

A Probabilistic Look into the Semantics of Medicine

Peter Haug, MD



International Conference on
Semantic Computing 2014
June 16-19, Newport Beach, CA

Agenda

- Medical Models
- “Big” Healthcare Data
 - Capture as a part of the care process
 - Strong Temporal Component
 - Inherently messy
- Extracting Models from Data
 - A place for Ontologies
 - Semi-Automatic extraction of Diagnostic Models
- Bayesian Modeling
 - Variations on the theme

Agenda

- **Medical Models**
- “Big” Healthcare Data
 - Capture as a part of the care process
 - Strong Temporal Component
 - Inherently messy
- **Extracting Models from Data**
 - A place for Ontologies
 - Semi-Automatic extraction of Diagnostic Models
- **Bayesian Modeling**
 - Variations on the theme

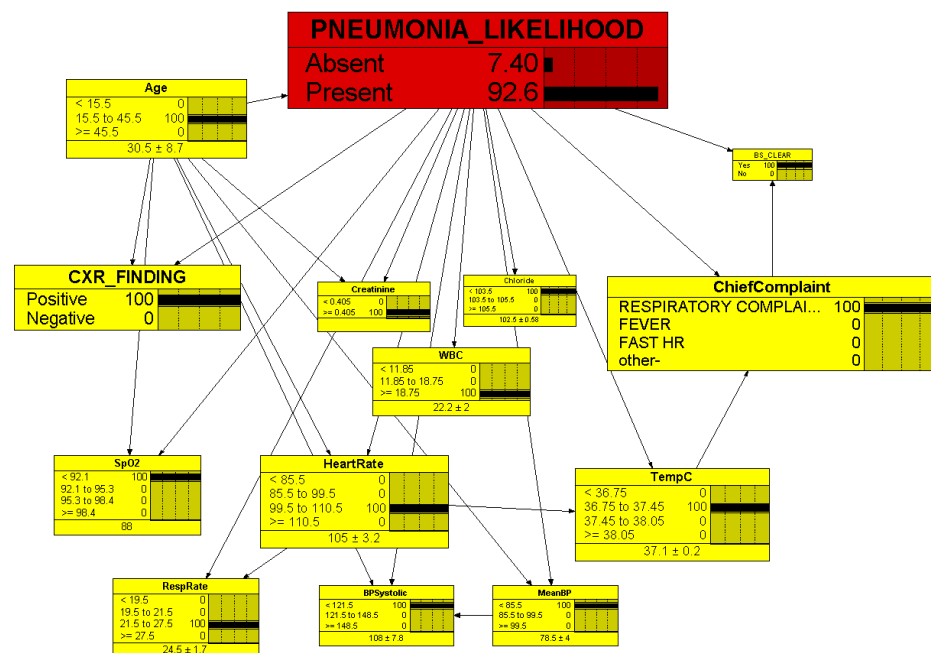
Models for Healthcare Delivery

A foundation for clinical research and care delivery

Modeling and Model Implementation

- Medical Data Models
- Diagnostic Models
- Therapeutic Models
- Workflow/Business Process Models
- Clinical Trials Models
- Natural Language Models
- Research Models
- Physiologic Models
- Ontologies
- Temporal Models
- Predictive Models

Translational Research
Data Visualization



Bayes Equation

Probability of Disease
When Finding exists

Probability of Both
The Disease and Finding


$$P(D | F) = \frac{P(F \text{ and } D)}{P(F)}$$

Probability of
Finding

Bayes Equation-A Graphical View

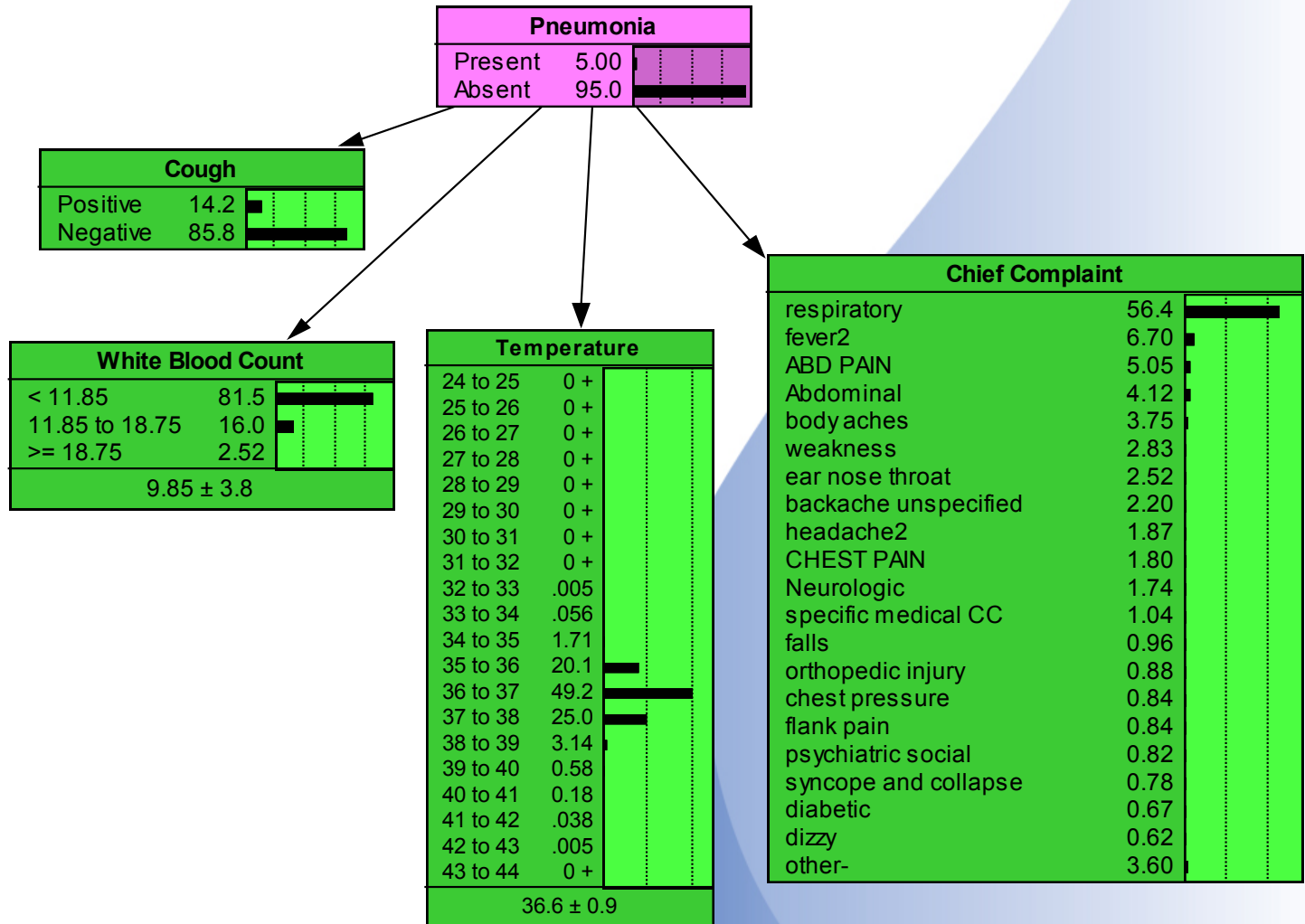
*Rendered as
a Bayesian
Network*

Disease		
Present	34.0	
Absent	66.0	



Finding		
Positive	38.9	
Negative	61.1	

Bayes Equation – Various Variables



*Rendered as
a Bayesian
Network*

Agenda

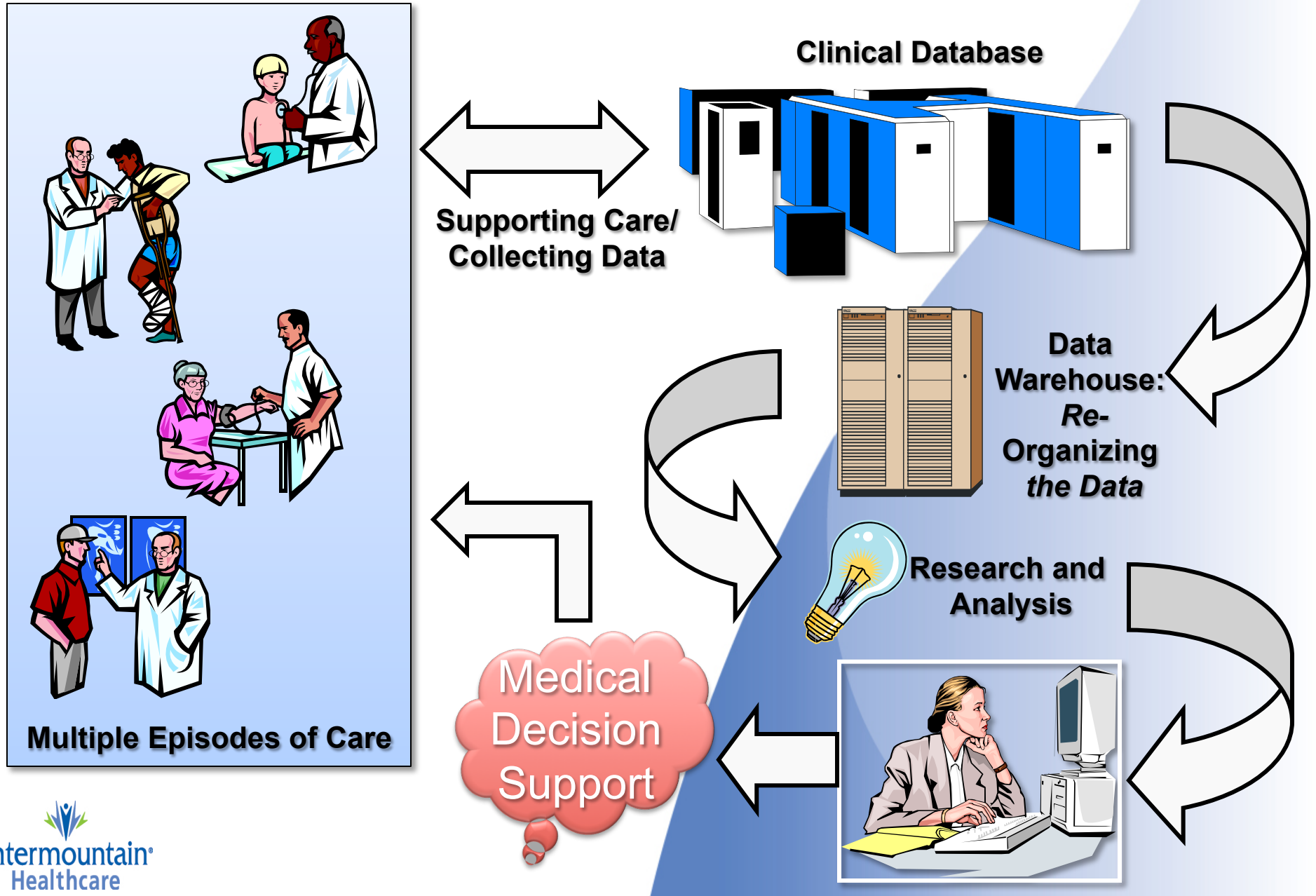
- Medical Models
- **“Big” Healthcare Data**
 - Capture as a part of the care process
 - Strong Temporal Component
 - Inherently messy
- Extracting Models from Data
 - A place for Ontologies
 - Semi-Automatic extraction of Diagnostic Models
- Bayesian Modeling
 - Variations on the theme

Sources for Building Medical Models

What are the sources of knowledge for Clinical Decision Support

- Clinical Trials
 - Prospective Trials
 - Observational Studies
- Physiologic Principals/Models
- Expert Opinion
- Common Sense
- CLINICAL DATA

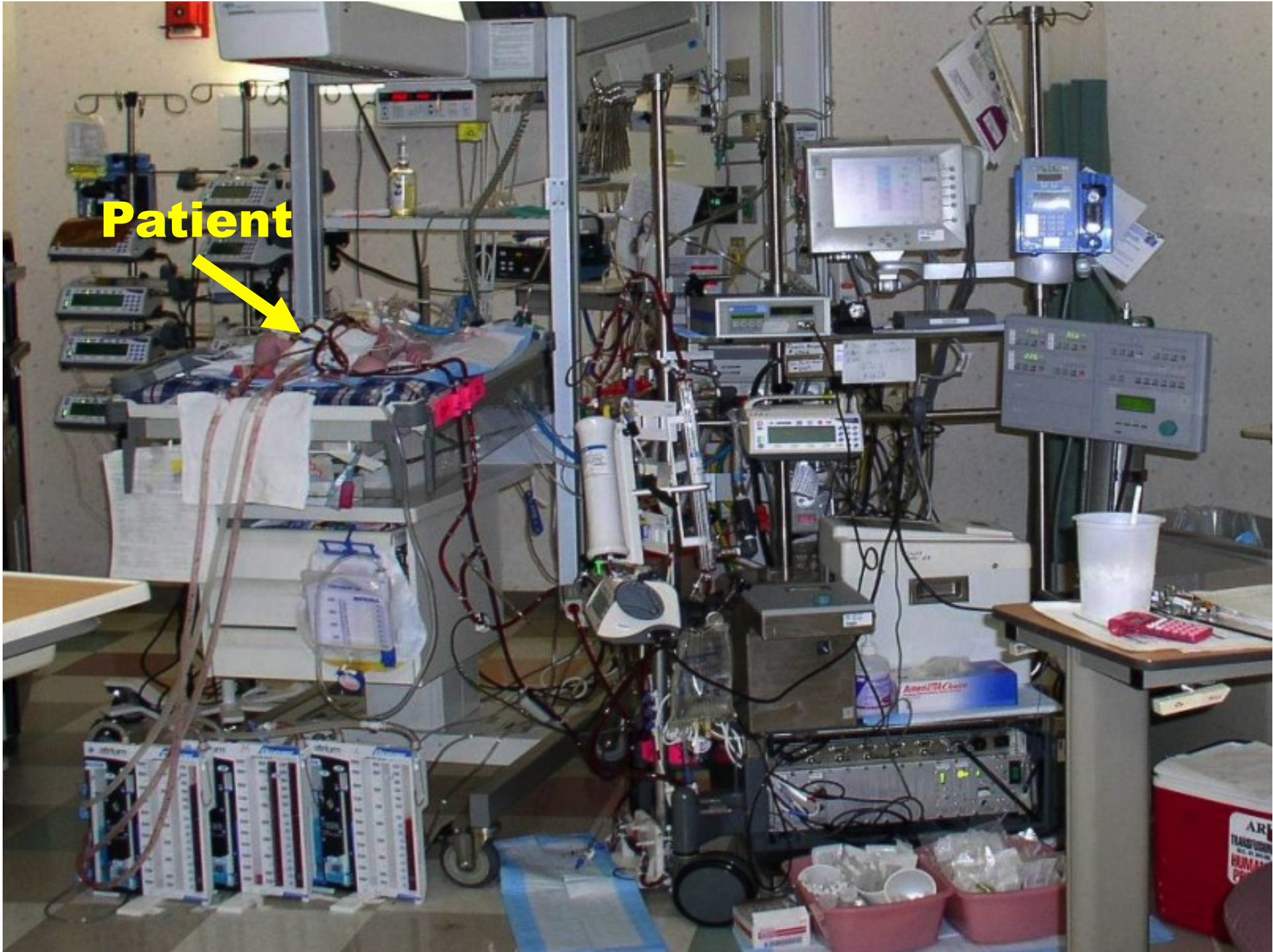
Clinical Data Serves Us Twice



EDW Interesting Facts...

<i>Fact</i>	<i>Measure</i>
# of EDW login accounts	2500
# of EDW “Consumers”	Thousands
# Records in EDW	70,000,000,000
Size (bytes)	~ 10 Terabytes (production)
Avg monthly queries	~ 150,000,000
Most queried table	LKUP.PATIENT_MASTER
Largest table (records)	HELP.PT_DATA (Patient Data Table from HELP1) 19.5 Billion Records
# queryable tables	12,500
Avg query run time	6 seconds
Avg rows returned per query	3500
# of years of data (for major data sets)	10 – 15 years (25+ for others)

Patient



Agenda

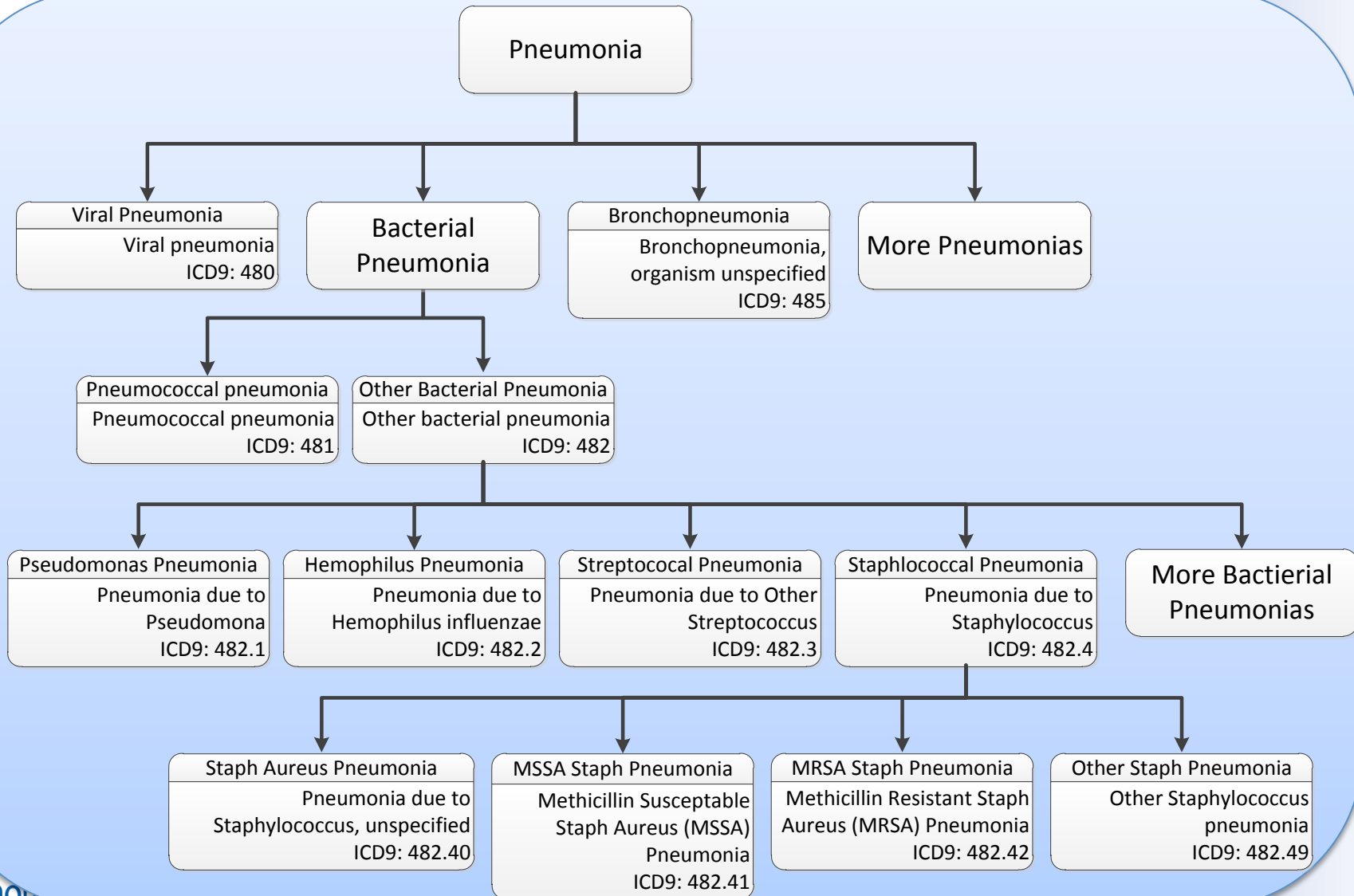
- Medical Models
- “Big” Healthcare Data
 - Capture as a part of the care process
 - Strong Temporal Component
 - Inherently messy
- **Extracting Models from Data**
 - **A place for Ontologies**
 - **Semi-Automatic extraction of Diagnostic Models**
- Bayesian Modeling
 - Variations on the theme

Using Ontologies to Extract Build Diagnostic Models

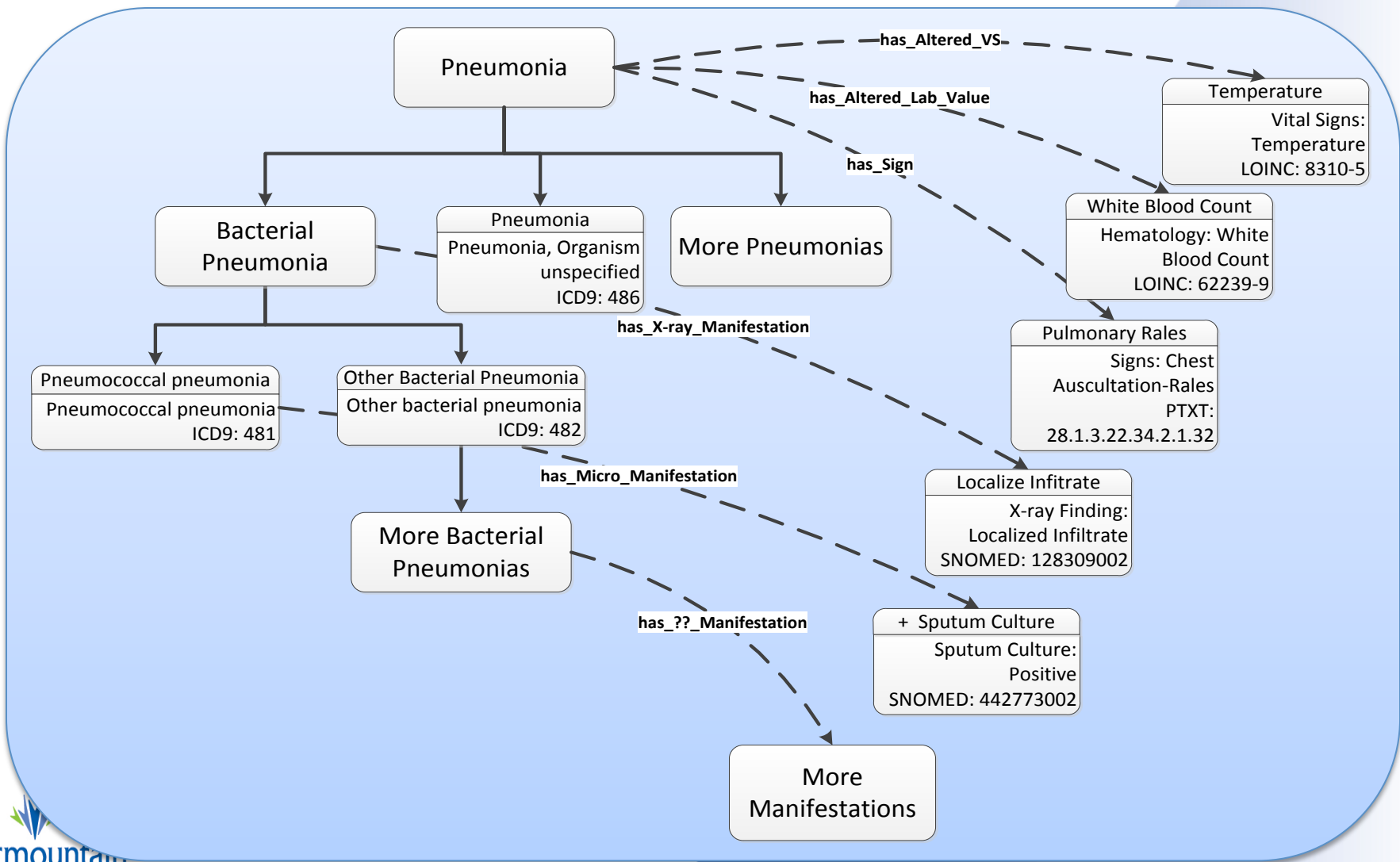
There is a need for predictive (diagnostic) models for care.

- Predictive models have value in clinical care environments
 - Have successfully deployed pneumonia and sepsis diagnostic models
- Building models from data is resource intensive.
- Ontologies can support diagnostic modeling
 - Requires a database containing data collected during routine care.
 - An **Ontology** to capture clinical relationships among data elements.
 - An application (the Ontology-driven Diagnostic Modeling System) is used to automate the initial analysis.

Ontologies Describe How Diseases Are Related (according to ICD9)

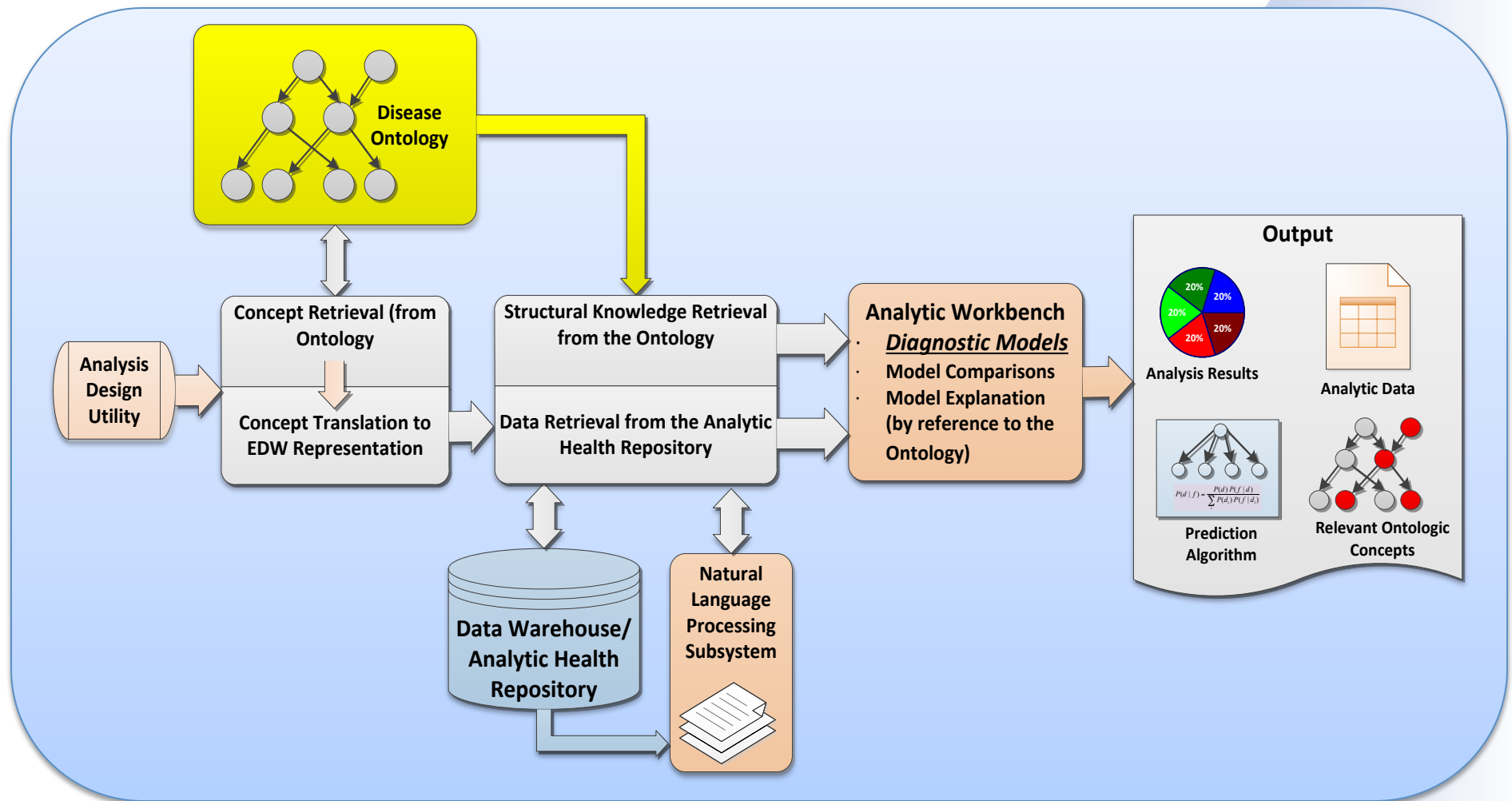


Ontologies Describe How Clinical Data are Related to Diseases

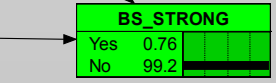
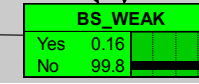
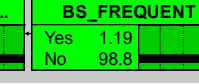
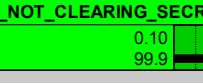
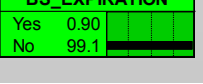
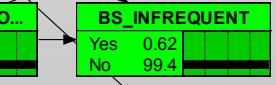
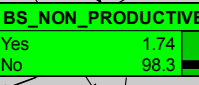
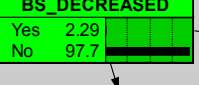
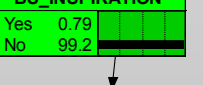
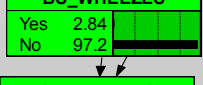
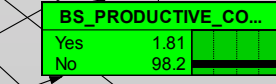
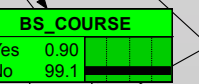
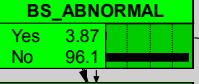
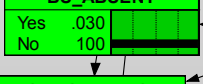
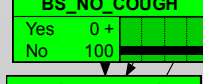
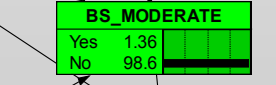
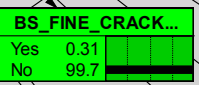
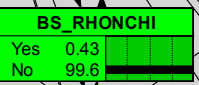
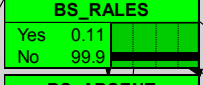
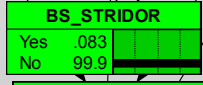
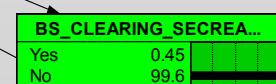
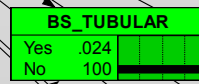
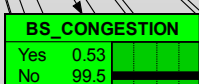
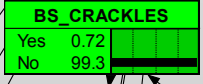
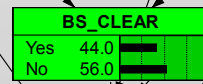
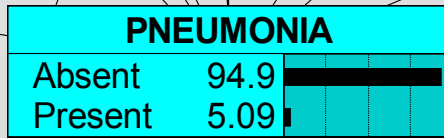
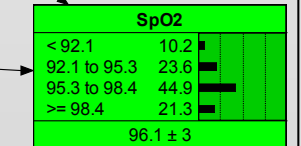
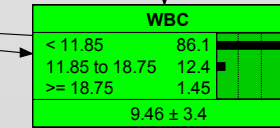
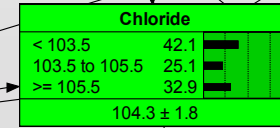
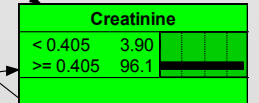
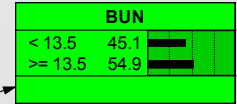
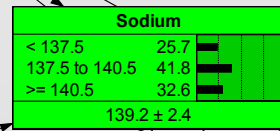
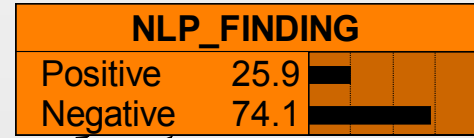
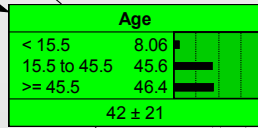
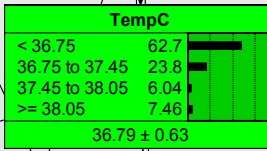
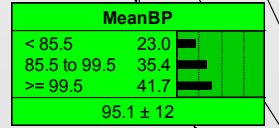
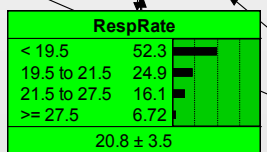
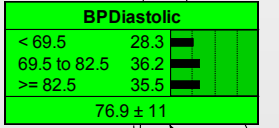
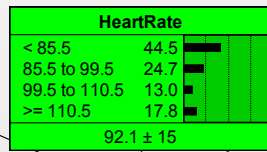
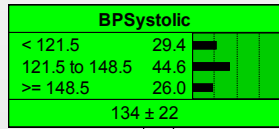


Ontology-Driven Model Discovery

Using knowledge embedded in ontologies to automate research?

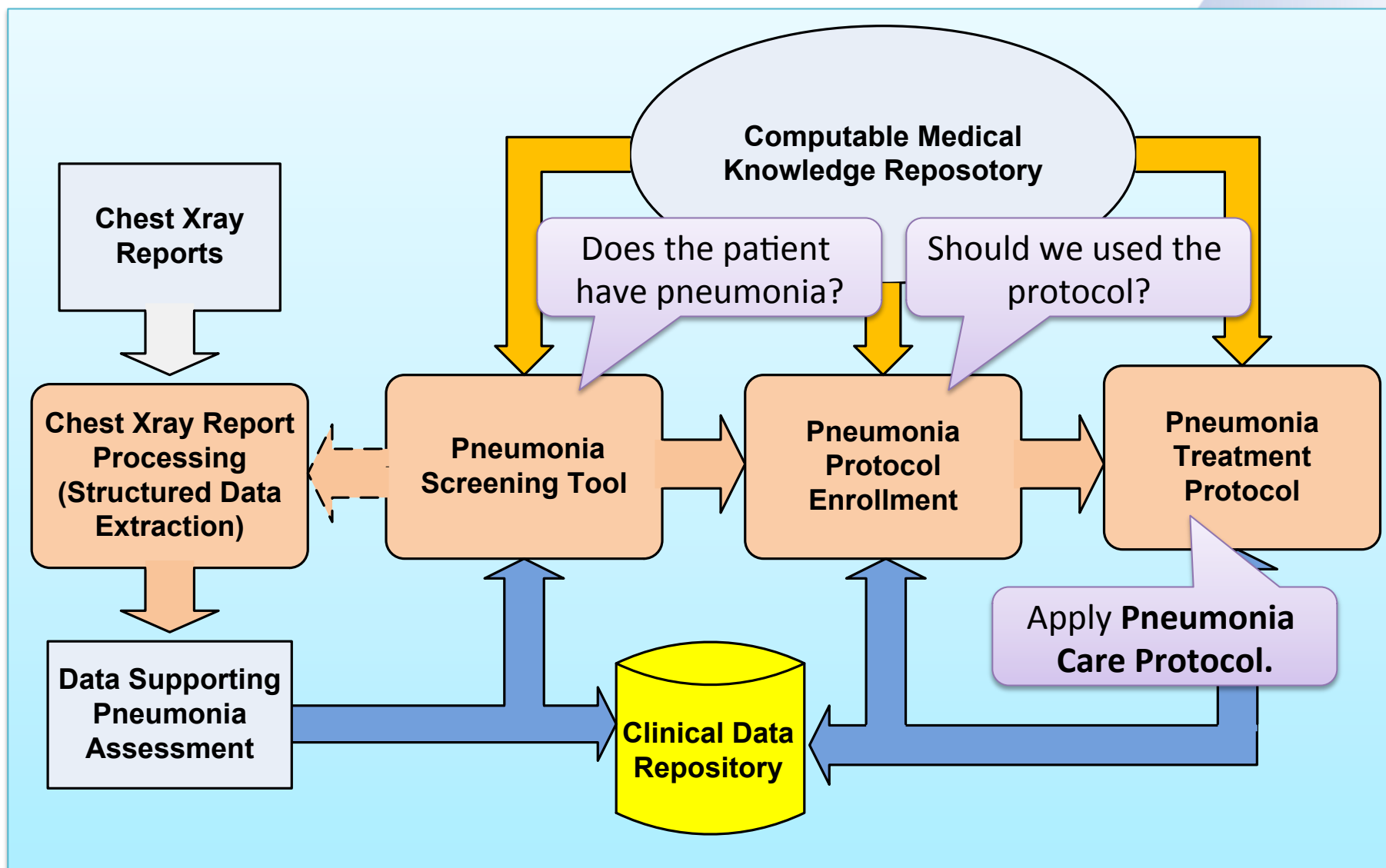


ChiefComplaint	
RESPIRATORY COMPLAINT	32.4
FEVER	6.96
ABD PAIN	6.05
ORTHO INJURY	4.26
CHEST PAIN	4.12
NEURO COMPLAINT	3.69
FALL	3.62
TRAFFIC INJURY	3.50
ABD PROBLEMS	3.45
CHEST PRESSURE	3.10
BACK PAIN	2.82
WEAKNESS	2.79
SYNCOPE	2.28
ENT PROBLEM	2.19
BODY ACHES	1.88
CV COMPLAINTS	1.88
HEADACHE	1.83
DIZZY	1.77
FLANK PAIN	1.43
CV PROBLEMS	0.92
ASSAULT RAPE	0.87
PSYCHIATRIC	0.86
CHEST HEAVINESS	0.82
SKIN COMPLAINT	0.78
SPECIFIC DIAGNOSIS	0.51
DIABETIC	0.44
PAIN CHEST	0.37
HEART RACE	0.33
TRAUMA	0.31
GENITOURINARY PROBLEM	0.31
PALPITATIONS	0.31
HEART IRR	0.30
ALLERGIES	0.29
HIGH BP	0.28
FLUID NUTRITION	0.27
CONVULSIONS	0.25
INFECTION	0.20
RAPID HR	0.19
IRR HEARTBEAT	0.16
LACERATION	0.16
INGESTION	0.16
BP HIGH	0.13
UNCONSCIOUSNESS	0.11
VAGINAL BLEEDING	.098
MED REFILL	.091
UNKNOWN	.087
LOW BP	.064
CARDIAC ARREST	.059
EYE PROBLEM	.055
BP LOW	.054
other-	0.18



A Care Delivery Framework (multi-factor screening and treatment)

Example: Community-Acquired Pneumonia



HELP clinical desktop

LDS Hospital

Refresh Remove Filters Zoom In Zoom Out Add Patient Merge Columns Areas Location Rooms Reports Forms Me

Beds Available: 43 Waiting Room Patients: 2 Avg Acuity: 2.89 Door to MD Seen: 00:16 Charge Nurse: Allen, Alisha
 Beds Occupied: 9 Waiting for Triage: 2 Avg LOS: 01:27 Door to RN Seen: 0 Triage Nurse: Allen, Alisha
 Patient Count: 11 Waiting for IP Bed: 0 Avg Lab TAT: 00:17 # Consults: 0 Second Triage Nurse:
 Pts Since Midnight: 26 Avg X-Ray TAT: 00:05 Greeter:
 View Clinician/Patient Ratio

Rm	I	Sex	Age	Complaint	LOS	Reg	Status	MD	Res	RN	ORD	L	U	R	E	RT	PROT	CN	Cons	Comments
+ 01		F	93	SOB/CARDIO ...	00:59	<input checked="" type="checkbox"/>	PreFloor	Shane		Kim	N	R			R	TC		<input checked="" type="checkbox"/>		PF50/70
+ 02		F	6	SZ ...	03:19	<input checked="" type="checkbox"/>	Seen	KMG		Kurt	C	R	R					<input checked="" type="checkbox"/>		
+ 03		M	42	CHEST PX	02:24	<input checked="" type="checkbox"/>	Seen	Shane		Randy	C	R		R	R			<input checked="" type="checkbox"/>		repeat cardiac enzym...
+ 06		M	72	CARDIO ...	01:05	<input type="checkbox"/>	Seen	KMG		jkehl	C	R		R	R			<input checked="" type="checkbox"/>		
+ 07		F	36	DETOX ...	00:12			KMG		Kim								<input checked="" type="checkbox"/>		
+ 09		M	66	POST OP PAIN ...	02:24	<input checked="" type="checkbox"/>	Seen	KMG		Randy	C	R						<input checked="" type="checkbox"/>		i-stat cr 1.0/Joy to...
+ 10		F	33	OBGYN ...	03:38	<input checked="" type="checkbox"/>	Seen	Shane		Kim	C	R		*				<input checked="" type="checkbox"/>		
+ 14		M	37	RESP ...	00:54	<input type="checkbox"/>	Seen	Shane		jkehl	C	P	O				P	<input checked="" type="checkbox"/>		
+ 22		M	27	HA/FEELS SICKS ...	01:03	<input checked="" type="checkbox"/>	Cons	KMG		jkehl	C	D	C					<input checked="" type="checkbox"/>		crisis 2C
+		F	24	RT ANKLE INJ ...	00:03					Kurt										
+		M	73	LT HAND NUMBNES...	00:05			KMG		Kurt										

Alerting for Pneumonia in the Patient Tracking System

- System Watches the Data Flow in the ED
- Identifies Possible Pneumonia Patients

Room: H2... DOB: Age: 78 Sex: M EMPI:

Screening CAP Protocol

Review Restart

Hospital Ward Admission recommended. The patient has Moderate CAP based on no HCAP factors, a 30 day mortality risk of 8.2%, and a 33% likelihood of needing ICU treatment, based on 2 Severe CAP criteria.

Agree with Recommendation

Disagree with Recommendation

Click the icons to add information and click Save Changes to update.



Health Care Acquired Pneumonia Factors

A single positive HCAP factor means the patient has HCAP.

- Hospitalization >= 2 days within 90 days: Yes No
- Nursing Home Resident: Yes No
- Wound Care or Infusion Therapy within 30 days: Yes No
- Chronic Dialysis within 30 days: Yes No



Vital Factors

- Age: 78 Years
- Confusion (patient not oriented to person, place, or time): Yes No
- Temperature: 37.5 C
- Respiratory Rate: 26 BPM
- Systolic Blood Pressure: 102 mm/Hg



Labs

- BUN: 21 mg/dL
- WBC: 29300 cells/mm3
- Platelet Count: 268000 cells/mm3
- PaO2/FiO2 Ratio: 252.3 mm/Hg



Radiology

- Pleural Effusion: Yes No
- Infiltrates: Single Lobe Multilobar

P/F Ratio = 52.974 / 0.210 = 252.255
 SpO2 --> PaO2 - conversion by Ellis equation = 52.974
 Calculated FiO2 = 0.210 + (0.000 * 0.03) = 0.210
 SpO2 on room air = 87.000
 FiO2 at room air = 0.21

Save Changes

Agenda

- Medical Models
- “Big” Healthcare Data
 - Capture as a part of the care process
 - Strong Temporal Component
 - Inherently messy
- Extracting Models from Data
 - The classic research paradigm
 - Semi-Automatic extraction of Diagnostic Models
- **Bayesian Modeling**
 - **Variations on the theme**

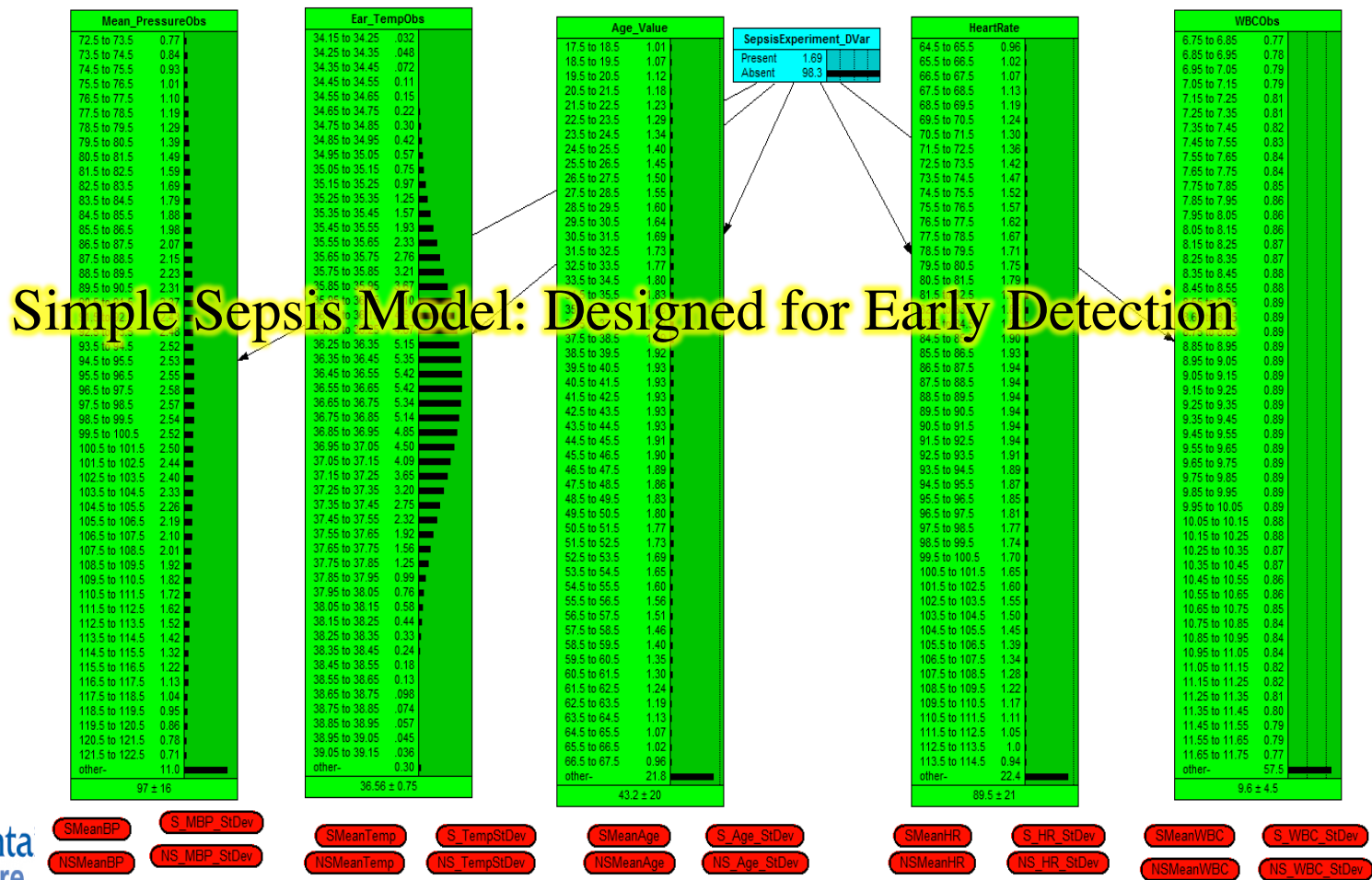
Intermountain Sepsis Bundle

To be successful, 11 components of the Sepsis Bundle should be accomplished.

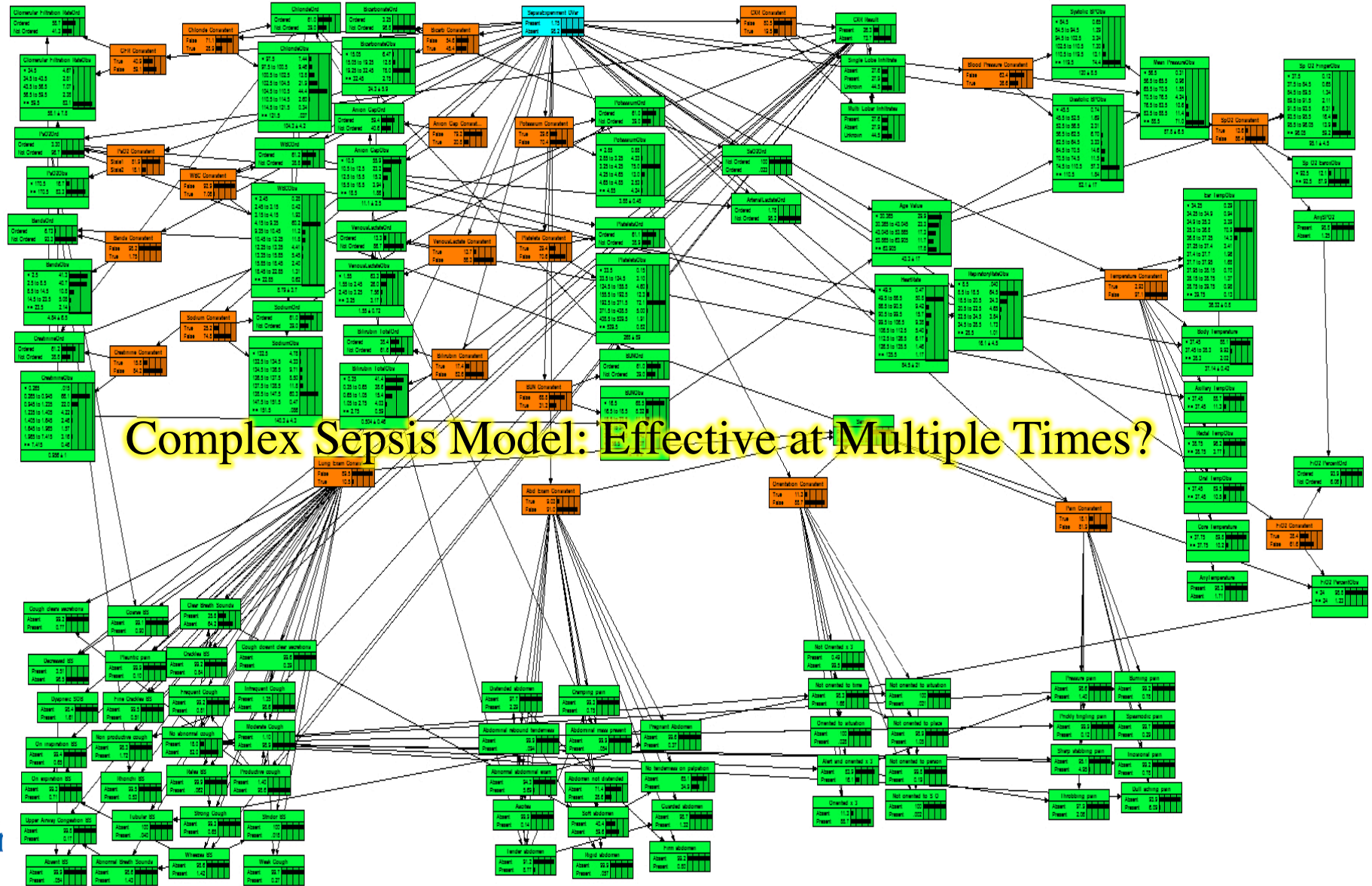
IMCP Adult Severe Sepsis/Septic Shock Bundle Handoff/Checklist - 2014		
Patient Name: _____ ENC: _____ Date of Admission: ____/____/____		
:	← "Time Zero" (ED admits = time of arrival. -- Floor admits = time of admission to ICU or IMC)	
:	← Enter Patient's Height	
Complete tasks within 3 or 1 hour from Time Zero	:	← Enter "Time Zero" + 3 hours for ED patients or + 1 hour for other inpatient unit arrivals
	Severe Sepsis Resuscitation Bundle	
	Compliant?	Goal: to be done within a max. of 3 hours from "Time Zero" above or for non-ED patients within 1 hour from "Time Zero" above
		1. Measure serum lactate. (if > 2 , see #7 below) Time Drawn: ____:____
		2. Obtain blood cultures prior to antibiotic administration Time Drawn: ____:____
	3. Broad-spectrum antibiotic administration*: (see below) Time started: ____:____	
	4. Fluid bolus of 30 ml/kg Predicted Body Weight (PBW) of crystalloid IV over 1 hour PRN MAP < 65 and/or Lactate ≥ 4 mmol/L (See height to fluid bolus conversion table)	
Complete tasks within 6 hours from Time Zero	:	← Enter the above "Time Zero" + 6 hours
	Septic Shock Bundle	
	Compliant?	<ul style="list-style-type: none"> If persistent hypotension (MAP < 60) after fluid bolus proceed to step 5-6 If persistent hypotension (MAP < 60) after fluid bolus and/or most recent lactate ≥ 4, begin further volume resuscitation and vasopressor support as needed to achieve step 6
		5. Administer Vasopressors (norepinephrine is preferred if not contraindicated)
	6. Place central monitoring line and measure CVP and ScvO ₂ or use NICOM for further fluid	

Screening Sepsis Model

Goal: Identify Sepsis Patients within 2 hours of admission.

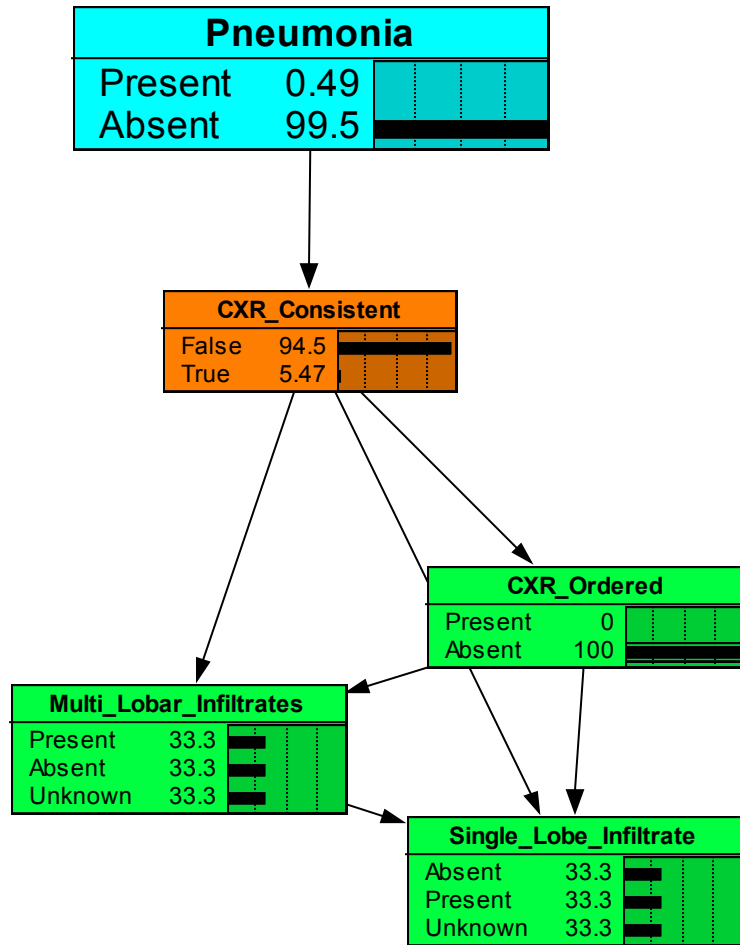


Sepsis Model: Extending Its Coverage

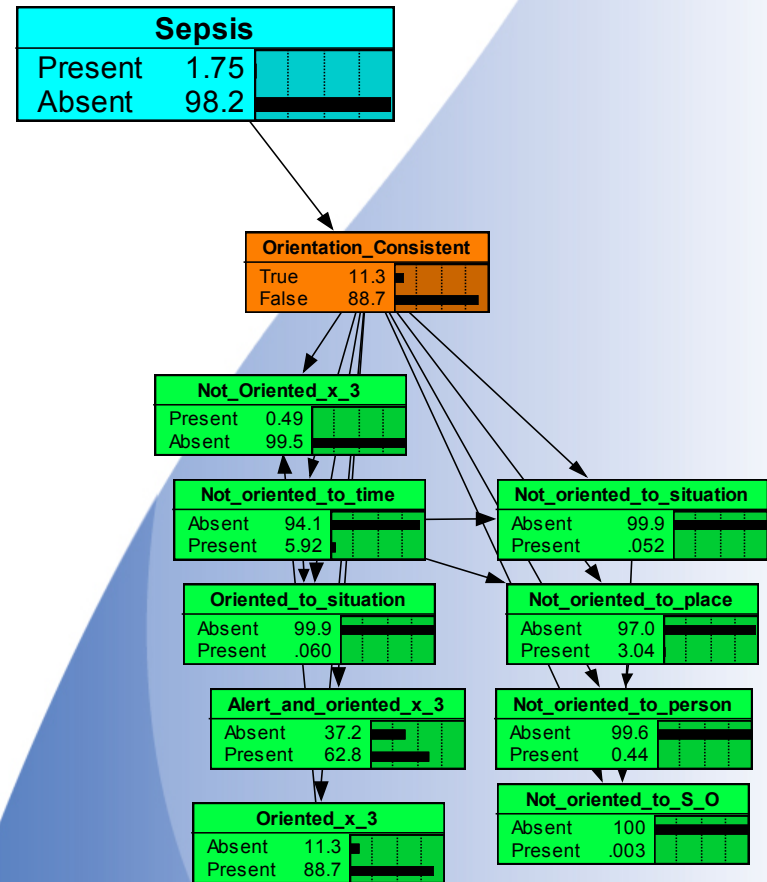


Adding Semantics to the Model

Saying "Findings are consistent with Diagnosis"



"Chest X-ray Results consistent with Pneumonia"



"Mental Status consistent with Sepsis"

Conclusion

“All models are wrong; some models are useful.”

Attributed to statistician George Box

- Probabilistic Models have a role in clinical care
- Enterprise Data Warehouses can contribute to Model development
- Stored Medical Knowledge (in the form of Ontologies) can accelerate Modeling
- Bayesian Models can flexibly represent a variety of clinical conditions.

Comments and Questions

