

SINI 2016
26th Summer Institute in Nursing Informatics
Informatics at the Crossroads of Care Coordination
July 20-22, 2016
University of Maryland School of Nursing, Baltimore, Maryland

Data Analytics: Applications in Clinical Settings

Laverne Perlie, MSN, RN, Senior Nurse Consultant
Helen Caton-Peters, MSN, RN Senior Nurse Consultant
Office of the National Coordinator
Laverne.Perlie@hhs.gov
Helen.Caton-Peters@hhs.gov

Let's get Started!
Objectives and Agenda

- Define data analytics and predictive analytics
- Discuss how to work with and understand Big Data and data patterns
- Describe how to get started with predictive analytics
- Provide an understanding on how to apply and use predictive models in clinical settings
- Describe how ONC contributes to policy on big data and analytics

Florence Nightingale (1820-1910)

- Used data and invented sophisticated predictive analysis tools, such as the polar-area diagram to transform care and save lives



3

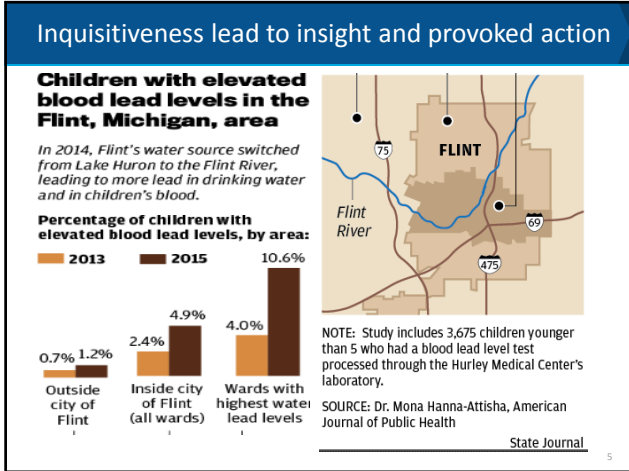
A practicing physician who was inquisitive...

- *"...if we did not have (electronic medical records), if we were still on paper, it would have taken forever to get these results."*



Mona Hanna-Attisha M.D.
Hurley Medical Center, Flint, MI

4



- ### Nursing Transforming Care
- Using nursing data science to address challenges in care delivery
 - » Big data analysis requires interoperable, standardized nursing data sets
 - Interoperable data contains data elements that are defined, measured, and retrievable in the exact same format.
 - Work with vendors, standards development organizations to incorporate nursing data into health IT in a standardized and interoperable manner.
 - » Shaping health policy
 - Nurses advocating for adoption of nursing data standards and making the case for inclusion of nursing data in value based care models.
 - Demonstrate how nursing data supports clinical decision making that improves clinical outcomes.

Define Data Analytics



Defining Analytics

- **Analytics** is the discovery and communication of **meaningful patterns** in data using **simultaneous application of statistics, computer applications and operations research.**



Other important analytic definitions

- **Decision Support** – much more than alerts and reminders! Clinical Decision Support (CDS) includes:
 - **Business Intelligence (BI)** – computerized for managerial decision making
 - **Business Performance Management (BPM)** – combines enterprise information systems (EIS) and BI for decision making. BPM feeds your rapid cycle improvement processes, such as LEAN, Six Sigma, or Plan-Do-Check-Act, which help measure your progress toward improving your **key performance indicators**
 - **Visual analytic tools** – Scorecards, dashboards (*with drill down capabilities*)
 - **Predictive analytics** – Use of tools to determine the probable future outcome or likelihood of a situation occurring.



Building Blocks: Understanding DIKW

- **Data** = vital sign data, static values
- **Information** = electronic medical record
- **Knowledge** = analytics
- **Wisdom** = application



Define the Progression from Descriptive to Prescriptive Analytics



What is predictive analytics?

- Predictive analytics is the practice of extracting information from existing data sets in order to determine patterns and predict future outcomes and trends.
- It does not tell you what will happen in the future.
 - It **forecasts what might happen** in the future with an acceptable level of reliability, and includes what-if scenarios and risk assessment.



Punxsutawney Phil in his movie debut – Goundhog’s Day

14

Gartner Analytic Ascendancy Model

Gartner Analytic Ascendancy Model



15

Why Predictive Analytics?

Predictive Analytics Drives Predictive Care



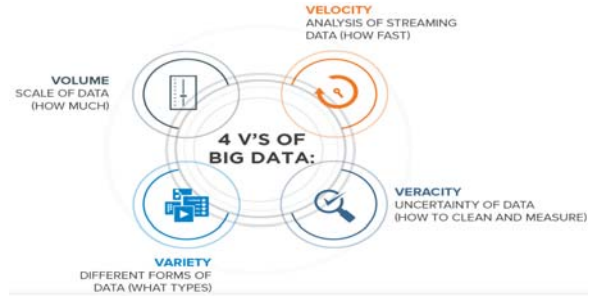
How to Understand and work with Big Data



17

Big Data definition: the 4 Vs

- **Big data** is high volume, velocity, variety & veracity information assets that require new forms of processing to enable enhanced decision making, insight discovery and process optimization.



18

Big Data provides Pattern Recognition:

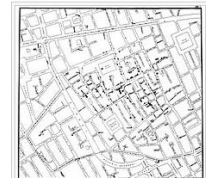
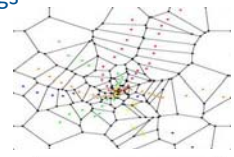
Remember C.A.R.P

- **C**lusters
- **A**ssociations
- **R**elationships
- **P**redictions

What is a Cluster?

Clusters identify natural groupings based on known characteristics

- Use optimization techniques
 - K-means (statistics) – cluster to the nearest mean & data partitioning using Voroni cells
 - Self organizing maps
- Examples:
 - Defining catchment areas for a new hospital or stroke clinic based on population health needs
 - Epidemiology – 1854 Snow's map of cholera outbreaks mapped to contaminated water pumps

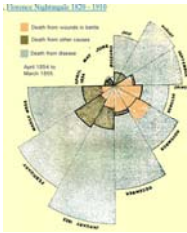


John Snow's original diagram

67

What is an Association?

- Commonly occurring groupings
 - Apriori – identifies the frequent individual items in the database and extends them to larger items in the database.
 - FP-Growth, OneR, ZeroR and Eclat (frequent item mining set), Polar Area Diagrams
- Examples:
 - “Beer and Diapers”
 - Amazon – people who bought “x” also bought “y”
 - Florence Nightingale’s association between unsanitary clinical practices and mortality – demonstrating the value of prevention in saving lives



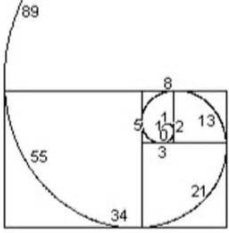
Source: Nightingale, 1859, 110-1102

What are sequential relationships?

Sequential relationships discover time ordered events

- Fibonacci Sequence
- Apriori algorithm

- Examples:
 - Population growth (F)
 - Clinical trials and chemotherapy -- oncology trials follow a modified dose escalation (F)
 - Triggers for adverse events for proactive interventions i.e. MI (A)

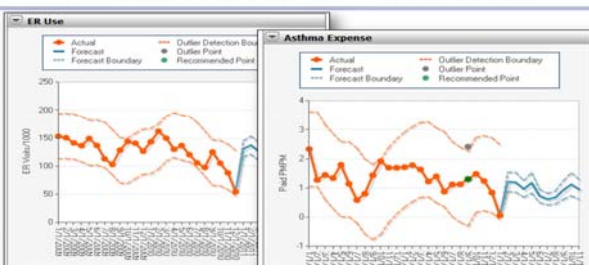


22

What are Predictions?


Predictions forecast the future based on past trends

Predictive Analytics – Care Trend



23

Getting Started Using Predictive Analytics



24

What is Predictive Modeling?

- Predictive modeling is a commonly used statistical technique to predict future behavior.
- Predictive modeling solutions are a form of data-mining technology that work by analyzing historical and current data and generating a model to help predict future outcomes.
- In predictive modeling, data is collected, a statistical model is formulated, predictions are made, and the model is validated (or revised) as additional data becomes available
- Example: identifying a patient’s risk of readmission through the use of predictive models using data abstracted from the EHR

25

Access data via an enterprise data warehouse (EDW)

26

Sources for Analytics

Assembling the data -- putting it all together:

- Clinical billing going to or claims data coming from the payer
- Pharmacy
- Biometric data
- EHRs
- Ancillary claims
- Hospital cost & use data (ADT, patient accounting, GL)

27

Apply the 3 steps of predictive modeling

28

Other “must haves”

Program Needs	Healthcare Challenge
Executive Support	Executives have to manage organization’s staff to get their cooperation and buy-in.
Well-Defined Business Challenge	Business challenges are everywhere. The real problem is prioritizing which one to address first.
Lots of Data	There’s lots of data but a lot of it is locked in departmental silos which ultimately makes all the data useless.
Right Team	The challenge will be finding qualified people in an already scarce resource pool and getting them to accept the lower wage healthcare may pay. Outsourcing might need to be an option. Bottom Line: GET HELP!
Integral Part of Organization	Everyone must buy-in to the results of the analytics program including clinical, finance and operational staff.
Track Results and Update Models	With the right team in place this should not be an issue.

29

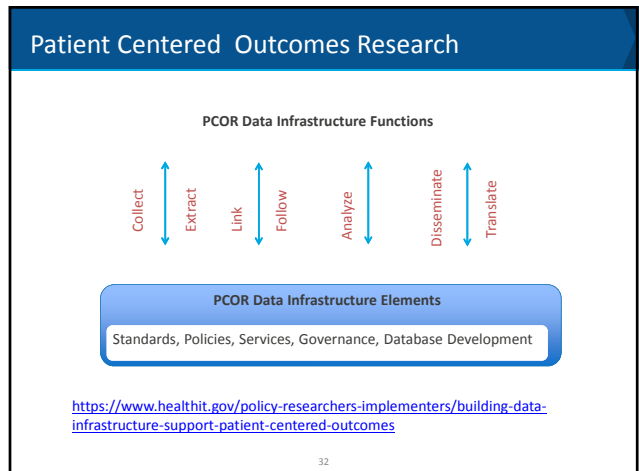


The Learning Health System

National LHS Requirements


A. Governance of Policy & Standards	H. Consistent representation of authorization to access health information
B. Interoperable Business and Regulatory Environment	I. Stakeholder Assurance that health IT is interoperable
C. Individuals are empowered	J. Consistent data formats and semantics
D. Care Providers partner with individuals	K. Standard, secure services
E. Ubiquitous, secure network infrastructure	L. Consistent, secure transport techniques
F. Verifiable identify and authentication	M. Accurate identify matching
G. Consistent representation of permission to collect, share and use identifiable health info	N. Reliable Resource Location

31



PMI

THE PRECISION MEDICINE INITIATIVE®



- **The Precision Medicine Initiative** -- An approach for disease prevention and treatment that takes into account both the population and individual variations in genes, environment, lifestyle, etc.
 - » Creates a research cohort of **>1 million American volunteers** who will share genetic data, biological samples, and diet/lifestyle information, all linked to their electronic health records if they choose.
 - » Pioneers a **new model for doing science** that emphasizes **engaged participants, responsible data sharing, and privacy protection.**
 - » Tests whether **mobile devices** can encourage healthy behaviors.
 - » Lays **scientific foundation** for precision medicine for **many diseases.**

33

Treatment, Research and Population Health Developments


- Clinical Data Registries
 - » Analyzing data from multiple sources to improve care
 - Developing new practice guidelines
- Zika Response
 - » EHR's and public health response plan
 - Developing algorithms, vocabulary sets, order sets
 - » Global response - OpenZika
- Precision medicine vs. traditional medicine
 - » Treatment protocols that involve precision medicine and personalized approaches tend to lead to improved patient wellness and prolonged periods of remission (University of California San Diego)

34

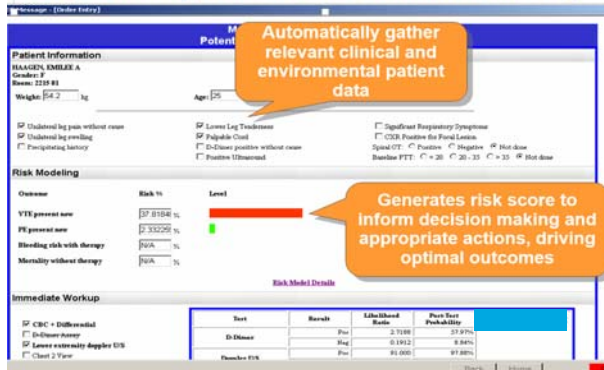
Clinical Examples:

Predicting

- Deep Vein Thrombosis
- 30-day Readmissions
- Patient Flow
- Disease Outbreaks



Deep Vein Thrombosis: Augmenting clinical decision making



Automatically gather relevant clinical and environmental patient data

Generates risk score to inform decision making and appropriate actions, driving optimal outcomes

Test	Result	likelihood Ratio	Post-test Probability
D-Dimer	Pos	2.7183	33.33%
Lower extremity Doppler US	Neg	0.1912	0.34%
Pre-test PR	Pos	0.1000	0.100%

Personalized Medicine Projecting Best Course of Care

Test Results

Test	Result	Lab/Ref Range	Post Test Probability
D-Dimer	Pos	2.7188	57.97%
Duplex US	Mag	0.1912	9.84%
Duplex US	Pos	91.000	97.80%
Contract Venography	Pos	99.000	98.02%
Contract Venography	Mag	0.0100	0.02%

Therapy Factors

Warfarin Initiation

Suggested Warfarin Regimen

Tablet dose: 1 mg, 2 mg, 2.5 mg, 3 mg, 5 mg, 10 mg, Out-pile

Actual average daily dose: 3.6 mg

Suggests patient-specific warfarin regimen based on individual genetic and environmental factors

Patient Flow: Is the patient tracking according to an expected cohort?

Patient Progression provides visibility and actions to advance care and manage deviations

Care algorithms keep a patient's care on track

Outcomes Likelihood (%)

- PE Diagnosis: 7%
- PE Future: 1%
- PE Mortality (7 Days): 1%
- PE Recurrence: 1%
- Antithrombotic Bleed: 7%

Therapy Treatment - Prescriptions & Treatment Choices

Drug/Nutrient	Dose	Schedule	Route	Purpose	Hold the CV
Warfarin C	3.6 mg	QD	PO	anticoagulant	
Aspirin	81 mg	QD	PO	antiplatelet	

Some Popular Readmission Reduction Predictive Models

Model	Description	Details
LACE	Length of Stay; Acuity; Comorbid condition; ED visits in 6 months	1-19 points. Patients with >10 points singled out for follow up
STAAR	State Action on Avoidable Re-Hospitalization	iHI initiative – focuses on improved transitions of care
Project RED	Re-engineered Discharge	Developed at Boston Univ Med Ctr (BUMC). An 11-step nurse-led process
H2H	Hospital to (2) Home	American College of Cardiology & iHI to reduce readmissions & improve care transitions for cardiovascular patients
BOOST	Better Outcomes by Optimizing Safe Transitions	8-PS (1. Problems with medications; 2. Psychological; 3. Principal diagnosis; 4. Psychosocial limitations; 5. Poor health literacy; 6. Poor social support; 7. Prior hospitalization; 8. Palliative care)
Custom	Develop own predictive model	Tools: Statistica R, Strata, MATLAB, Vensim

Predictive analytics is used to calculate risk scores

Patient Name	MRN	Adult Type	Gender	Age	Risk Score	Phone
Perkins, Sherene Kiyah	301420120249	Urgent Care	Female	62	71	(208) 577-7062
Ortiz, Yolanda Krystal	301420120252	Urgent Care	Male	73	111	(208) 498-4239
Harris, Margaretta Quibeth	301420120254	Urgent Care	Female	40	11	(208) 552-9769
Osby, Dora Dominique	301420120258	Urgent Care	Male	79	89	(208) 663-0623
Perkins, Kaiti Malai	301420120259	Urgent Care	Female	39	84	(208) 559-4857
Bull, Jerry Jale	301420120264	Urgent Care	Female	86	81	(208) 604-8927
Truitt, Amy Cheryl	301420120272	Urgent Care	Male	72	80	(208) 604-8728
Victor, Anita Scott	301420120276	Urgent Care	Male	67	79	(208) 279-7121
Johnson, Geudreco Chatherine	301420120288	Urgent Care	Female	62	77	(208) 649-5242
Smith, Fred Gerald	301420120289	Urgent Care	Female	60	74	(208) 275-2942
Vasquez, Prudencio Sierra	301420120296	Urgent Care	Male	72	74	(208) 555-1328
Castro, Heira Tanyanka	301420120308	Urgent Care	Male	37	73	(208) 713-7960
Jedryka, Ronale Alex	301420120327	Urgent Care	Female	41	79	(208) 668-2876

Specific Heart Failure Cohort
 Regression based, pre-calculated
 Current PPV (precision) is 91.9%
 Integrated in current workflow

We are at the finish line!

- We defined analytics & predictive analytics
- We talked about Big Data and data patterns
- We talked about how to get started
- We talked about using predictive analytics in the clinical setting
- We described policy efforts

Thoughts to leave you with...

- Who are data scientists? - Clinicians, epidemiologists, public health officials, policy specialists, etc., can work with big data – asking questions, seeking answers, and making a difference
- Data validity begins at the point of data capture. The value and reliability of data are only as good as its validity.
- Use C.A.R.P. to detect patterns
- Data, information, knowledge and wisdom can transform care – as we move from predictive analytics to prescriptive care.

Be the change you wish to see in the world.

- Mahatma Gandhi



Thank you for:

- Your time,
- Your attention,
- Your role as being inquisitive agents for change, and
- For the work you do everyday!