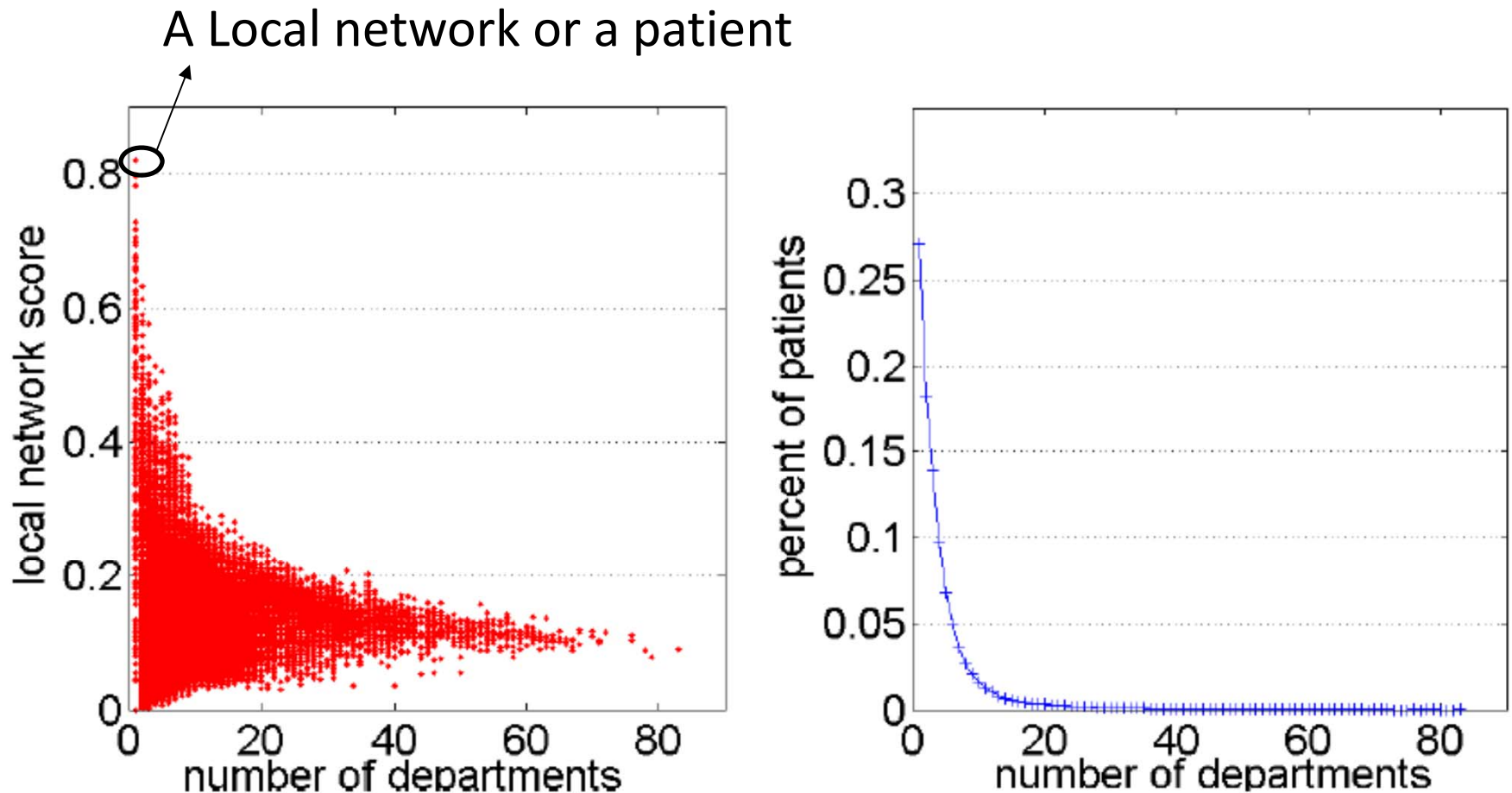
Each point in the  $P_{start}$  corresponds to a local network

A Local network for  $p_6$

$$Score(p_k) = \frac{\sum_{\forall d_i, d_j \in D_k} Cert(d_i \rightarrow d_j)}{2 \times |E_k|}$$

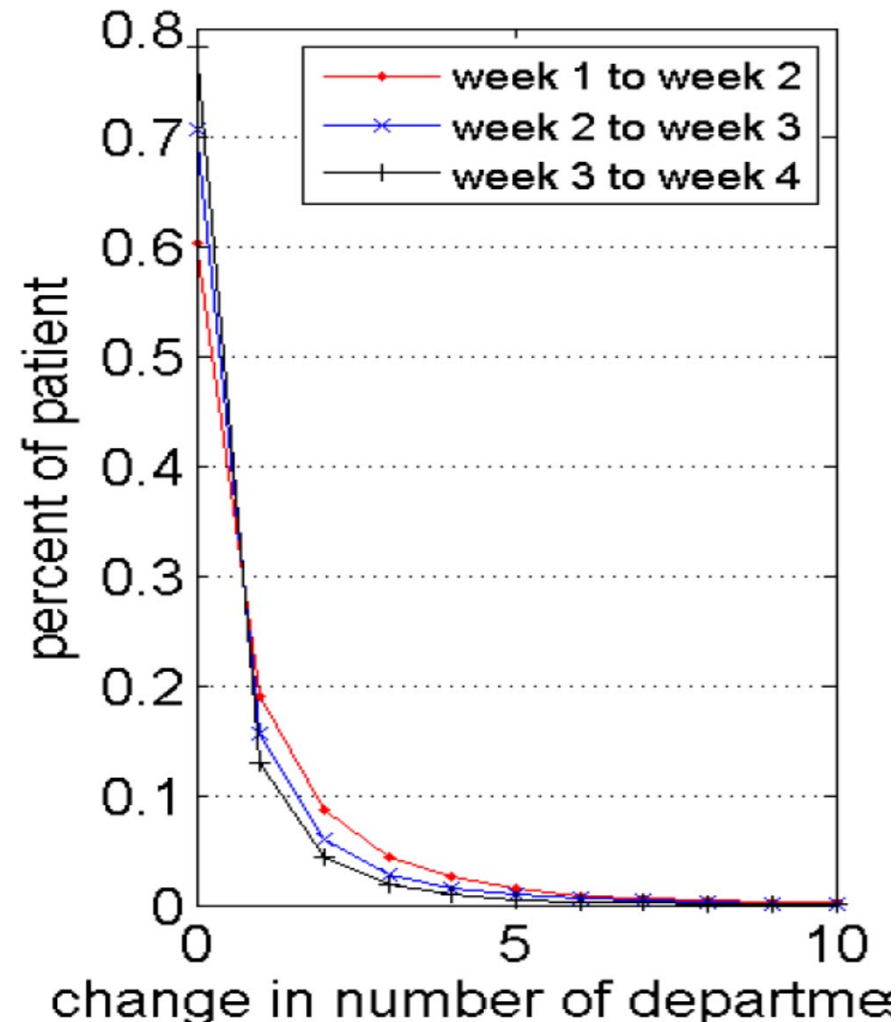


# Over 80% of local networks whose size is less than 5

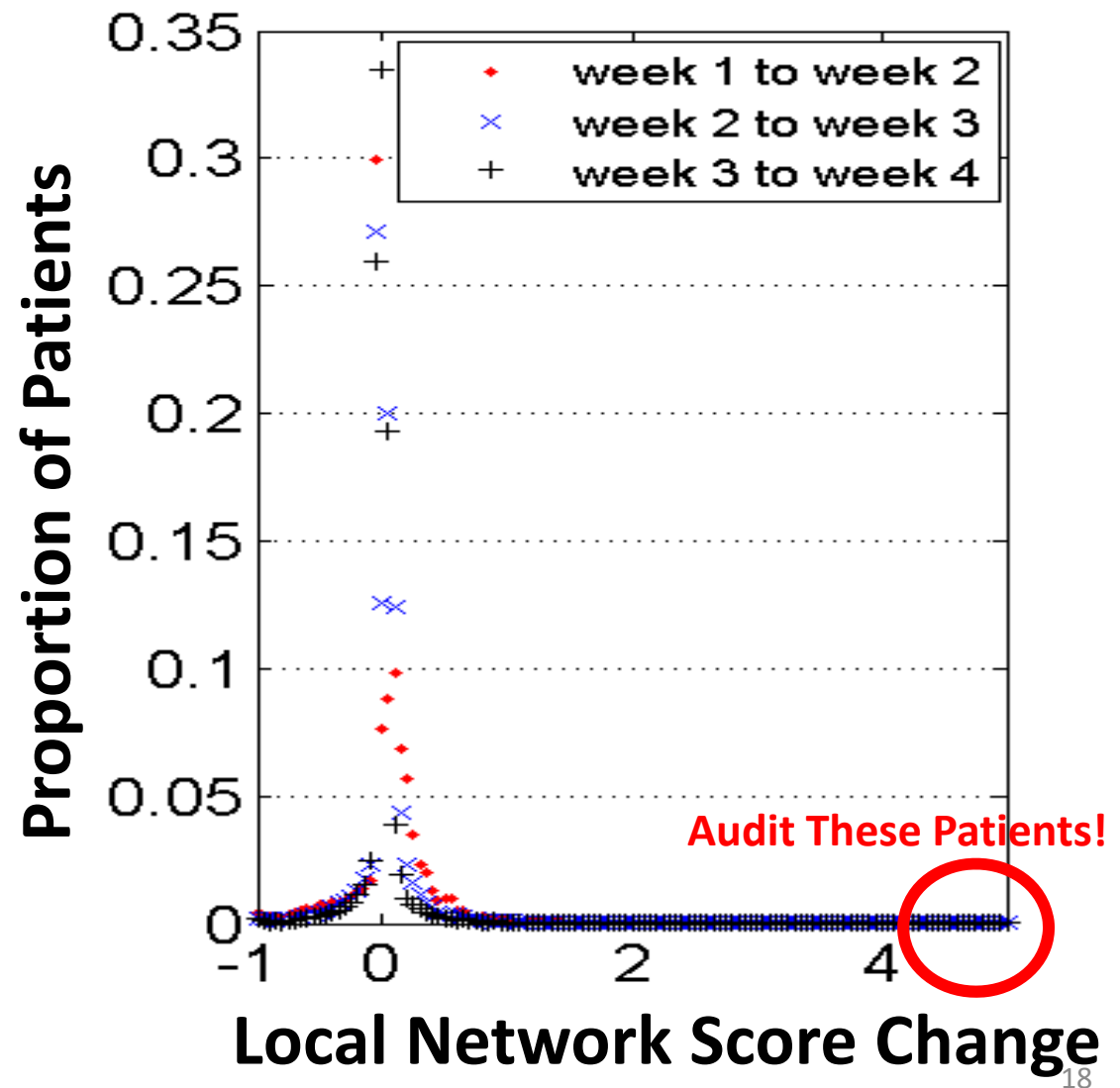




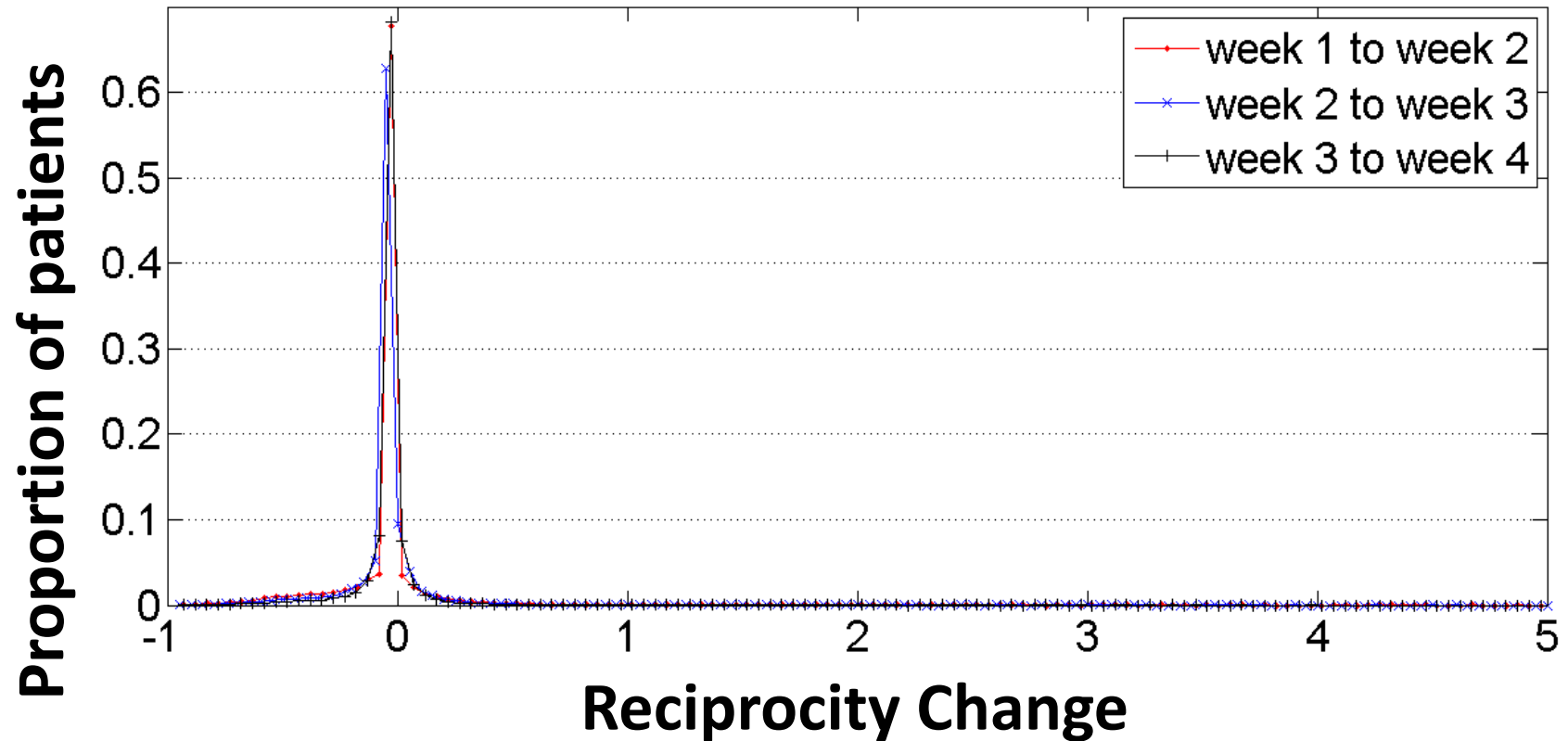
Over 80% of local networks has number of departments change less than 2



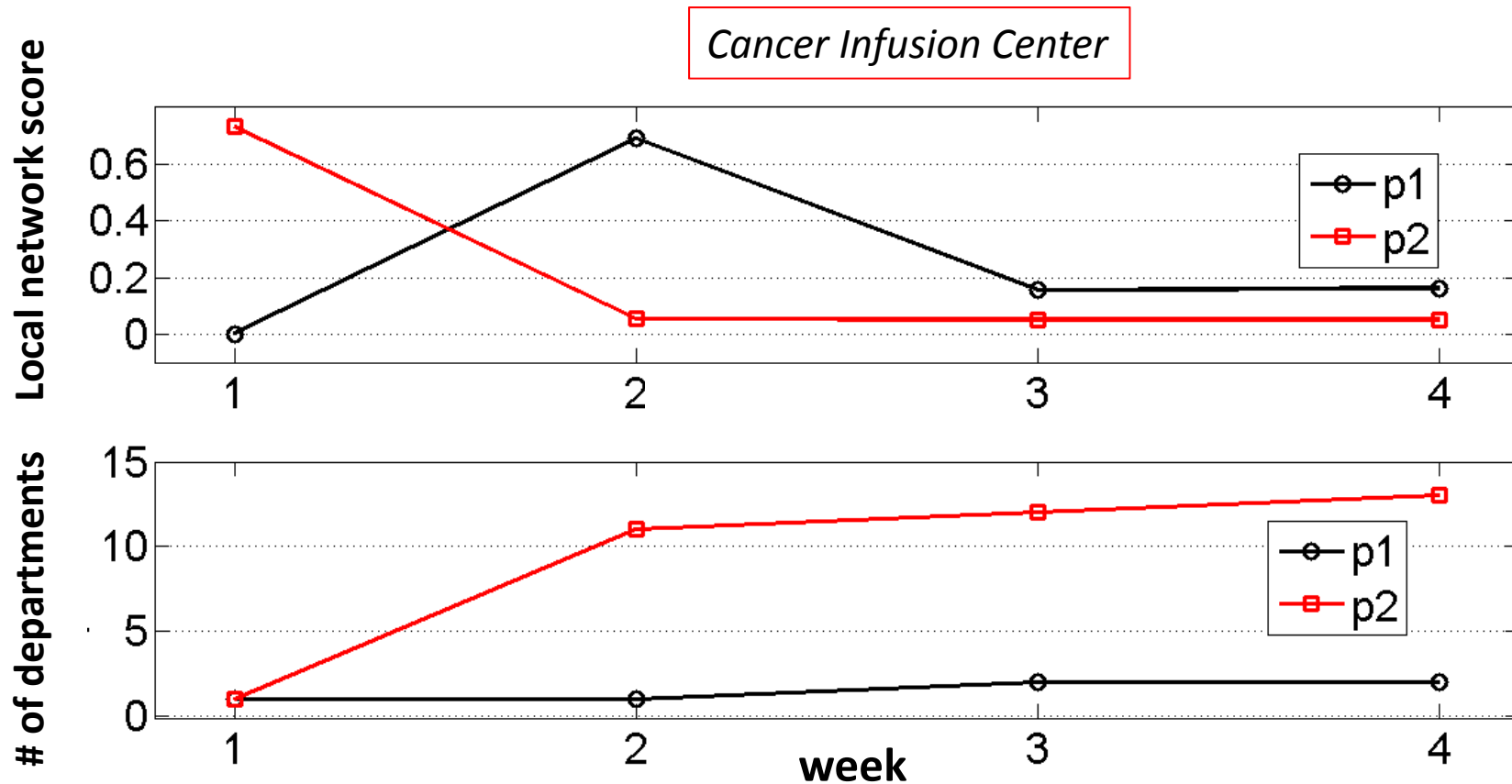
# Most Patients Network Suggest They Are “Normal”



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



**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*

# Findings

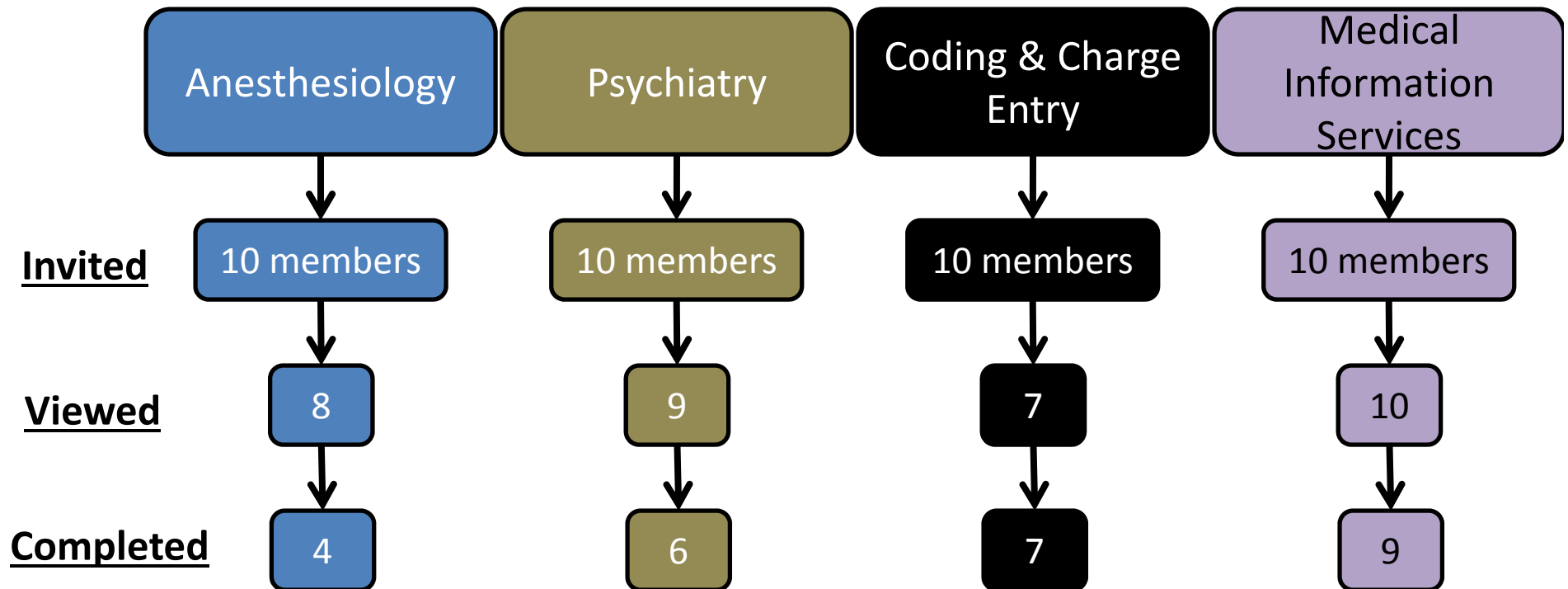
- We hypothesized an HCO would exhibit strong stability  
→ **confirmed by our experiments**
- The changes in the score of local networks do not justify the claim that the patient has been intruded upon, but may provide a reason for an investigation that incorporates more nuanced domain knowledge

But Do You  
Believe the Data?

# Survey Population

(Chen, Lorenzi, Nyemba, Schildcrout & Malin – IJMI 2014)

- Vanderbilt University Medical Center areas

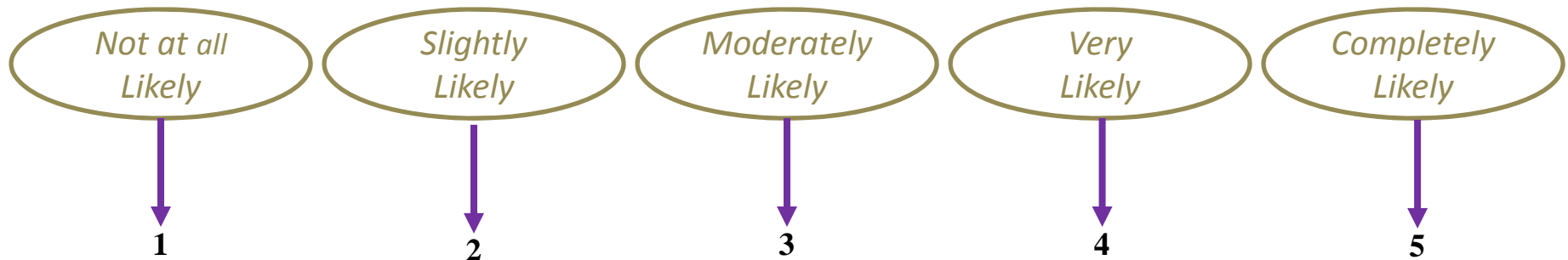


34 respondents did the survey and 26 of them are valuable

# Survey Questions

- Departmental interactions
- Conditional probabilities of accessing a record (conditioned on the HCO area)
- “Given someone from Coding & Charge Entry accessed the record, *what’s the chance someone from the following Area accessed the record?*”

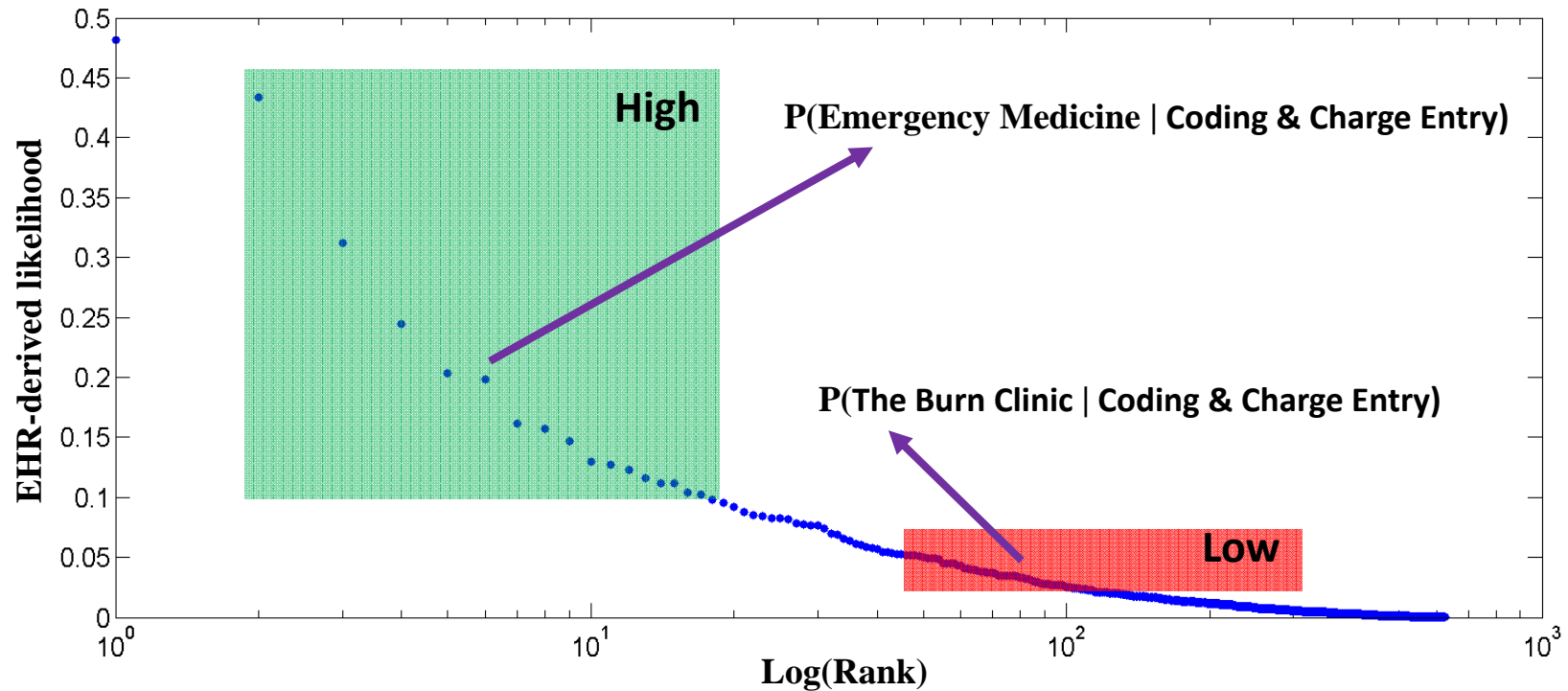
## Emergency Medicine



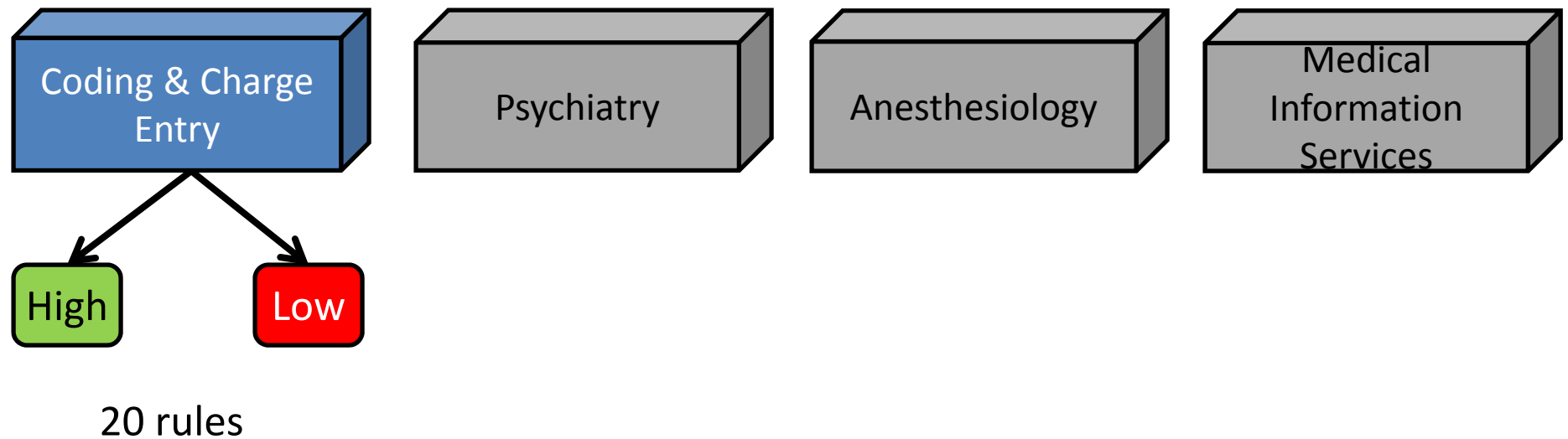


# Coding & Charge Entry Interactions

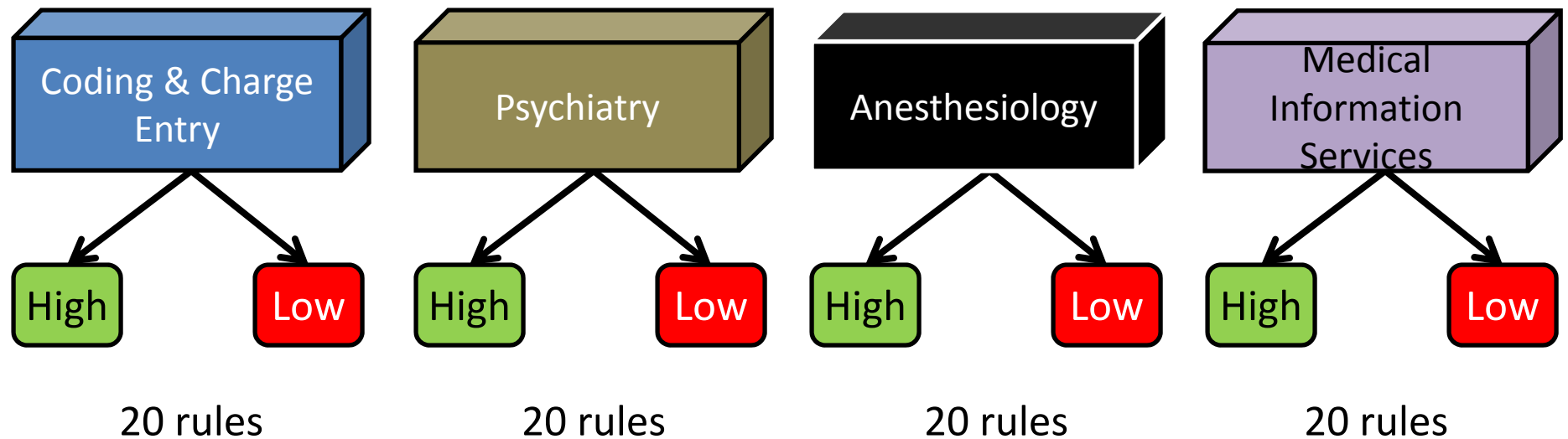
(one week, ~620 points)



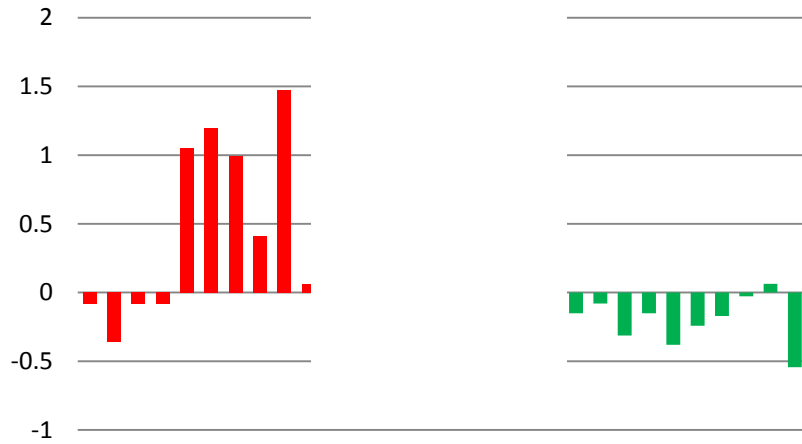
# Survey Questions



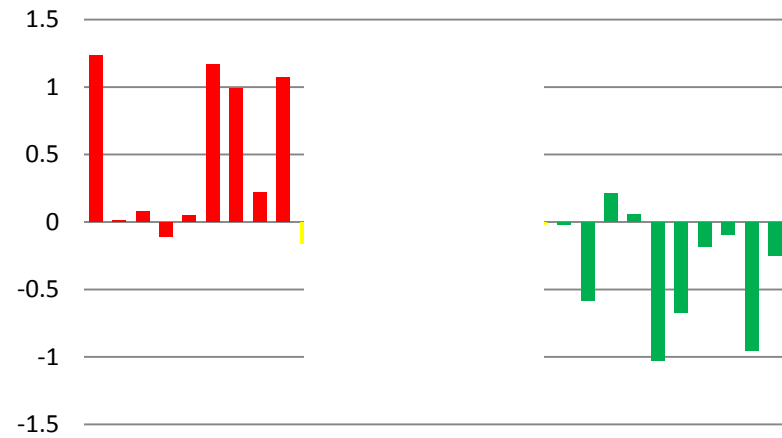
# Survey Questions



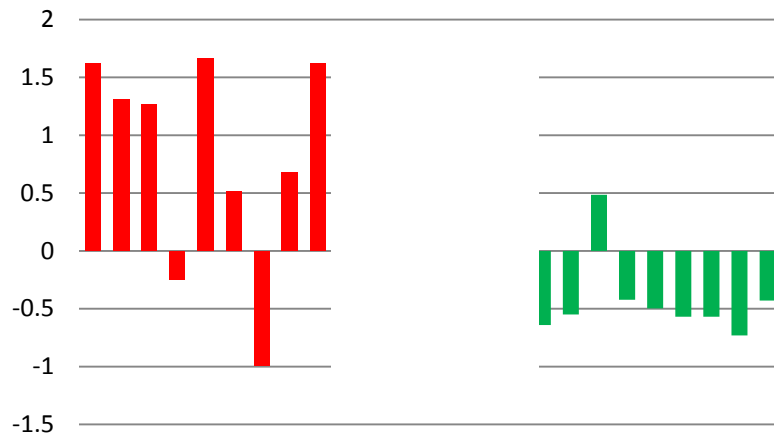
## MIS RESPONDENT(10)



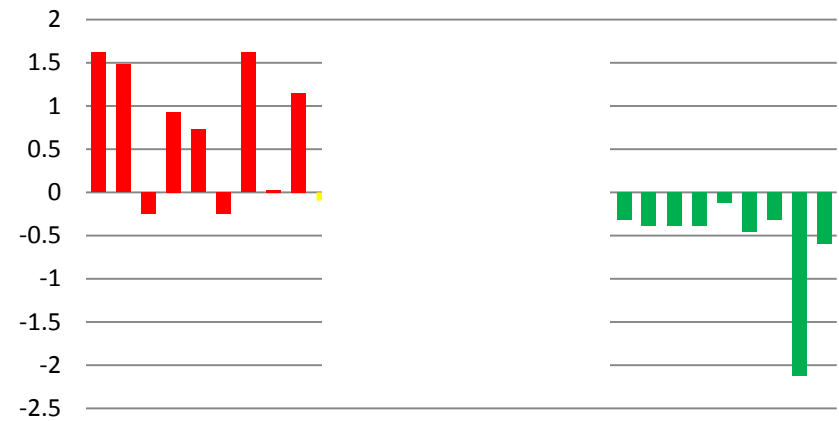
MIS Rules



CODE Rules

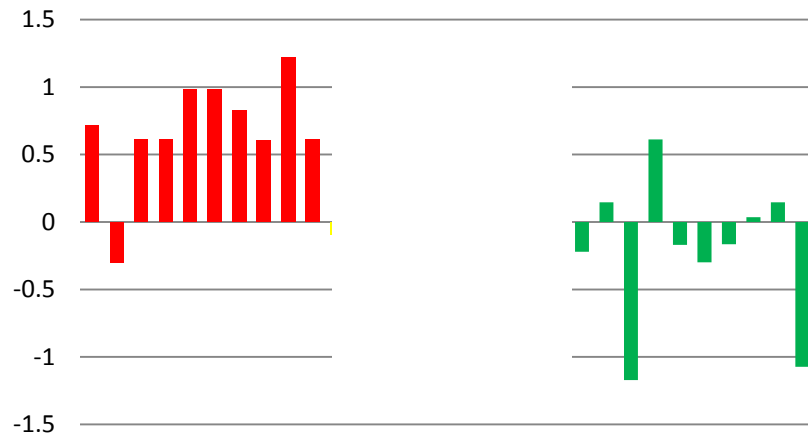


PSY Rules

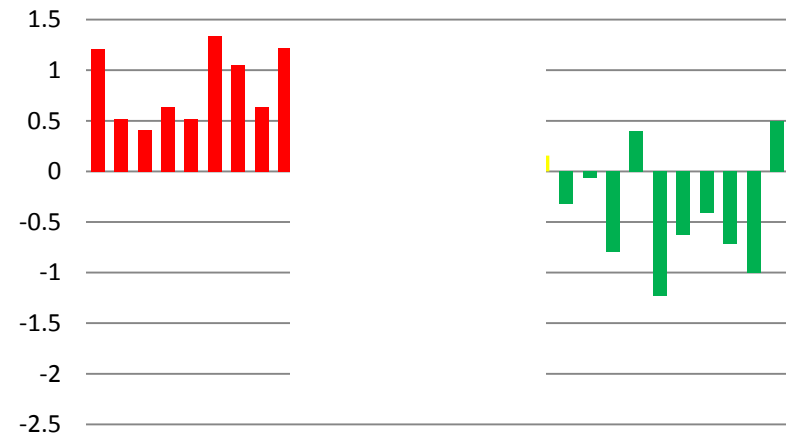


ANE Rules

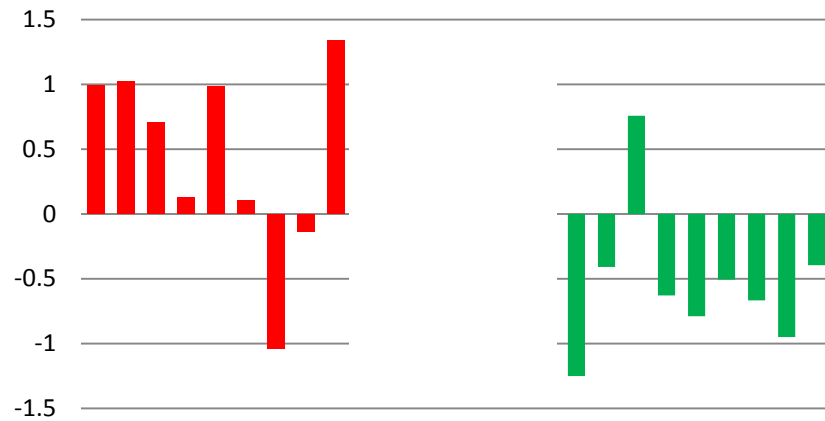
## CODE RESPONDENT(7)



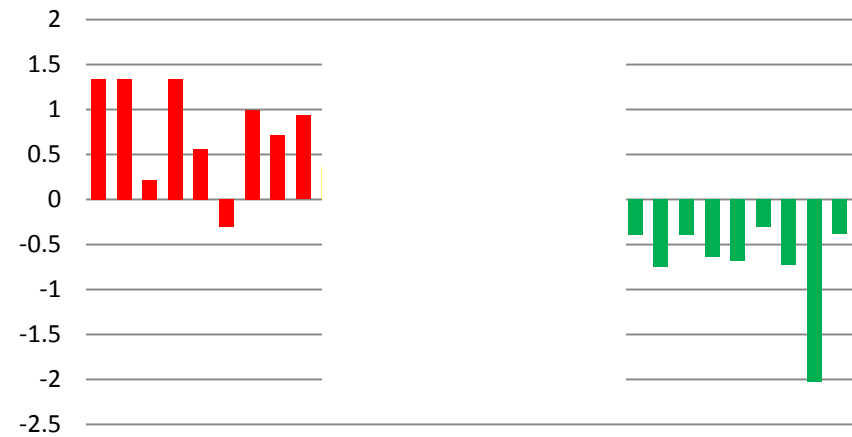
**MIS Rules**



**CODE Rules**

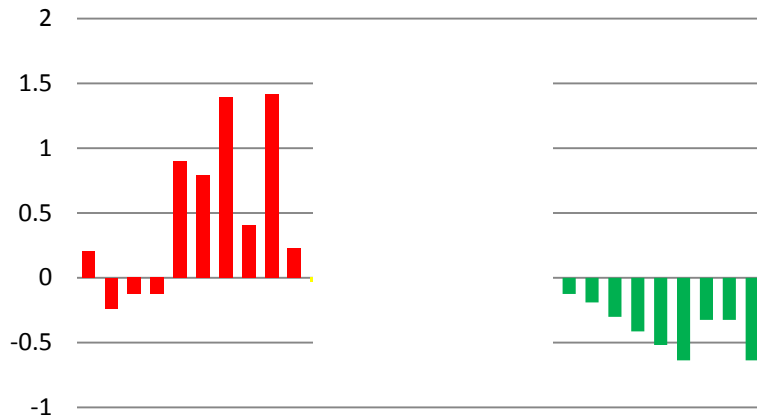


**PSY Rules**

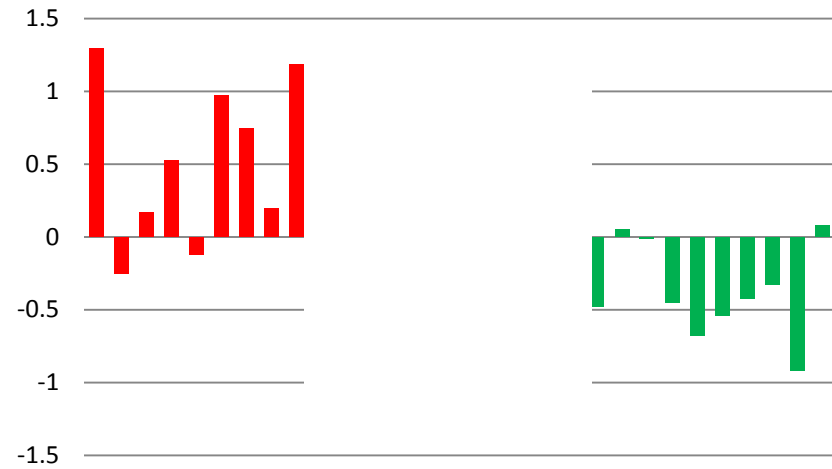


**ANE Rules**

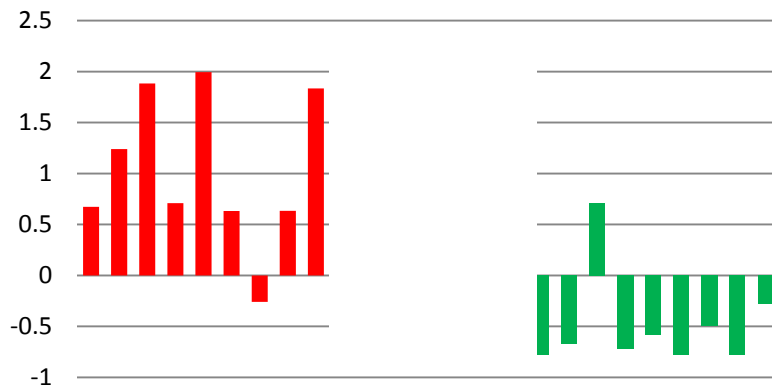
## PSY RESPONDENT(9)



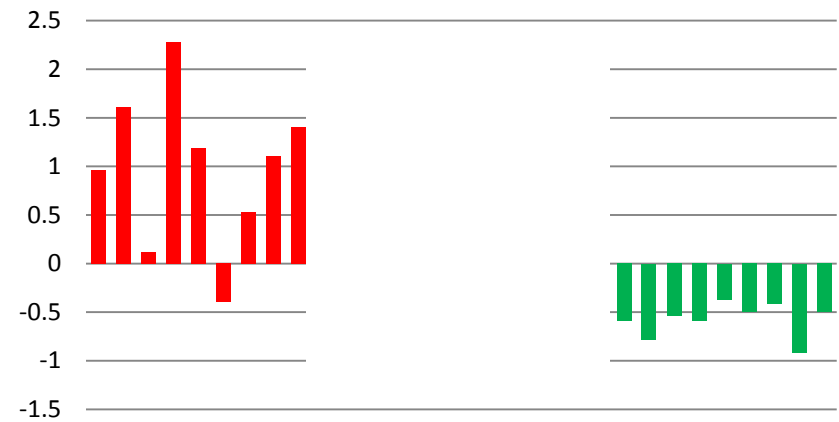
**MIS Rules**



**CODE Rules**

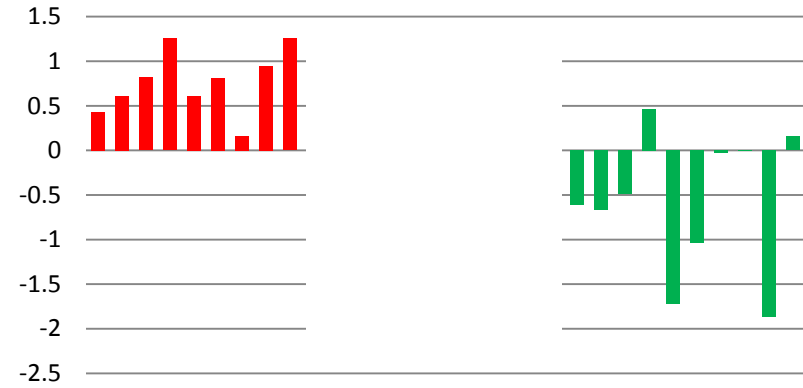
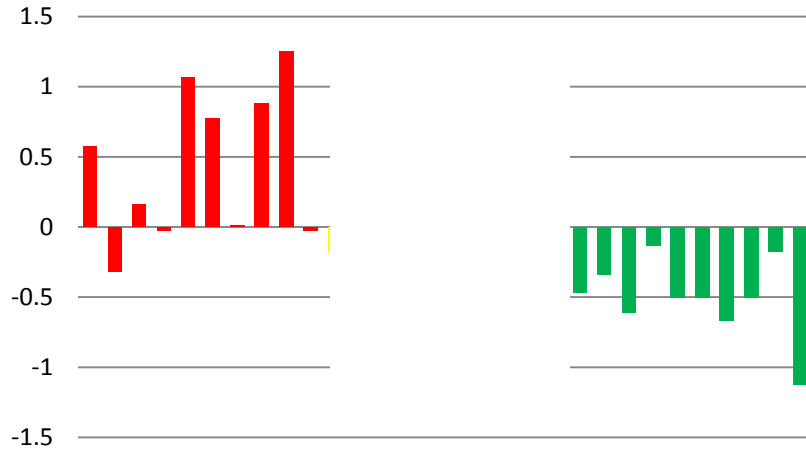


**PSY Rules**

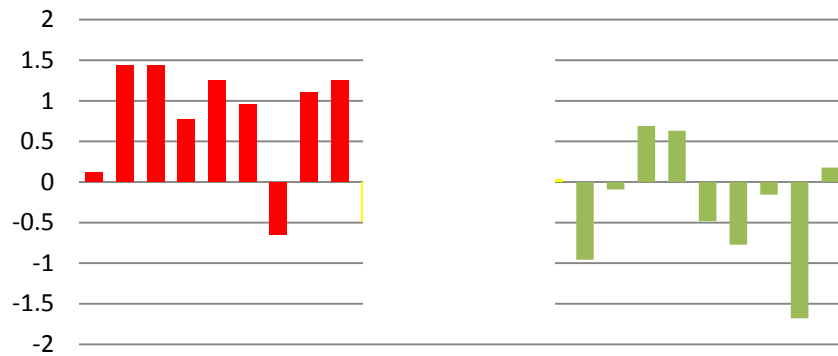


**ANE Rules**

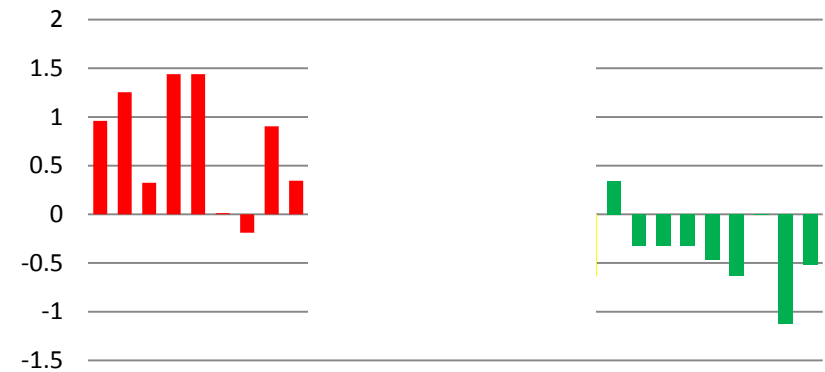
## ANE RESPONDENT(8)



## MIS Rules



## CODE Rules



## PSY Rules

## ANE Rules

# Hypothesis

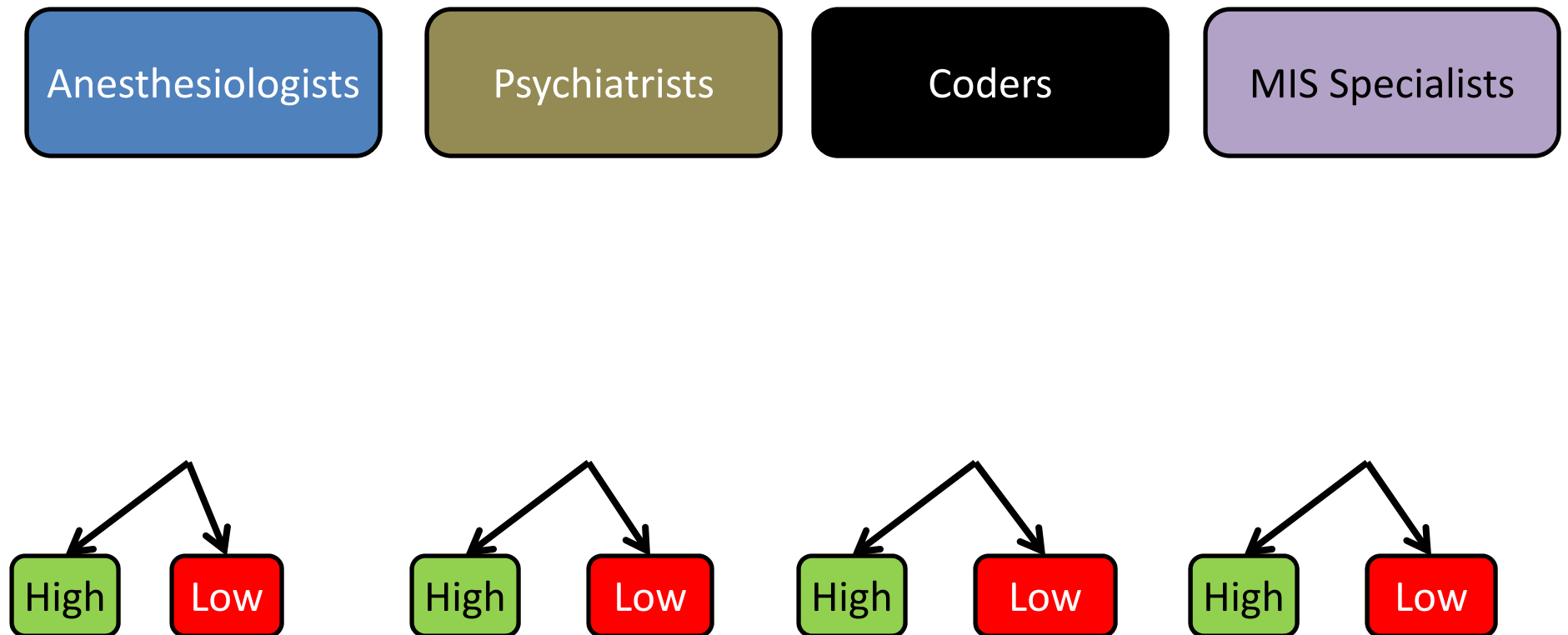
- 1) Employees can distinguish between high, and low likelihood rules for **all HCO areas**
- 2) Employees can distinguish between high and low likelihood rules for **their own HCO area**
- 3) employees can distinguish between high, and low likelihood rules in **their own** HCO area **better** than they can in **other** HCO areas



One respondent has 8 observations  
The total number of observations is  $8 \times 26 = 208$

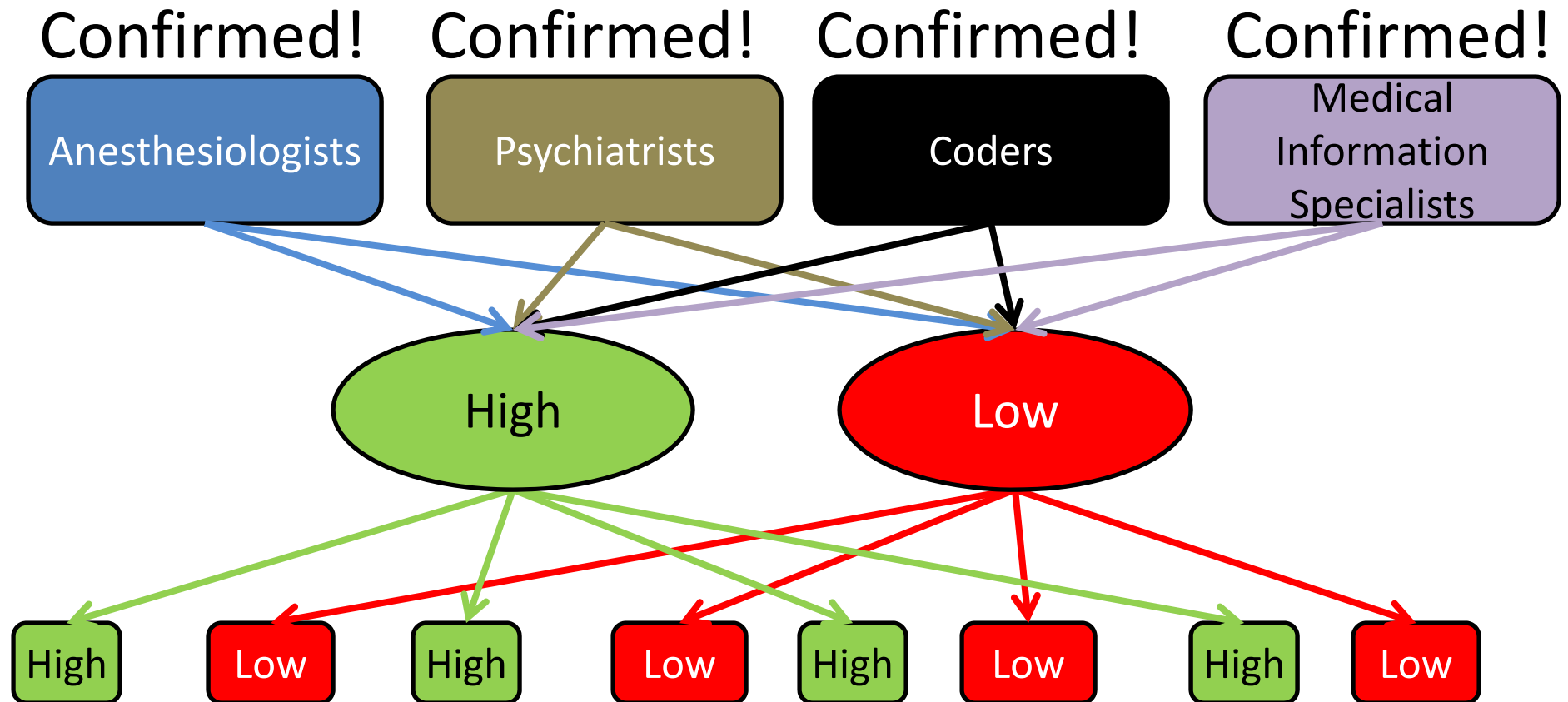
Respondent (ID)	Respondent Type (P)	Rule Type (R)	Rule Class (C)	Average Score of Responses
1	MIS	ANE	High	3
1	MIS	ANE	Low	2
1	MIS	CODE	High	3.3
1	MIS	CODE	Low	2.1
1	MIS	MIS	High	3.1111
1	MIS	MIS	Low	2.125
1	MIS	PSY	High	2.9
1	MIS	PSY	Low	2.05

# Hypothesis Test 1 – Rules of All HCO Areas:

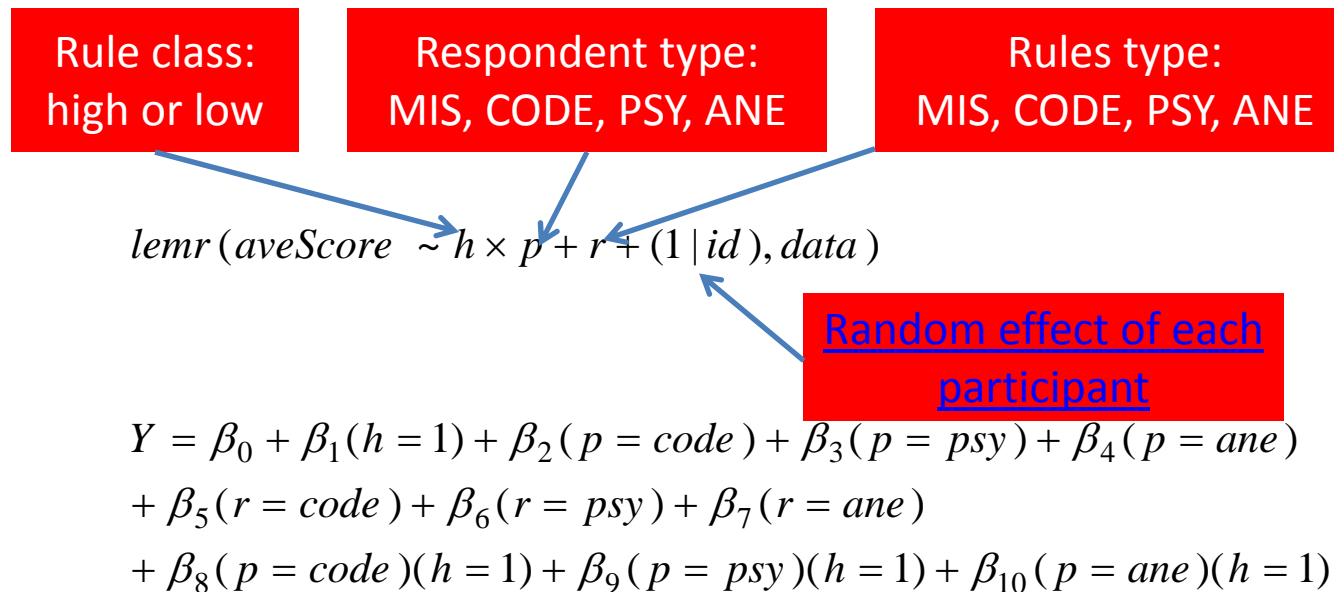


# Hypothesis Test1 – Rules of All HCO Areas:

One-sided t-test, 95% confidence



# Linear Mixed Effects Model



How **MIS respondents** distinguish high and low likelihood of rules for **all HCO areas**

$$E(y \mid p=mis, h=0, r=ALL) = \beta_0 + \beta_5 + \beta_6 + \beta_7;$$

$$E(y \mid p=mis, h=1, r=ALL) = \beta_0 + \beta_1 + \beta_5 + \beta_6 + \beta_7;$$

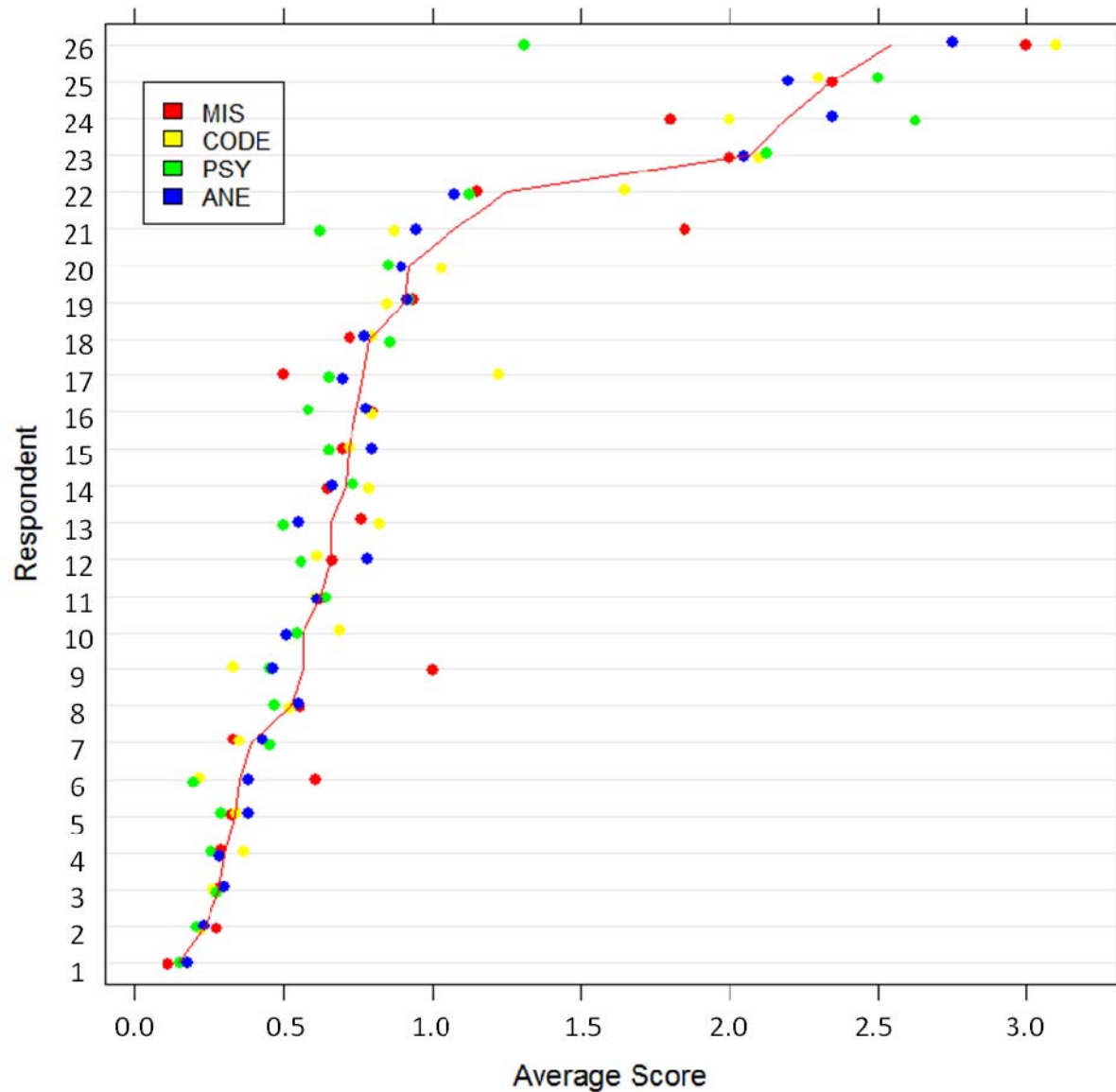
$$E(y \mid p=mis, h=1, r=ALL) - E(y \mid p=mis, h=0, r=ALL) = \beta_1$$

How about CODE, PSY and ANE?

## Distribution of respondents' average score for low rules

Certain respondents are inclined to assign large likelihoods (upper right section of the plot),

while others are included to assign small likelihoods (the lower left section of the plot).

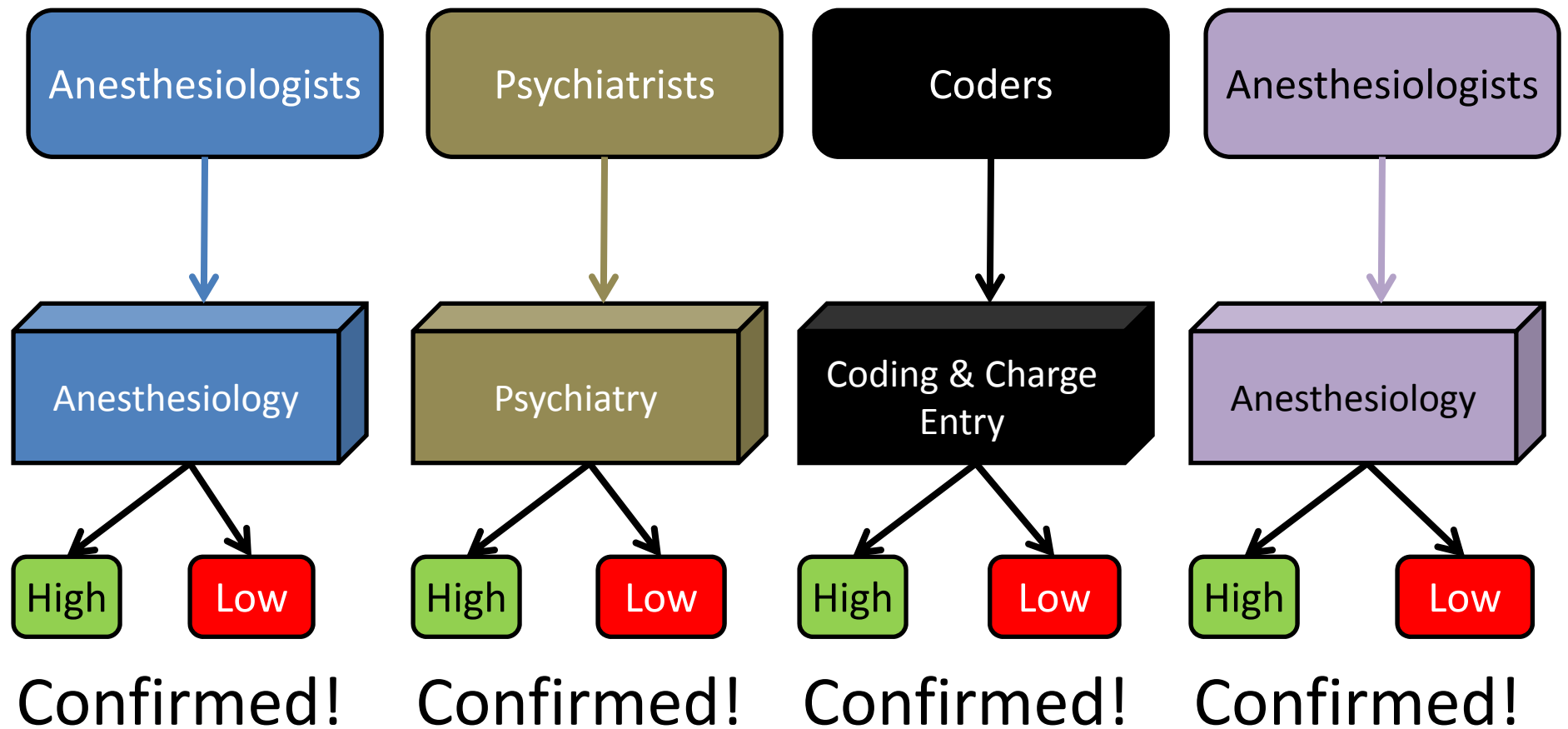


$\beta$	$\beta$ values	description	p value
$\beta_1$	0.351557	<b>MIS</b> respondents distinguish high and low likelihood rules for all HCO areas	$1.91 \cdot 10^{-9}$
$\beta_1 + \beta_8$	0.521492	<b>CODE</b> respondents distinguish high and low likelihood rules for all HCO areas	$1.11 \cdot 10^{-6}$
$\beta_1 + \beta_9$	0.677858	<b>PSY</b> respondents distinguish high and low likelihood rules for all HCO areas	$9.33 \cdot 10^{-8}$
$\beta_1 + \beta_{10}$	0.691166	<b>ANE</b> respondents distinguish high and low likelihood rules for all HCO areas	$1.22 \cdot 10^{-8}$

Respondents from four areas can distinguishing between high and low likelihood rules for **all HCO areas**

# Hypothesis Test 2– Self Assessment:

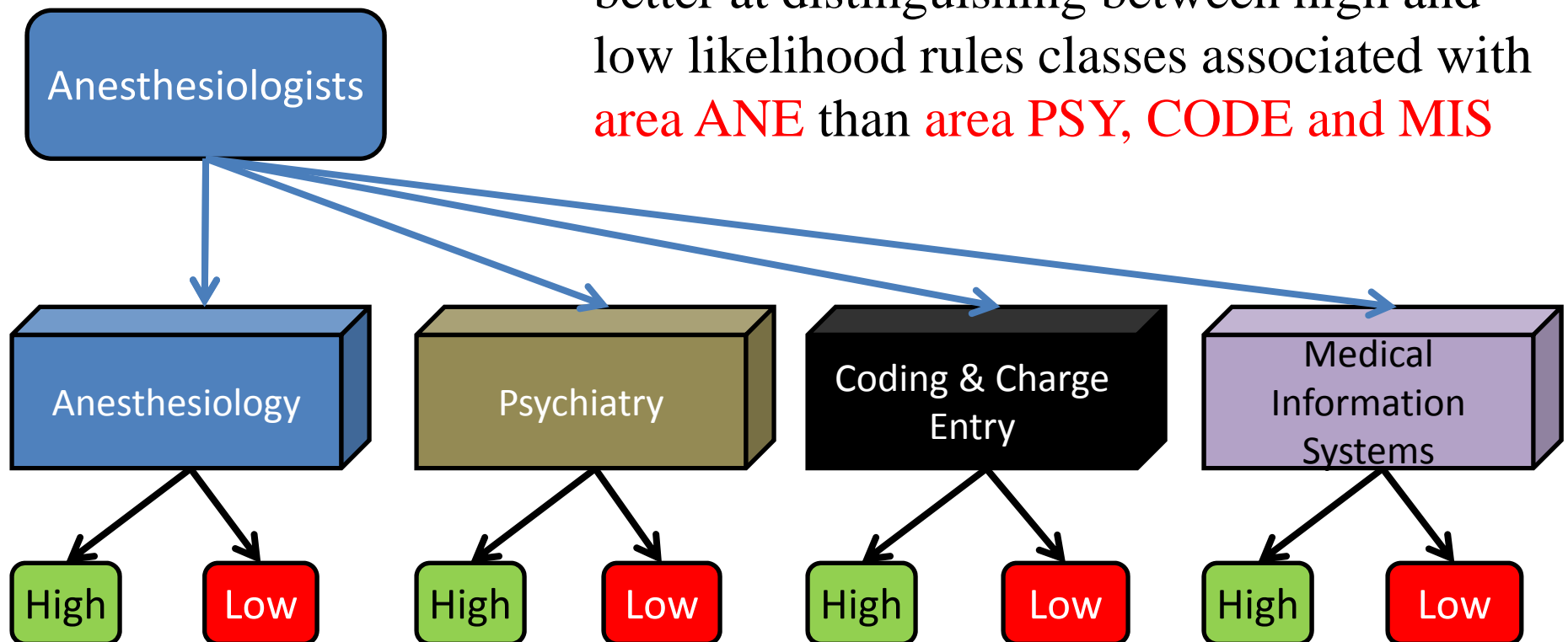
Linear Mixed Effects Model  
One-sided t-test, 95% confidence



# Hypothesis Test 3– Bias Toward Own Rules

Linear Mixed Effects Model  
One-sided t-test, 95% confidence

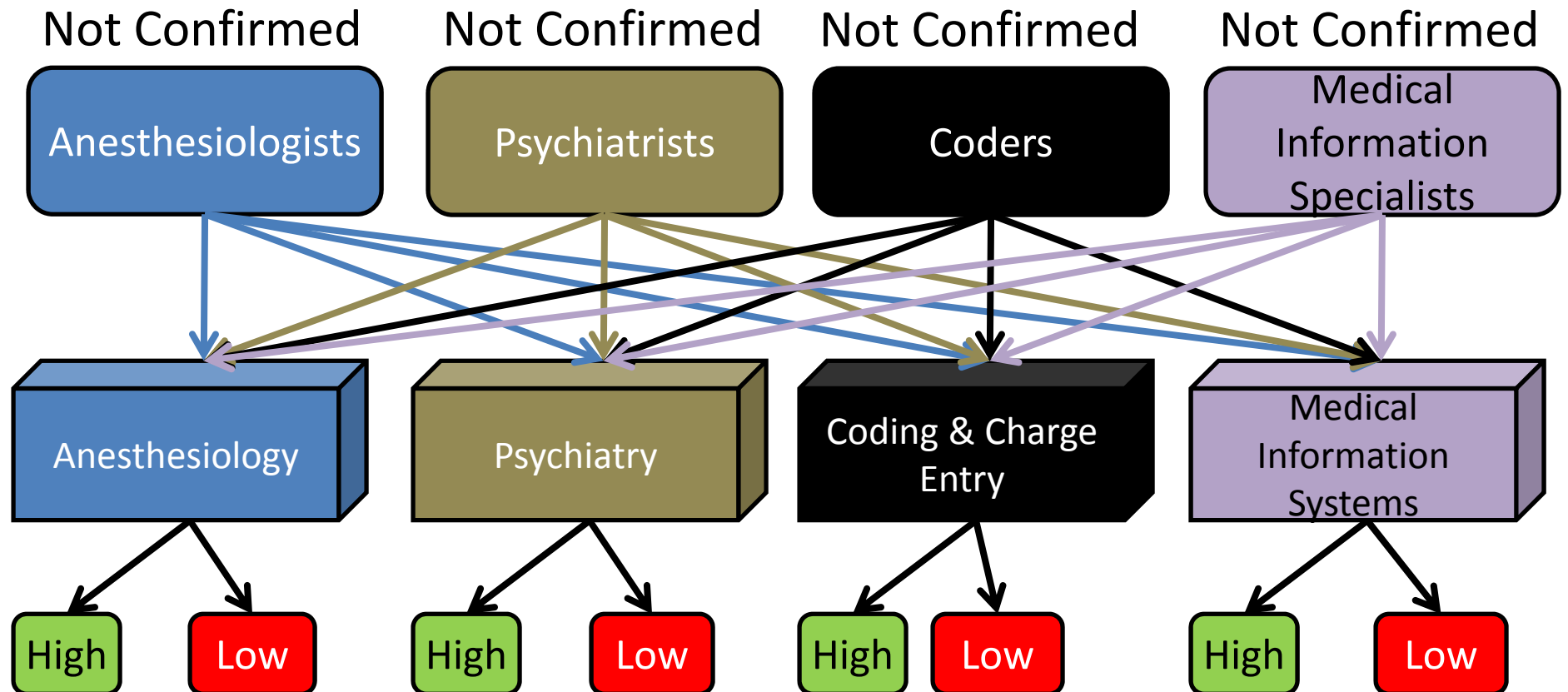
Respondents from **HCO area ANE** are better at distinguishing between high and low likelihood rules classes associated with **area ANE** than **area PSY, CODE and MIS**





# Hypothesis Test 3 – Bias Toward Own Rules

Linear Mixed Effects Model  
One-sided t-test, 95% confidence



# Conclusions

- Healthcare organization employees generally understand what goes on around them...  
... and for other sections of the organization as well!
- Automated healthcare organizational modeling may be possible.
- Anomalies detection through collaborative patterns may be reliable!

Q&A  
Thanks!