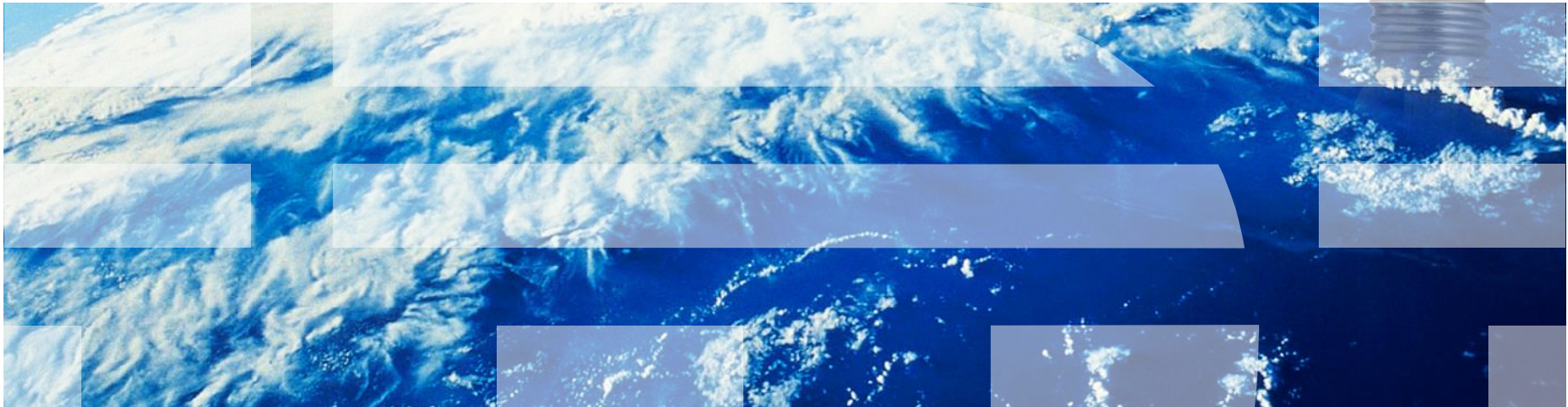
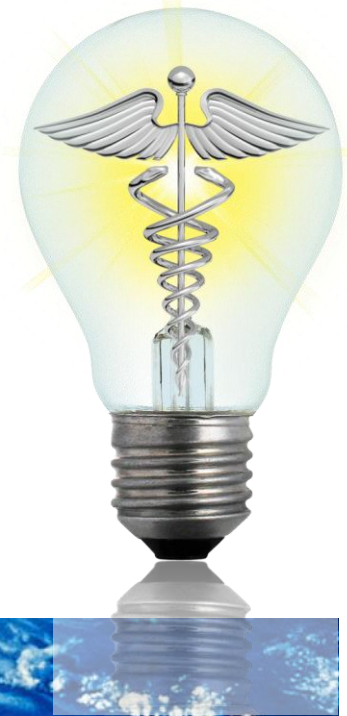


IBM Advanced Care Insights:

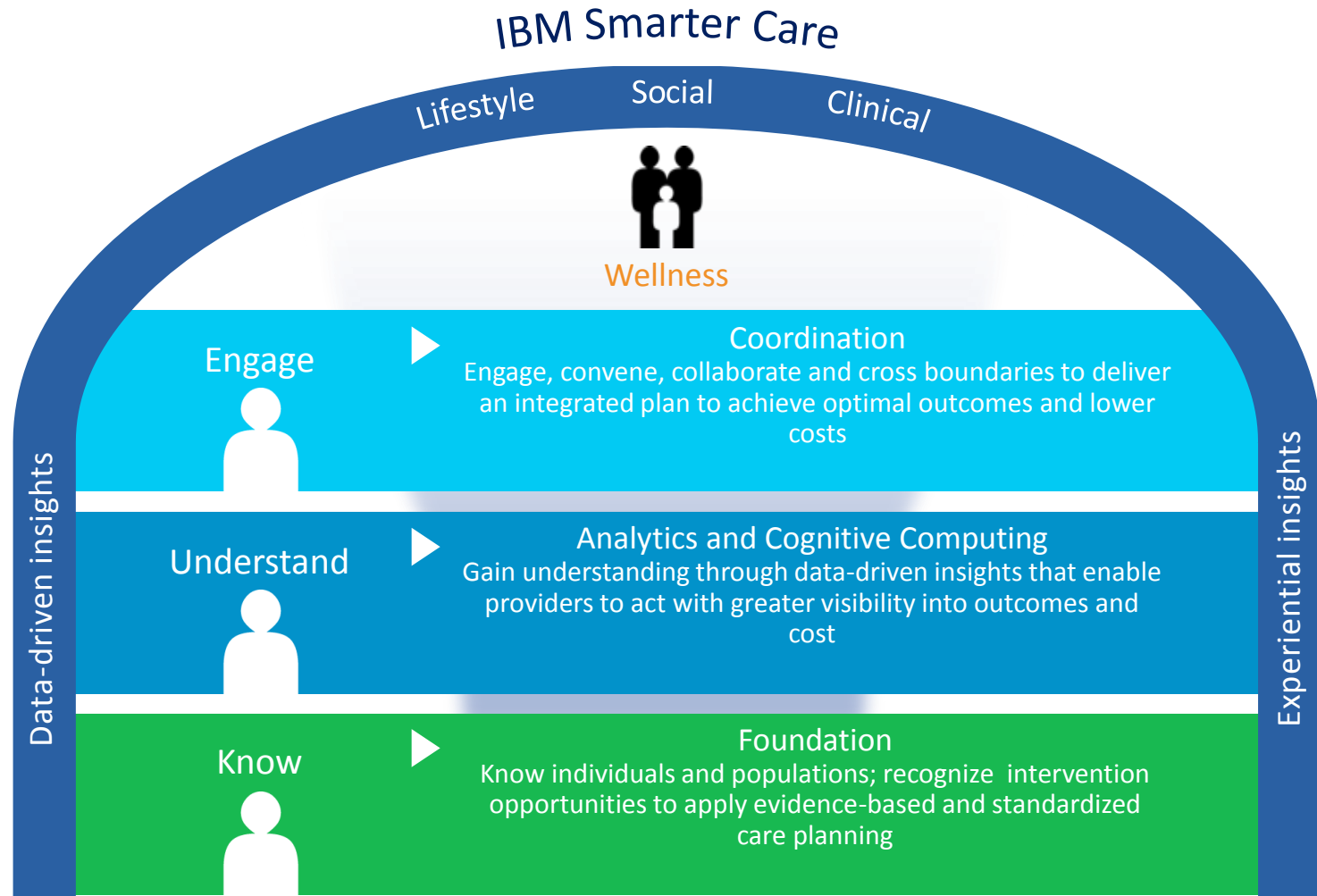
Analytics and Care Management to Reduce Readmissions

Paul Hake MSPA ([phake@us.ibm.com](mailto:phake@us.ibm.com))



# The path forward

... enabling holistic and individualized care to optimize outcomes and lower costs



# IBM integrated portfolio for Smarter Care

## Coordination

Care identification

Care planning

Care collaboration

Outcome evaluation

## Analytics and Cognitive Computing

Population analytics

Diagnostic support

Care pathways

Operational reporting

Cognitive computing

## Foundation

Data warehouse and data models

“Single view” customer EMPI (MDM)

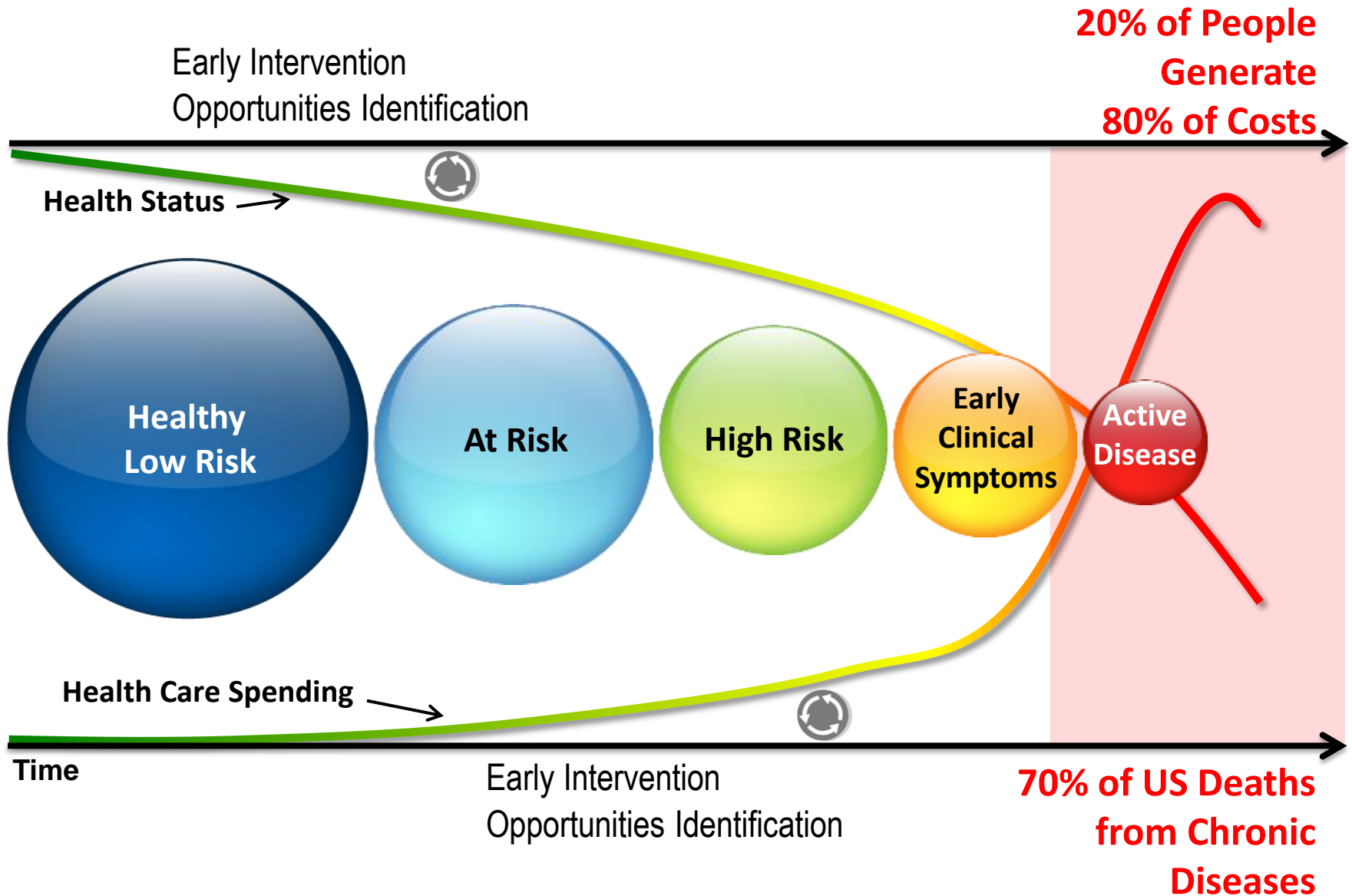
BI, reports and dashboards

Portals, mobile and collaboration

Remote monitoring and medical device connectivity

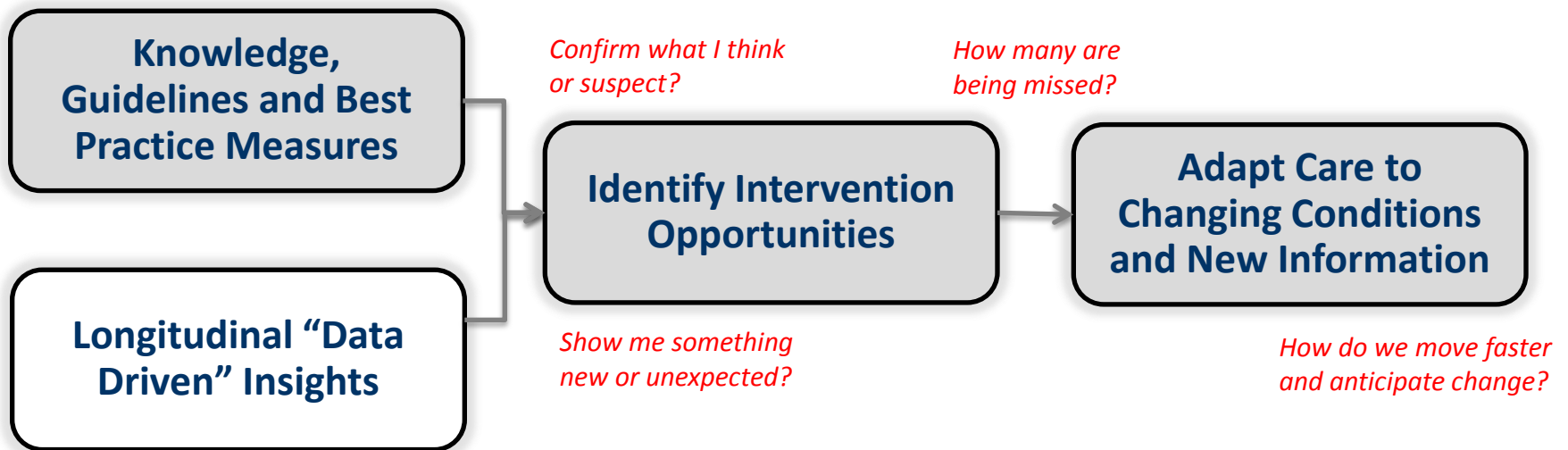
Paper and Fax capture, conversion and extraction

Comprehensive global consulting, technology, infrastructure and managed services



# Information Should Aid Us, Not Lie Hidden and Dormant

*If we could only activate the relevant information to bring insights to the point of care when needed most ...*



**Time once spent manually interpreting data ... becomes time spent healing patients**

- Aggregate, activate and enrich relevant patient information beyond what is known
- Surface new data driven insights that enable new intervention opportunities ... earlier
- Adapt to changes and proactively deliver individualized patient centered care

# What were the Readmissions Predictors at Seton?

## The value of adding unstructured Data

### The Data We Thought Would Be Useful ... Wasn't

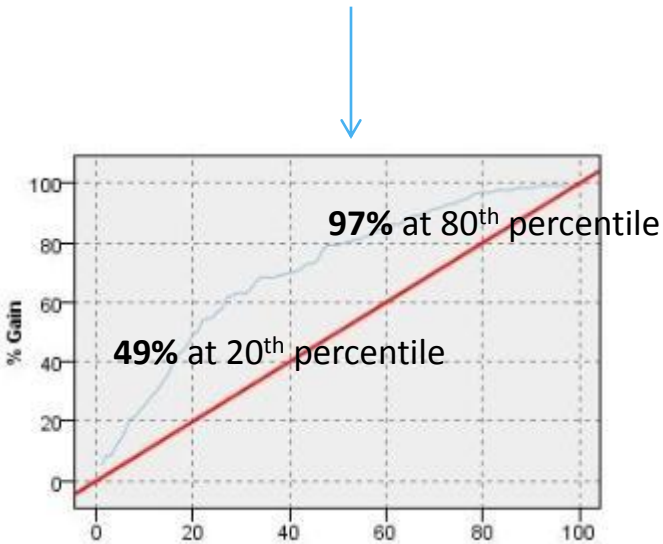
- Structured data not available, not accurate enough, without the unstructured data - which was more trustworthy

### What We Thought Was Causing 30 Day Readmissions ... Wasn't

- 113 possible candidate predictors expanded and changed after mining the data for hidden insights

### New Hidden Indicators Emerged ... Readmissions is a Highly Predictive Model

- 18 accurate indicators or predictors (see next slide)

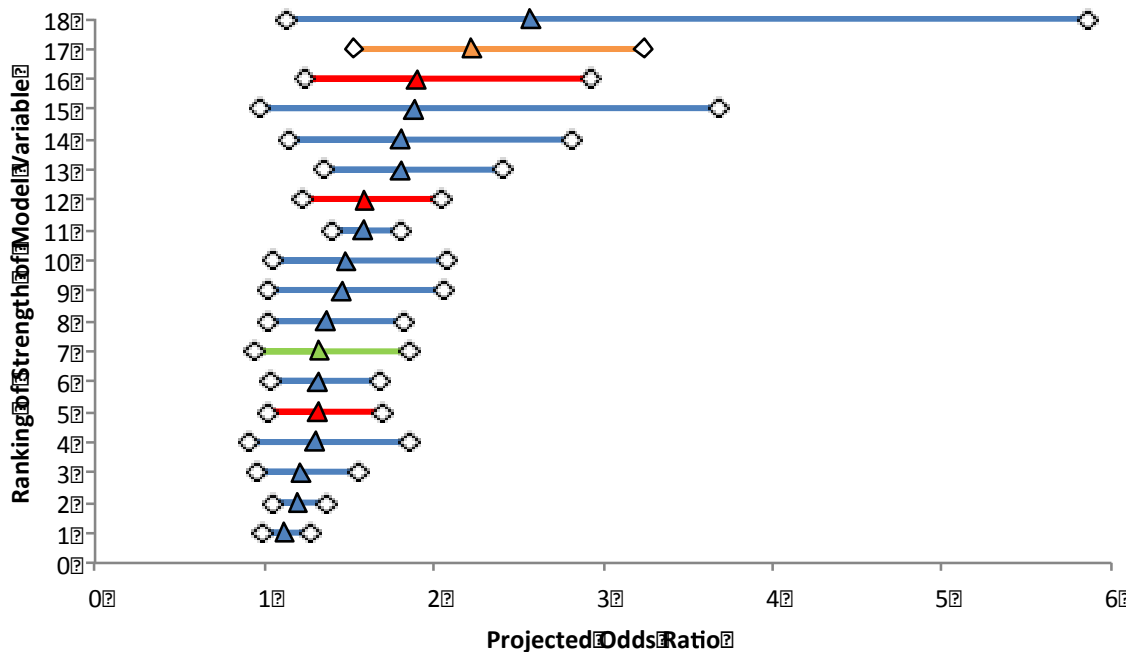


Predictor Analysis	% Encounters Structured Data	% Encounters Unstructured Data
Ejection Fraction (LVEF)	2%	74%
Smoking Indicator	35% (65% Accurate)	81% (95% Accurate)
Living Arrangements	<1%	73% (100% Accurate)
Drug and Alcohol Abuse	16%	81%
Assisted Living	0%	13%

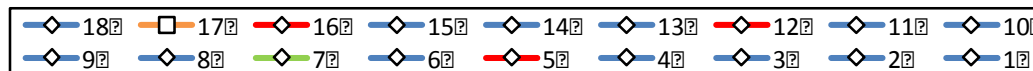
# Readmissions at Seton - Top 18 Indicators

## New Insights Uncovered by Combining Content and Predictive Analytics

- Top indicator JVDI not on the original list of 113 - as well as several others
- Assisted Living and Drug and Alcohol Abuse emerged as key predictors - only found in unstructured data
- LVEF and Smoking are significant indicators of CHF but not readmissions
- A combination of actionable and non-actionable factors cause readmissions



- 1. Jugular Venous Distention Indicator**
2. Paid by Medicaid Indicator
3. Immunity Disorder Disease Indicator
4. Cardiac Rehab Admit Diagnosis with CHF Indicator
5. Lack of Emotion Support Indicator
6. Self COPD Moderate Limit Health History Indicator
7. With Genitourinary System and Endocrine Disorders
8. Heart Failure History
9. High BNP Indicator
10. Low Hemoglobin Indicator
11. Low Sodium Level Indicator
- 12. Assisted Living**
13. High Cholesterol History
14. Presence of Blood Diseases in Diagnosis History
15. High Blood Pressure Health History
- 16. Self Alcohol / Drug Use Indicator**
17. Heart Attack History
18. Heart Disease History





# The Impact of Readmissions at Seton

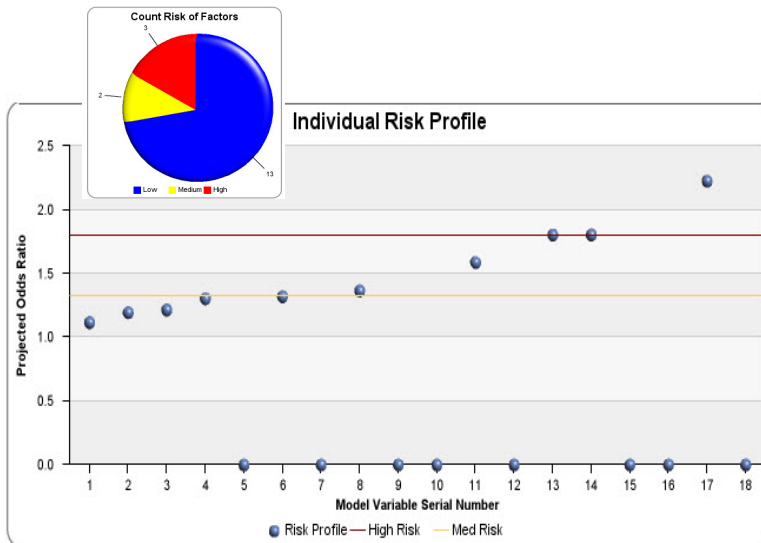
## CHF Patient X – What Happened?

- Admit / Readmission
- 30-Day Readmission

Patient X was hospitalized **6 times** over an **8 month period**. The same basic information was available at each encounter and Patient X's readmission prediction score never dropped below **95%** (out of possible 100%)



### Individual Patient Data at Each Encounter (Patient X @ Dec 20, 2009)



### Description of Model Serial Number

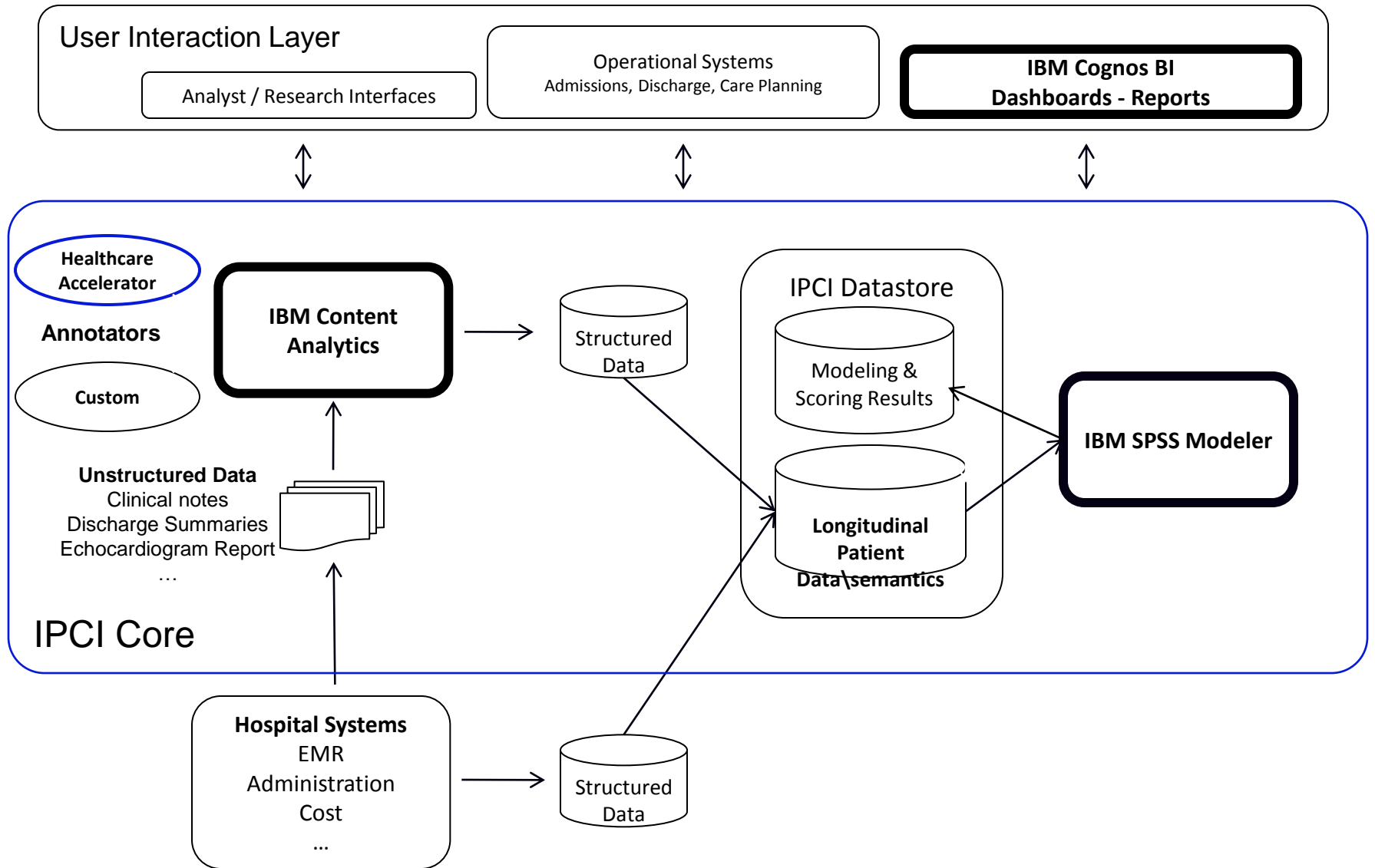
18. Jugular Venous Distention Indicator
17. Paid by Medicaid Indicator
16. Immunity Disorder Disease Indicator
15. Cardiac Rehab Admit Diagnosis with CHF Indicator
14. Lack of Emotion Support Indicator
13. Self COPD Moderate Limit Health History Indicator
12. With genitourinary system & Endocrine disorders
11. Heart Failure History
10. High BNP Indicator
9. Low Hemoglobin Indicator
8. Low Sodium Level Indicator
7. Assisted Living from ICA Extract
6. High Cholesterol History
5. Presence of diseases of the blood in diagnosis history
4. High Blood Pressure Health History
3. Self Alcohol/Drug Use Indicator (Cerner + ICA)
2. Heart Attack History
1. Heart Disease History

### Patient Population Monitoring Clinical and Operational Data





# Illustrative CHF Readmissions Architecture



- Annotators are used to identify valuable facts in unstructured documents (e.g. clinician notes, consult reports, free text fields in EMRs) and convert to a structured form
- Annotators execute in a sequence called the UIMA or Unstructured Information Management Architecture pipeline
- IBM Patient Care and Insights Annotators use UMLS to normalize discovered facts to coding systems
- Excellent application training services / annotators can be developed in IBM Content Studio



## Unified Medical Language System

The screenshot displays the IBM Content Analytics Studio interface. The main window shows a document titled "Transcribed Medical Transcription Sample Reports and Examples" with a sample name "Consult - Congestive Heart Failure". The description text is annotated with red boxes highlighting key medical terms and phrases. The right-hand pane, titled "Outline", shows a hierarchical list of annotations categorized by type, including "com.ibm.en.Age", "com.ibm.en.DictDrugIngredient", "com.ibm.en.EjectionFraction", "com.ibm.en.EjectionFractionValue", "com.ibm.en.Frequency", "com.ibm.en.Ind\_Drug", "com.ibm.en.Ind\_MedAmount", "com.ibm.en.LabValueInd", "com.ibm.en.Measurement", "com.ibm.en.MedicationInd", and "com.ibm.en.ProblemInd". The "ProblemInd" category is expanded, showing specific annotations like "250.00", "Congestive Heart Failure", "Congestive heart failure", "shortness of breath", and "fatigue".

# Healthcare Annotators example

## Problems

- Result of a series of interim annotations that identify diseases, symptoms, and disorders
- Normalize to standard terms and standard coding systems including SNOMED CT, ICD-9, HCC, CCS
- Capture timeframes of the problem
  - determine if past or current problem
- Determine confidence
  - Positive, Negative, Rule Out, etc.
  - Negation example
    - “abdominal pain”

HEENT: History of blurry vision and hearing impaired. No glaucoma.  
 CARDIOVASCULAR: Shortness of breath, congestive heart failure, and arrhythmia. Prior history of chest pain.  
 RESPIRATORY: Bronchitis and pneumonia. No valley fever.  
 GASTROINTESTINAL: No nausea, vomiting, hematemesis, melena, or abdominal pain.  
 UROLOGICAL: No frequency or urgency.  
 MUSCULOSKELETAL: No arthritis or muscle weakness.  
 SKIN: Non-significant.  
 NEUROLOGICAL: No TIA. No CVA or seizure disorder.  
 ENDOCRINE: Non-significant.  
 HEMATOLOGICAL: Non-significant.  
 PSYCHOLOGICAL: Anxiety. No depression.  
 PHYSICAL EXAMINATION:  
 VITAL SIGNS: Pulse of 60, blood pressure of 120/70, afebrile, and respiratory rate 16 per minute.

Property	Value
conceptid	@ 21522001
confidence	@ negative
Covered text	@ abdominal pain
hccCode	@ empty
icd9	@ 789.00
modifiers	@
normalized	@ abdominal pain
origin	@ symptom



## Risk-stratifying patients

- Focus costly, resource-intensive interventions on patients who are at highest risk
- Example: nurse home visits, home tele-monitoring

## Risk prediction models

- Performance is generally poor
- Typically only use structured EMR and/or claims data
- Psycho-social determinants of readmission risk usually not in structured data
- Free-text diagnostic test results not included in risk model

## Structured

Age  
Gender  
Race/ethnicity  
Insurance type  
Diagnoses (ICD-9)  
Vital signs  
Laboratory results  
# previous readmissions  
LOS on previous hospitalizations  
# Medications



## Unstructured

- Physical exam findings
- Diagnostic test results
- Psycho-social factors
  - Lives alone, homeless
  - Substance abuse
  - Medication non-compliance
  - Estrangement from family/friends
  - Depression

SOCIAL HISTORY: Current smoker - pipe, as well as 10 cigarettes to 1/2 pack per day at present (2ppd since 14). Occasional alcohol use, but history of heavy alcohol use "up to a gallon per day until 1975" per the patient. Denies illicit drug use. Lives alone near Haw River.

SH: Currently unemployed but not on disability. Lives in Henderson, NC with his mother and father. Originally from Mexico, moved to US about 7-9 years ago. Previously smoked cigarettes, ~1 pack per week x10 years, quit age 22. Previously drank heavily in Mexico, ~2 bottles tequilla on each weekend night x3 years, quit age 16. Prior marijuana use in high school. Used cocaine once.

## Social History:

He quit smoking 22 years ago, prior to this, smoked 10 cigarettes per day since his teenage years. Denies any alcohol or other drug use. He has been married for 29 years and lives with his wife.

## Interpretation:

### Clinical Diagnoses and Echocardiographic Findings

Inferoposterior myocardial infarction

Anteroseptal and apical myocardial infarction

Decreased left ventricular ejection fraction (35-40%)

Dilated left ventricle

Diastolic left ventricular dysfunction

Elevated left ventricular filling pressures

Degenerative mitral valve disease

Mitral annular calcification

Mitral regurgitation (mild)

Dilated left atrium

Aortic sclerosis

Aortic regurgitation (trivial)

Pulmonary hypertension (moderate to severe - see detail below)

Segmental right ventricular contractile dysfunction (see detail below)

Tricuspid regurgitation (mild)

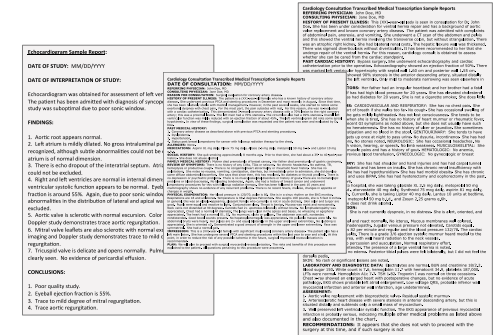
Elevated central venous and right atrial pressures (see detail below)



# What Have We Learned So Far?

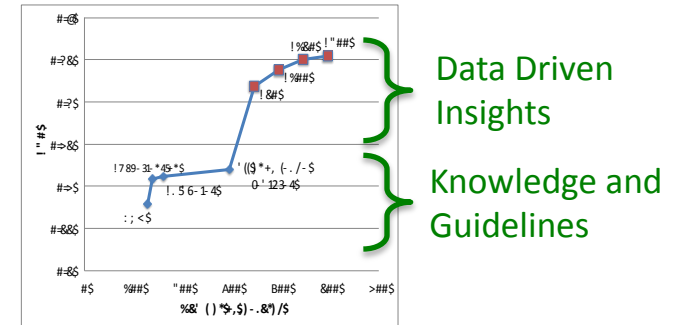
## Structured Data is Not Enough

- Unstructured data significantly increases the richness and accuracy of analysis and decision making ... including paper / faxes



## Today's Care Guidelines Only Get You So Far

- Not granular enough to deliver on the promise of personalized medicine with data driven insights <sup>1, 2</sup>



## Manual Processes and Traditional Workflow Approaches Don't Work

- Process complexity increases with disease complexity ... changing conditions require process adaptability <sup>3</sup>



1. Dijun Luo, Fie Wang, Jimeng Sun, Marianthi Markatou, Jianying Hu, Shahram Ebadollahi, SOR: Scalable Orthogonal Regression for Low-Redundancy Feature Selection and its Healthcare Applications. SDM'12  
 2. Jimeng Sun, Jianying Hu, Dijun Luo, Marianthi Markatou, Fei Wang, Shahram Ebadollahi, Steven E. Steinhilb, Zahra Daar, Walter F. Stewart. Combining Knowledge and Data Driven Insights for Identifying Risk Factors using Electronic Health Records. Under submission at AMIA'12  
 3. Blind Surgeon Metaphor Problem - W.M.P. van der Aalst, M. Weske, and D. Grünbauer. Case Handling: A New Paradigm for Business Process Support. Data and Knowledge Engineering, 53(2):129-162, 2005

A **Configurable Solution** designed to surface evidence based insights from longitudinal data that enables advanced population analysis, personalized interventions and proactive care delivery in complex and costly disease scenarios. Supporting doctors treating patients in collaborative care models with process complexity, interventions and care transitions.

## Configurable Solution Options

### Advanced Care Insights Solution Models

- Readmission Prediction and Prevention
- Condition Onset or Deterioration Prediction and Prevention
- Drug Treatment Efficacy and Effectiveness
- Physician, Care Team or Resource Matching
- Resource Utilization Pattern and Anomaly Detection
- Risk Adjusted Scoring Improvement
- Care Pathways Adherence and Deviation

### Care Management Solution Plans

- Disease and Scenario Specific Care Plans and Templates

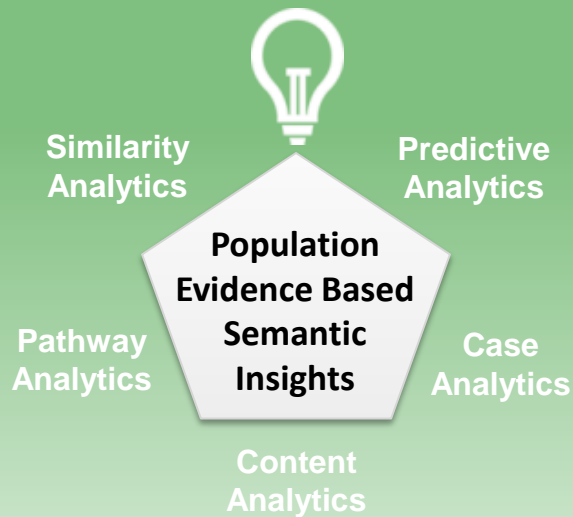
### Visualizations

- Care Pathway Flows
- Custom Population Analysis

### User Experience, Dashboards and Reporting

- Case Performance Analysis and Monitoring
- Semantic Powered Search

## Advanced Care Insights



## Care Management

### Care Management Solution



- Assess
- Plan
- Deliver
- Monitor
- Audit
- Analyze

### Care Management Platform

# Reducing Readmissions with targeted care management – Catalonia Region in Spain



- Nationalized Healthcare – Government Payor
- Healthcare Provider for the region of Catalonia
- ~7 million residents served
- \$4 Billion annual budget
- 8 Hospitals, 4500 beds, 130 OR, 450 primary care centers

- Existing IBM customer since 2005
  - SAP implementation for clinical healthcare and financial (8 ICS Hospitals)
- Smarter Care proof-of-concept delivered Dec'2012
- Phase 1 live March'2013
  - 300 patients, 10-20 Care Coordinators, 30-40 Doctors and others
  - Developed in 8 weeks



*Spain's most prosperous region*

To achieve the main objectives, care systems must **focus on areas of highest impact**

Improve **quality** of care



Lower **costs** of care



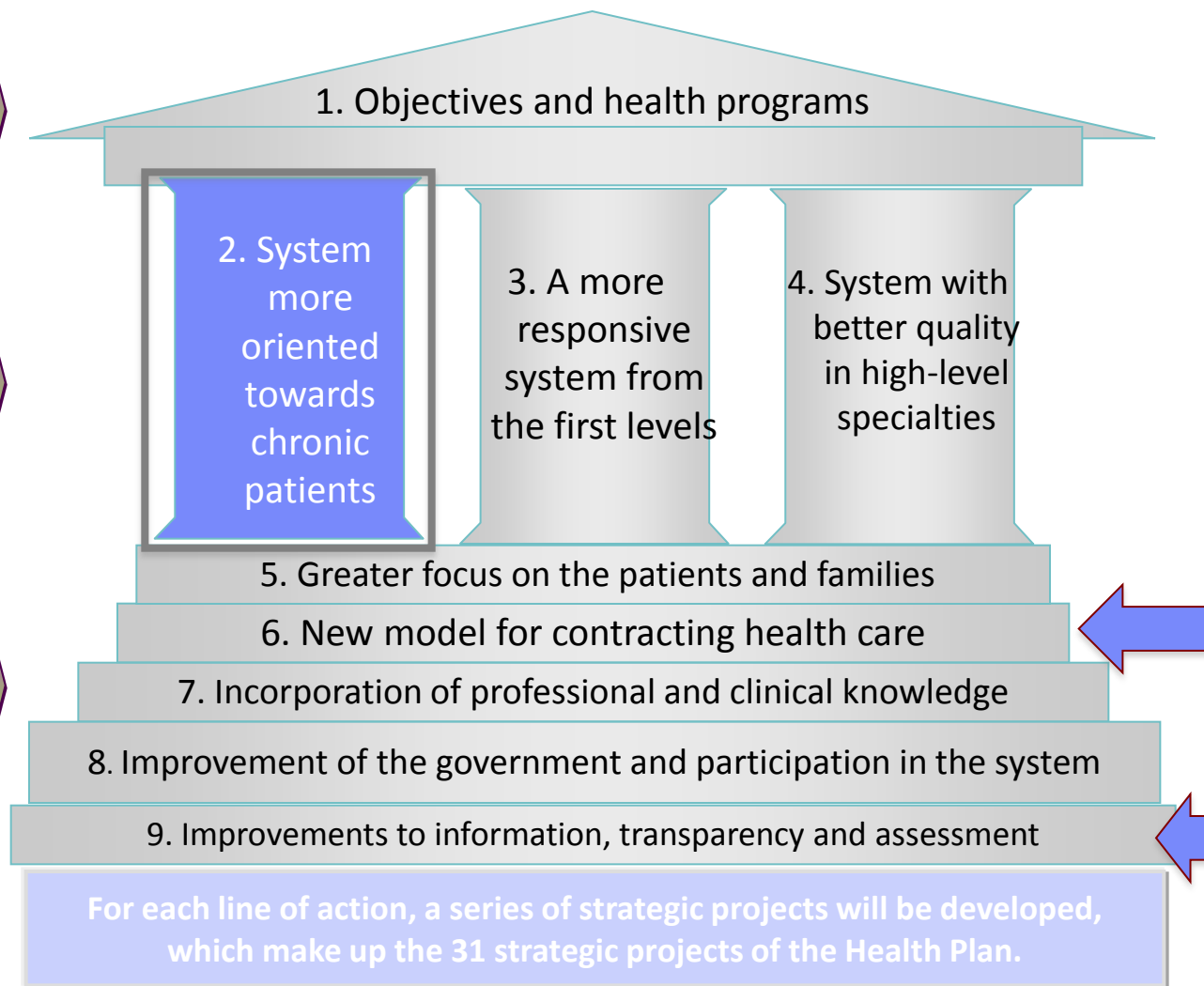
- **25% of population over 65 years**, 60% have chronic diseases and consume 70% of healthcare resources
- Complex needs require care by **providers across disciplines, acting as a team**
- Over time, progress must be tracked and care plans refined **to achieve desired outcomes**

## 3 pillars of transformation

I Health Programs:  
*Better health and quality of life for everyone*

II Transformation of the care models:  
*better quality, accessibility and safety in health procedures*

III Modernize the organizational models:  
*a more solid and sustainable health system*



## Catalonia Care Management Functional Objectives

### Approach care holistically

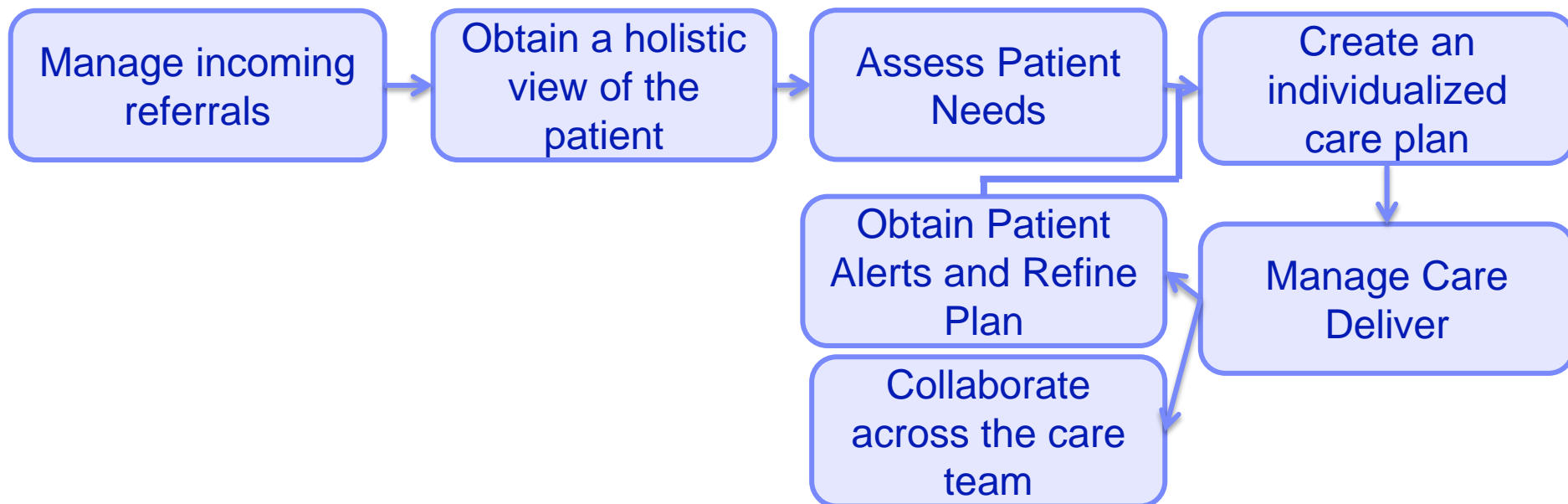
- Implement a care management program to effectively manage care
- **Overcome fragmented views of health** *Physical, mental, nutrition, education, employment & income, safety, family & community, living conditions*
- Design care approaches to **address holistic needs** of the patient

### Manage care plans for better outcomes

- **Reduce Aggressive Treatments:** Increase homecare, Reduce A&E cases, Reduce inpatient cases
- **Collaborate and coordinate all stakeholders**  
*Care providers, activities, services, medication, equipment*
- **Improve adherence** to care management program

### Empower Patients

- **Improve therapeutic adherence**
- **Increase the patient co-responsibility** in his/her care
- **Improve patient satisfaction** with the healthcare system





Indicators	Objective
<b>Prevalence</b> recruiting Complex Chronic Patients (PCC) and Advanced Chronic Patients (MACA)	At least to double PCC and MACA prevalence comparing with the rest of control territories
Proportion of PCC/MACA patients with a related <b>activated/reviewed Care Plan</b>	More than 70% patients with a Care Plan
<b>Avoidable emergency admission:</b> COPD / Heart failure / “composite”	Decreasing by 10%
<b>30-day Readmission:</b> COPD / Heart Failure / “Composite”	Decreasing between 5-10%
Mean <b>number of contacts with PHC services</b>	Increasing contacts with PHC by 15%
Patient <b>Satisfaction</b>	Satisfaction over 85 score
Introduction of <b>Quality of Life (Euroqol)</b> measure	Improvement Euroqol score
Regular <b>Medication Plan review</b>	Over 80% medication plan reviewed at least 2 times a year



- 1) Predictive Modeling/Risk identification is not enough to reduce readmissions
- 2) Care Management is equally as important
- 3) NLP can help augment both
- 4) Platform approaches integrating all 3 look promising

Thank You