



# Translational Research for Surveillance

Integrated Surveillance Seminar Series  
from the  
National Center for Public Health Informatics  
January 28, 2008

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<http://essence.jhuapl.edu/ESSENCE>





# Outline

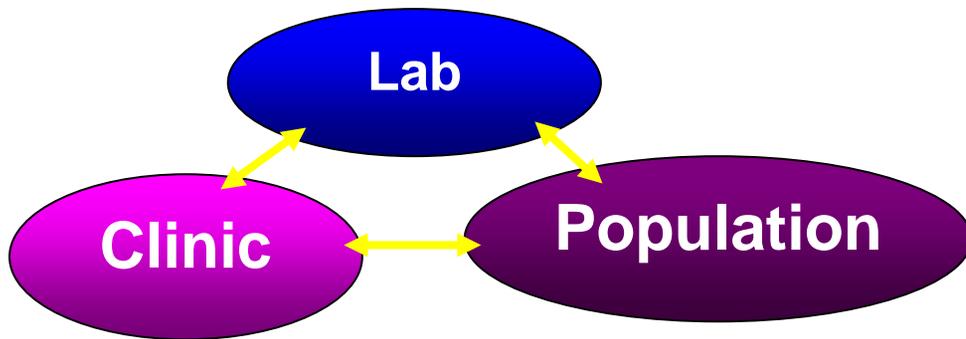
1. A definition of public health informatics translational research
2. Identify translational research opportunities in PHI needed to improve public health practice in surveillance
3. Introduce selected surveillance enhancement projects
  - a. Advanced querying to rapidly create more productive and timely analysis groupings (moving from syndromic to case specific surveillance)
  - b. Customizable alerting analytics
  - c. Public health collaborations (overcoming data sharing obstacles)
4. Discussion



# NCI Translational Research Defined

## National Cancer Institute Technical Working Group Definition

"Translational research transforms scientific discoveries arising from laboratory, clinical, or population studies

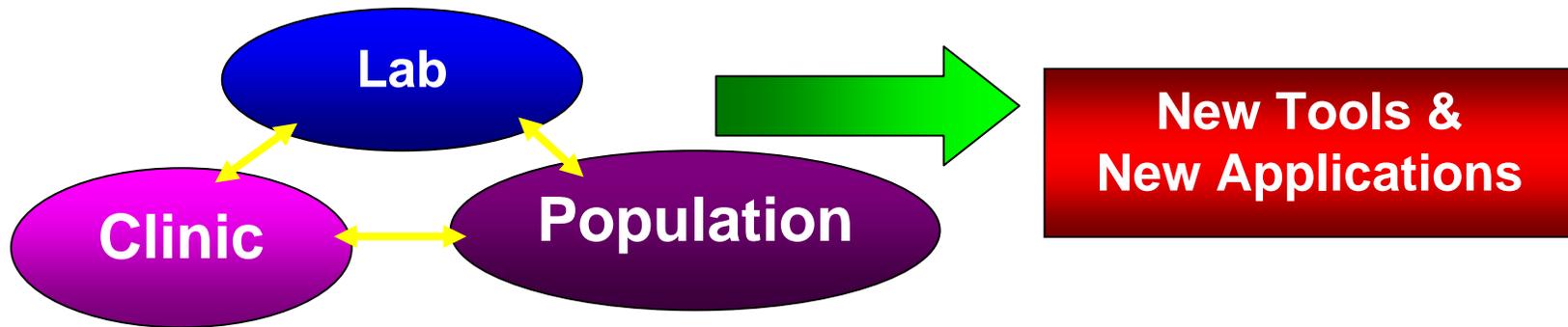




# NCI Translational Research Defined

## National Cancer Institute Technical Working Group Definition

"Translational research transforms scientific discoveries arising from laboratory, clinical, or population studies into clinical applications

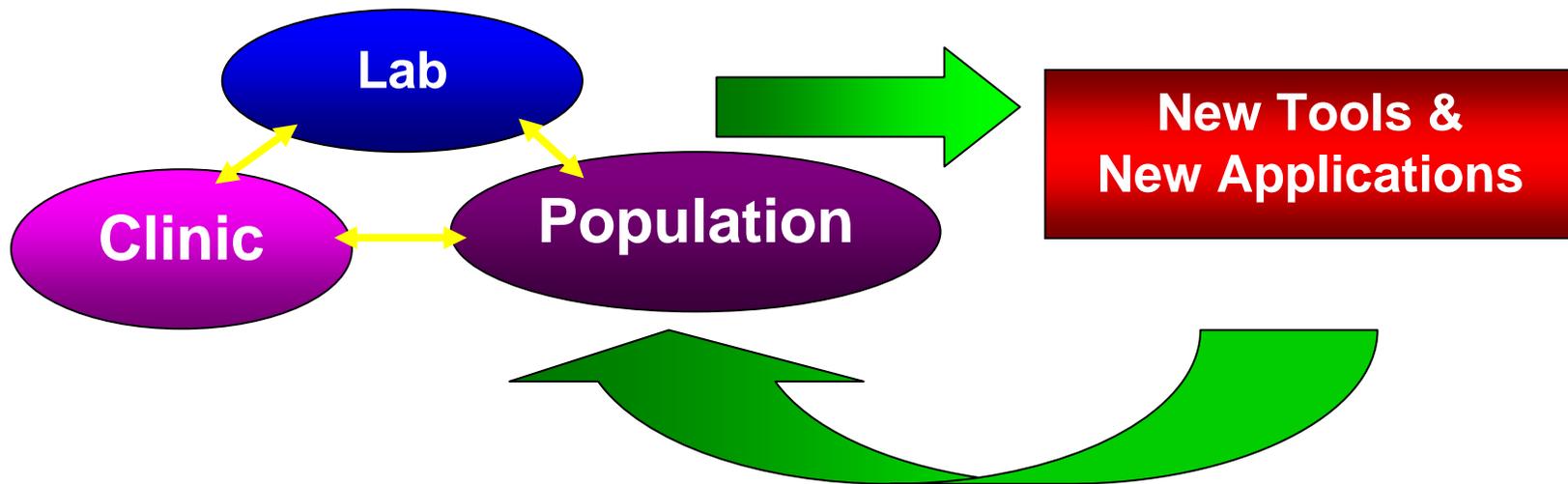




# NCI Translational Research Defined

## National Cancer Institute Technical Working Group Definition

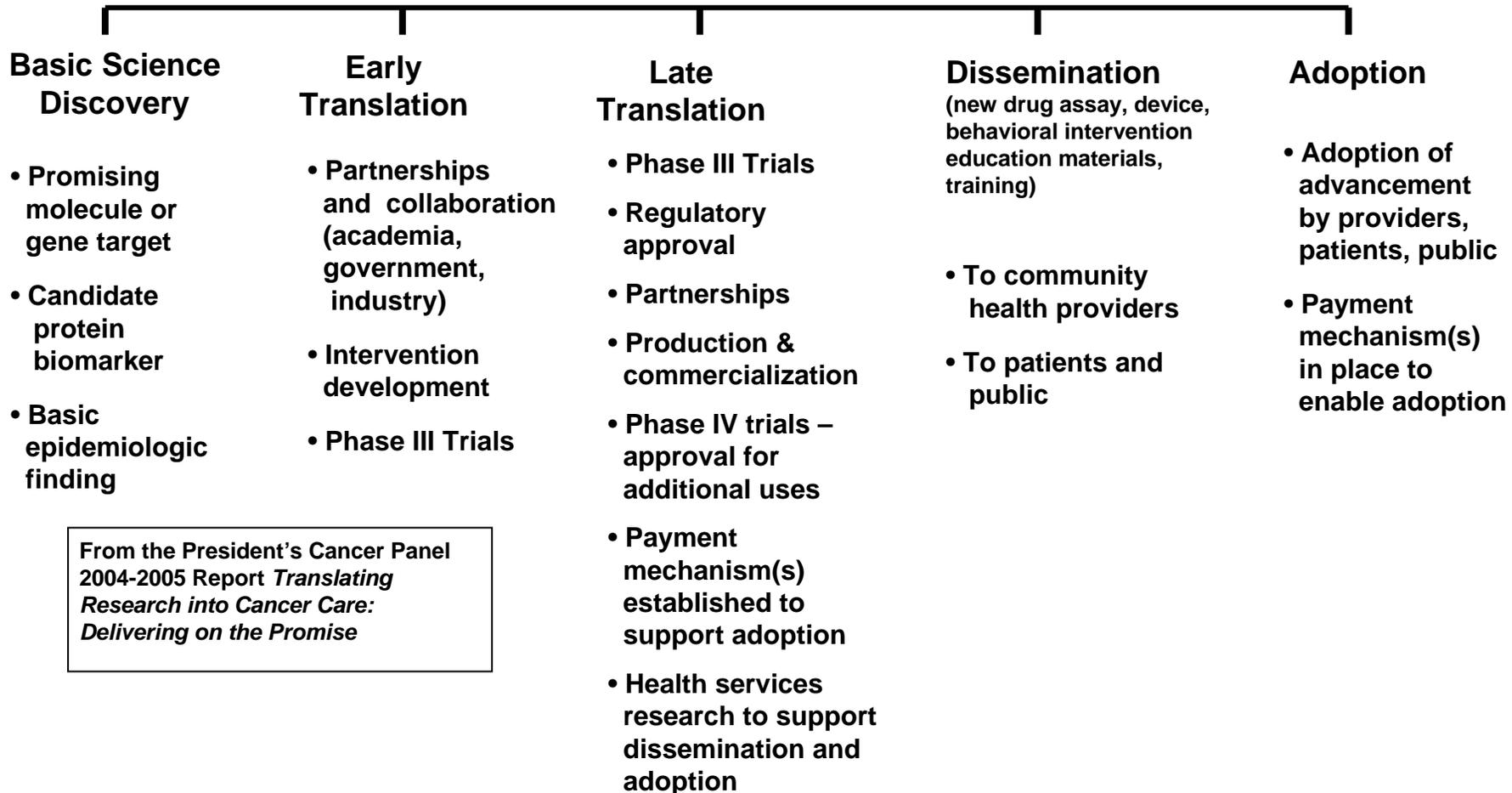
"Translational research transforms scientific discoveries arising from laboratory, clinical, or population studies into clinical applications to reduce cancer incidence, morbidity, and mortality."



<http://www.cancer.gov/trwg/TRWG-definition-and-TR-continuum>



# NCI's Translational Research Continuum



<http://www.cancer.gov/trwg/TRWG-definition-and-TR-continuum>



# Translational Research Applied to Public Health Informatics

**“Public Health Informatics** has been defined as the systematic application of information and computer science and technology to public health practice.”

Yasnoff WA, O'Carrol PW, Koo D, Linkins RW, Kilbourne E. Public health informatics: Improving and transforming public health in the information age. *J Public Health Management Practice*. 2000; 6(6): 67-75.



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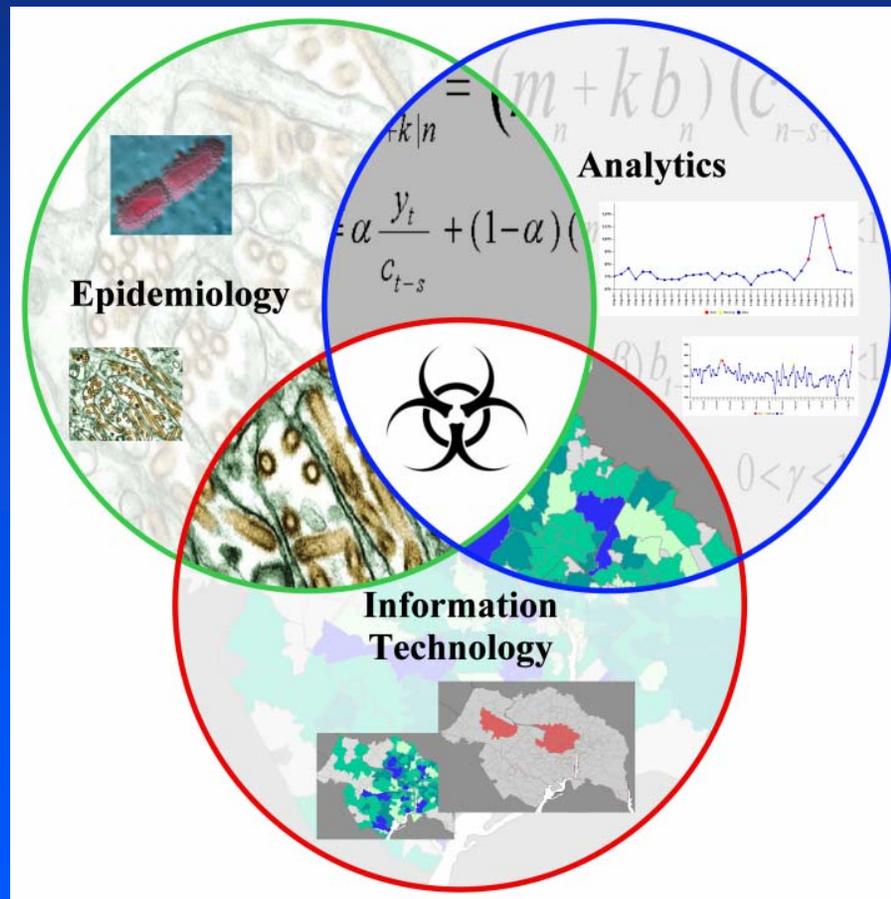
## Proposed Definition for Translational Research for Public Health Informatics:

Translational research in public health informatics is the conversion of advancements made in information and computer science into tools and applications to **support public health practice**.



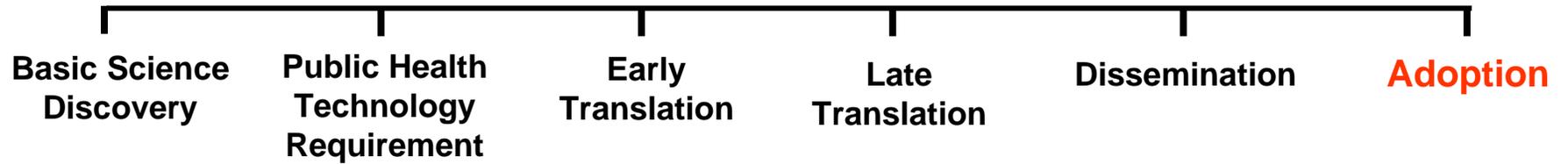
# Alternative Definition of Translation Research For Public Health Informatics

Translational research in public health informatics is the translation of advancements made at the intersections of information technology, mathematics, and epidemiology into tools and applications to support public health practice.



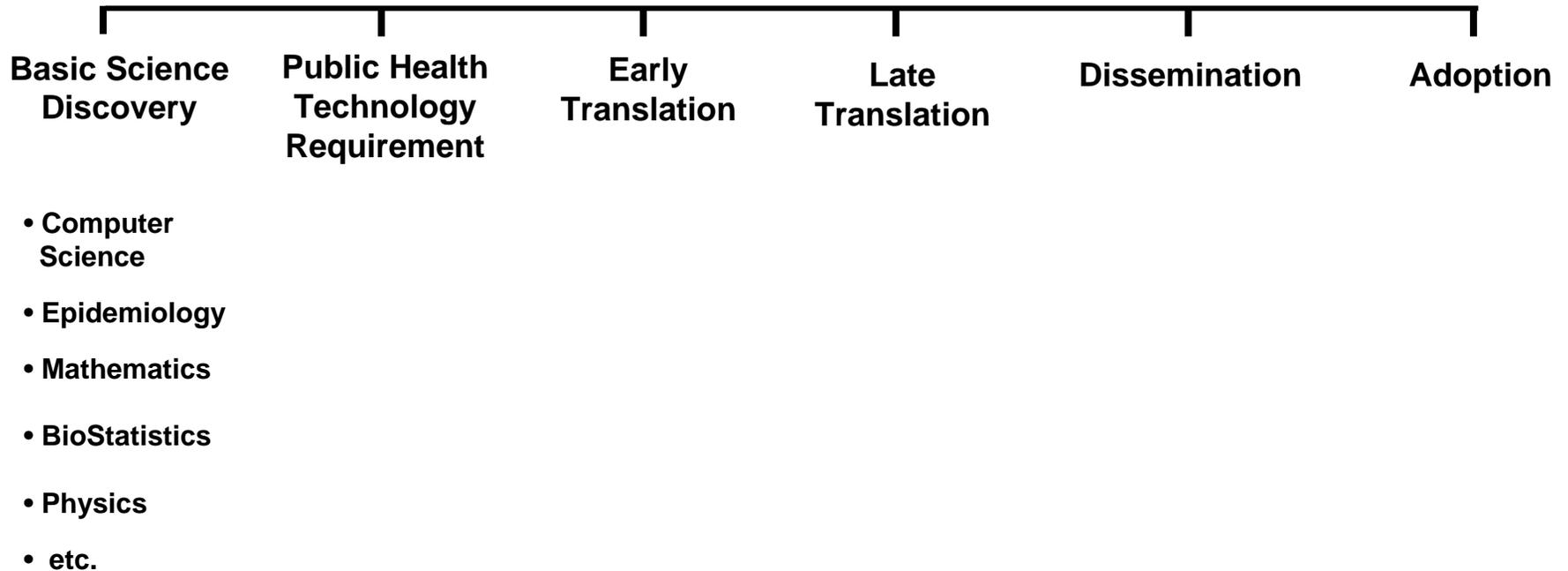


# A (proposed) Public Health Informatics Translational Research Continuum



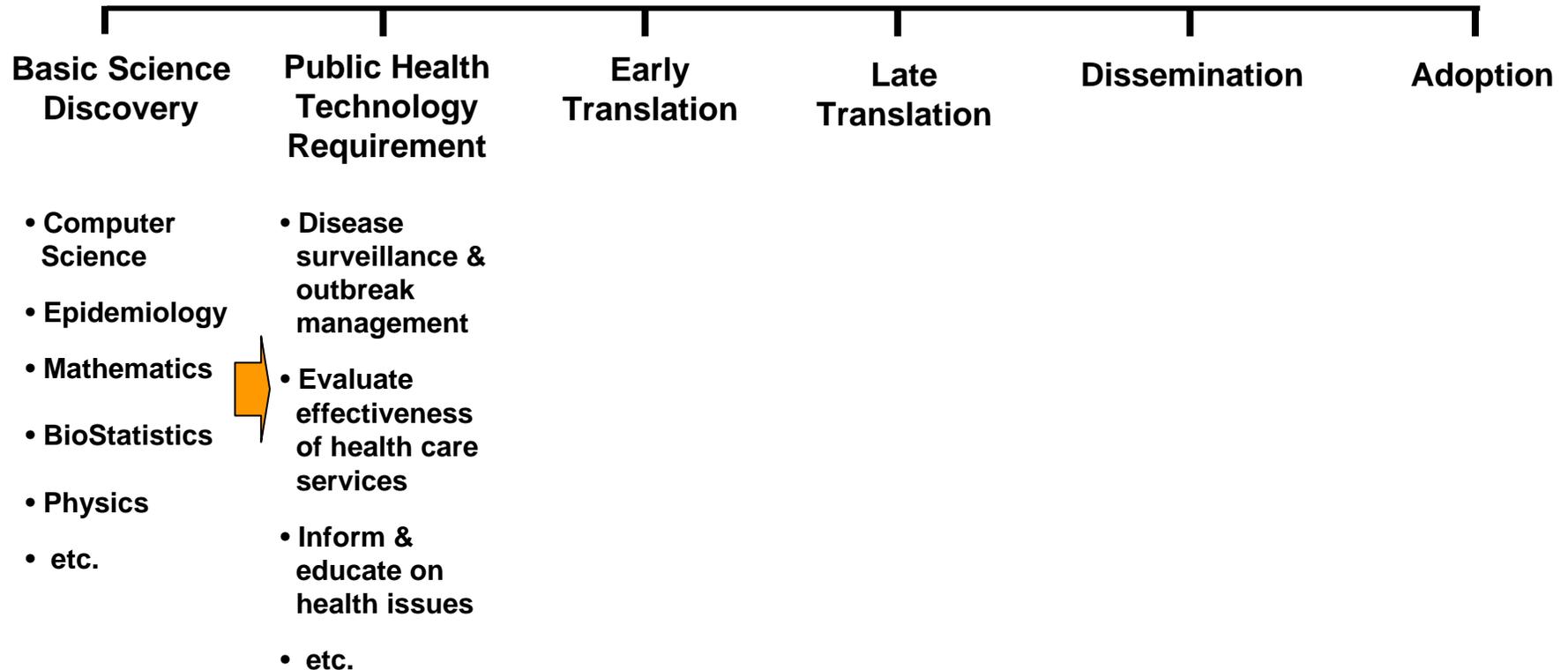


# A Public Health Informatics Translational Research Continuum



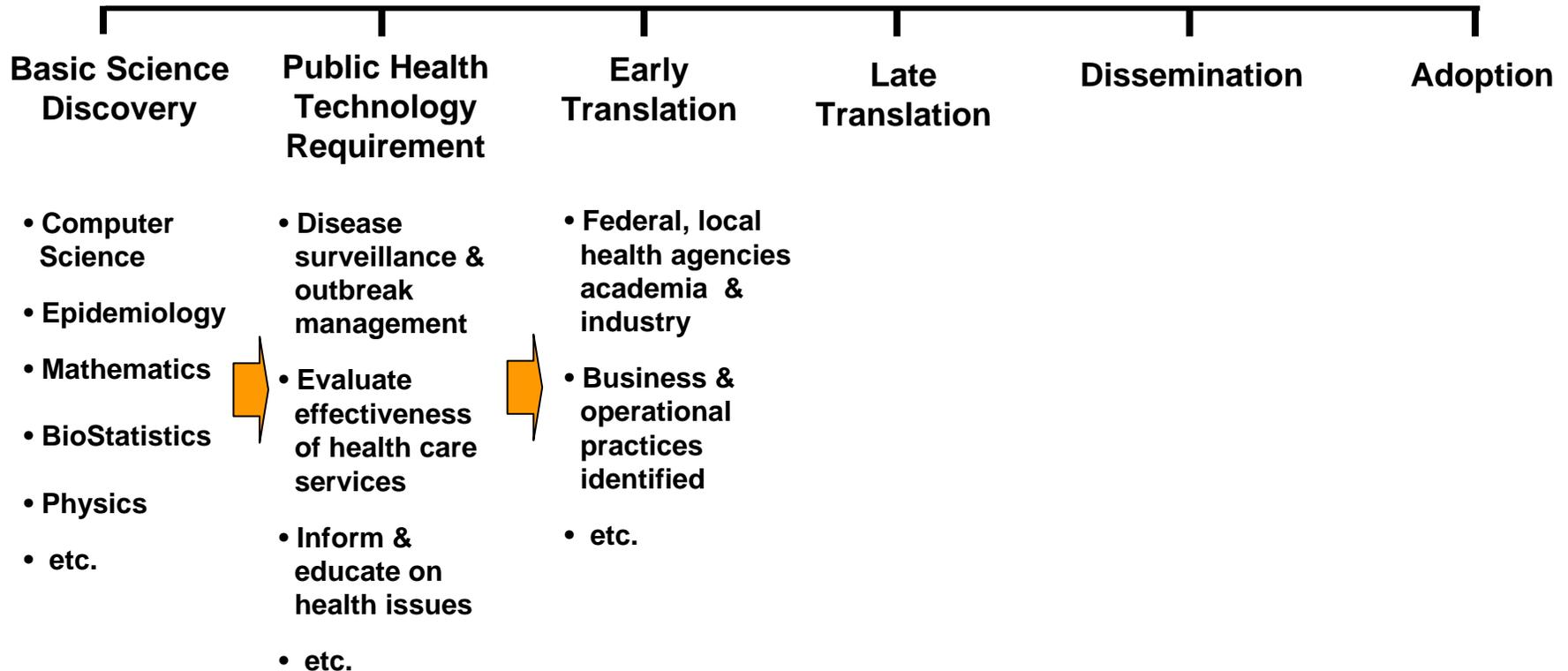


# A Public Health Informatics Translational Research Continuum



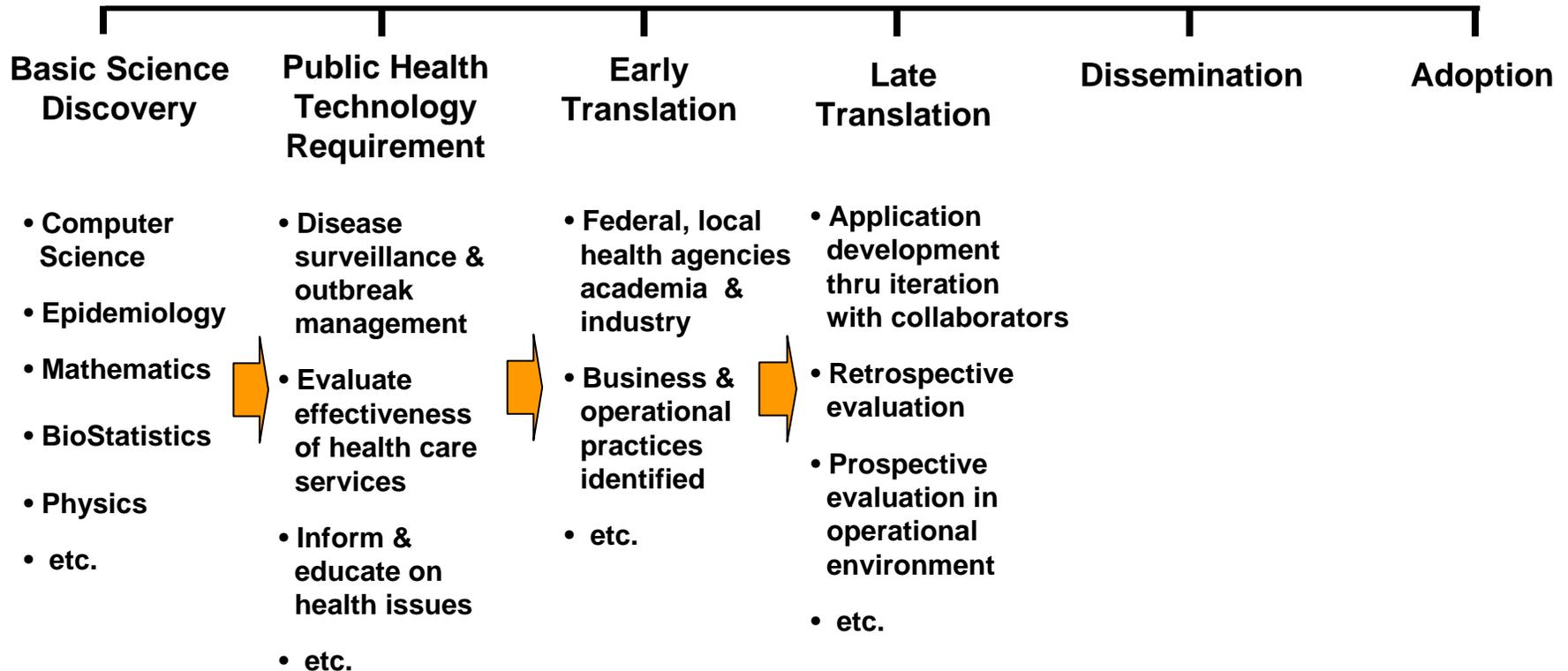


# A Public Health Informatics Translational Research Continuum



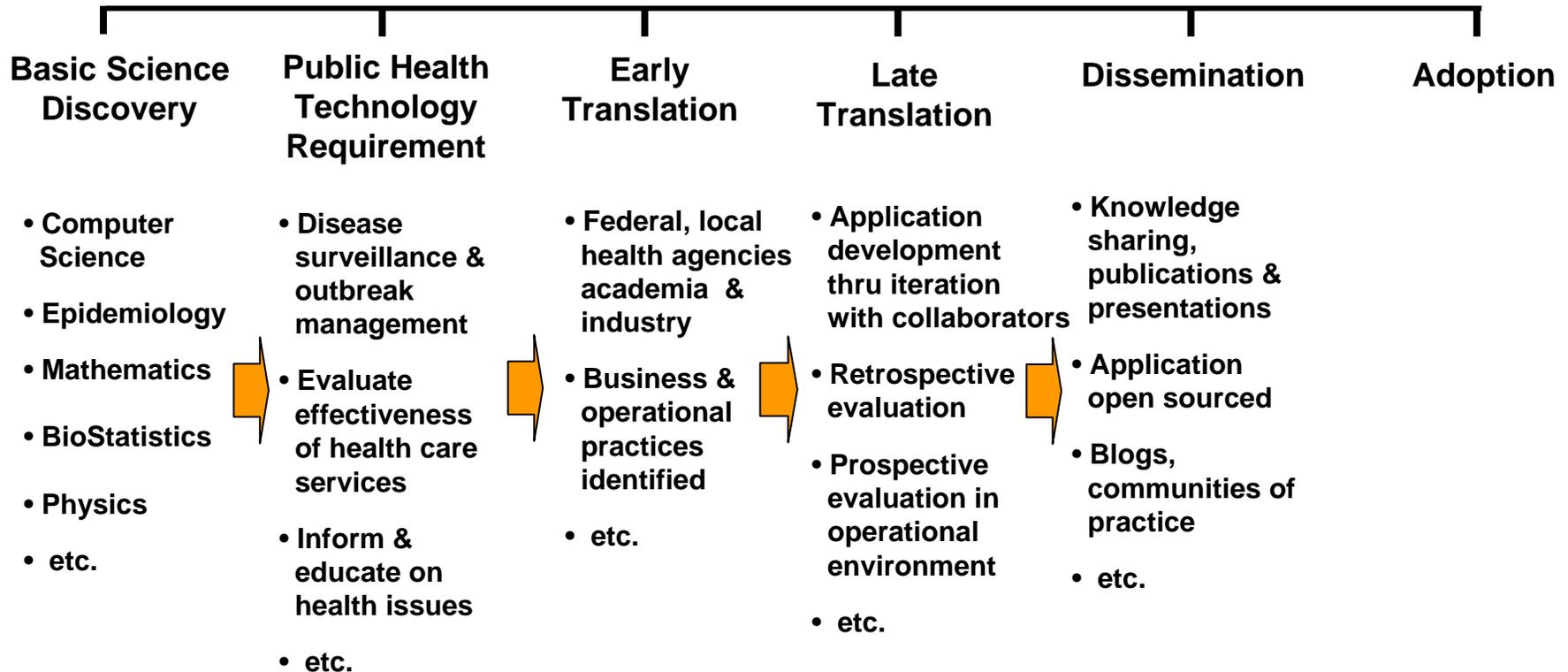


# A Public Health Informatics Translational Research Continuum



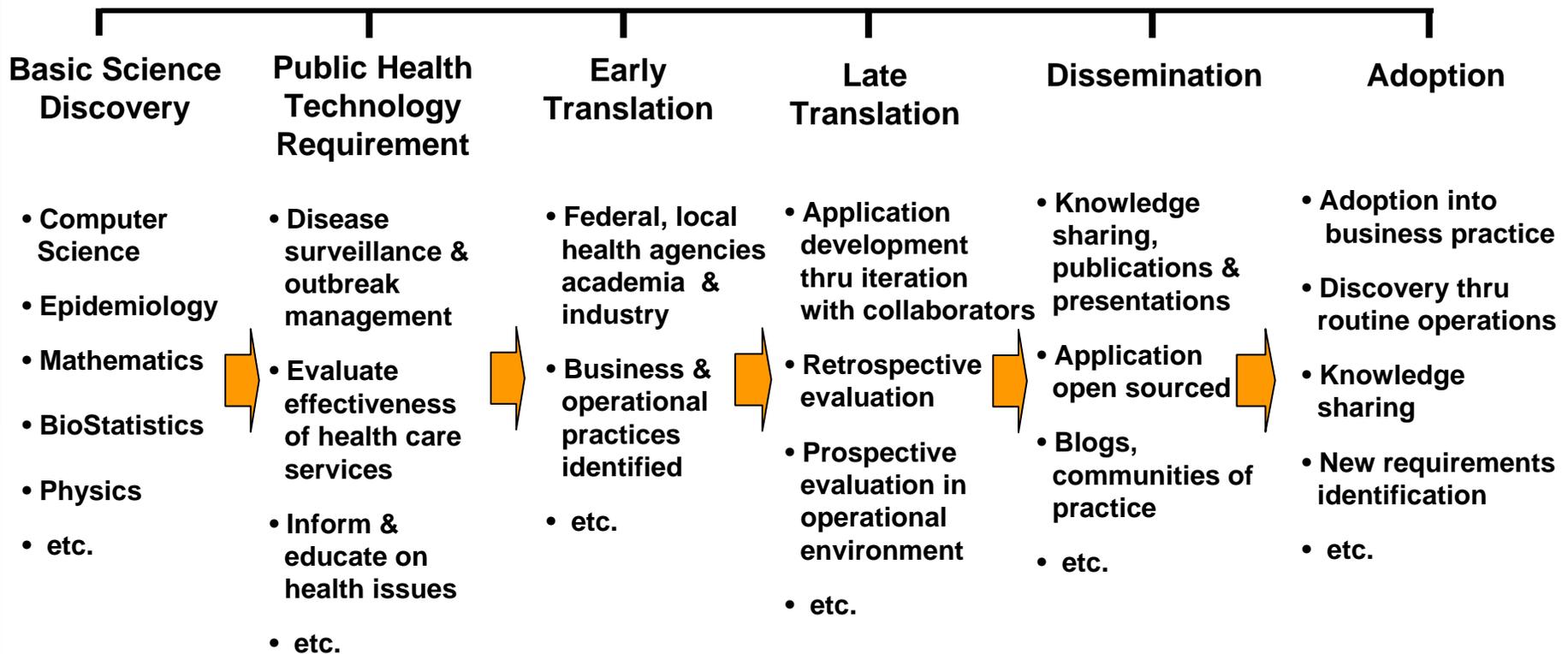


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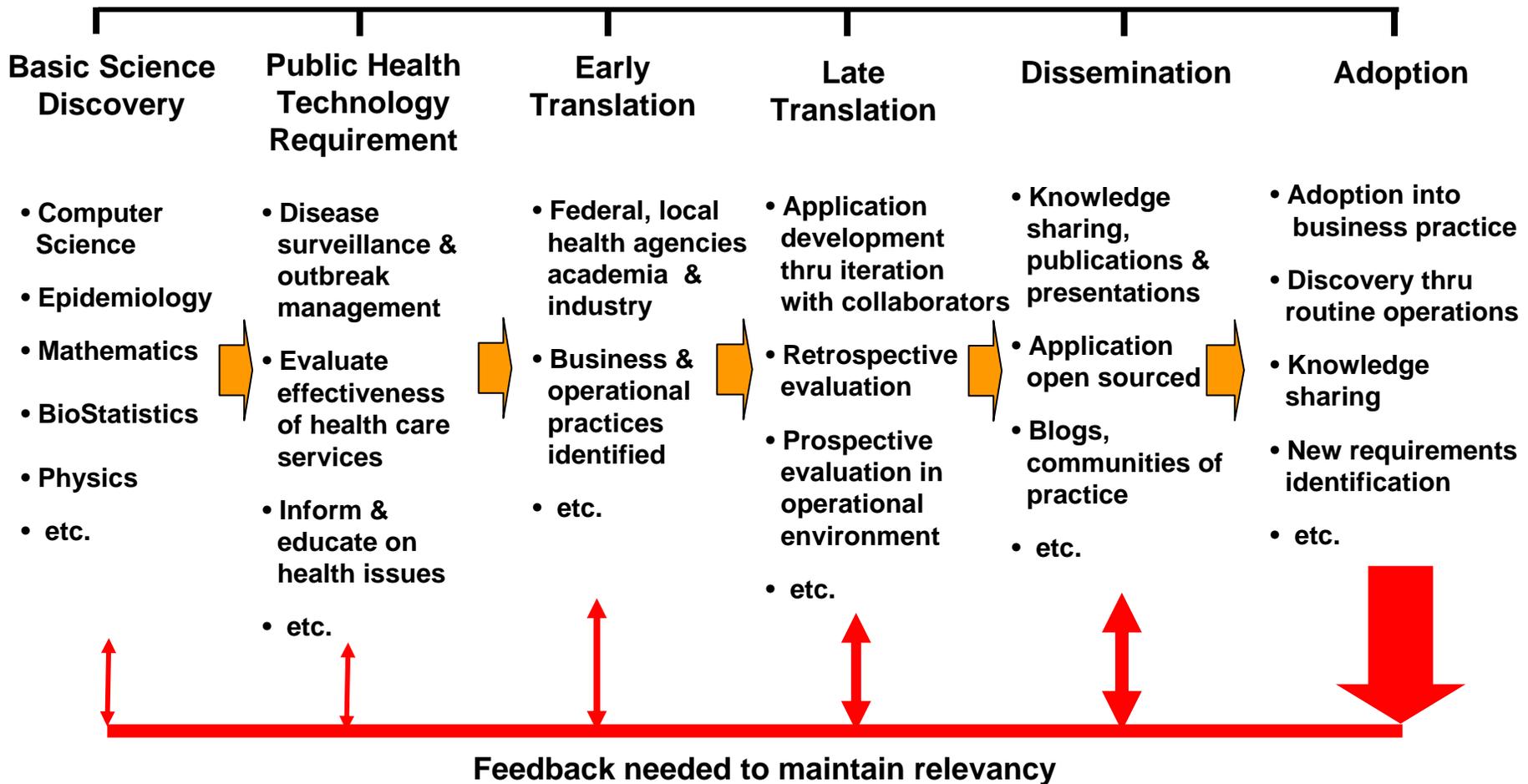


# A Public Health Informatics Translational Research Continuum





# A Public Health Informatics Translational Research Continuum



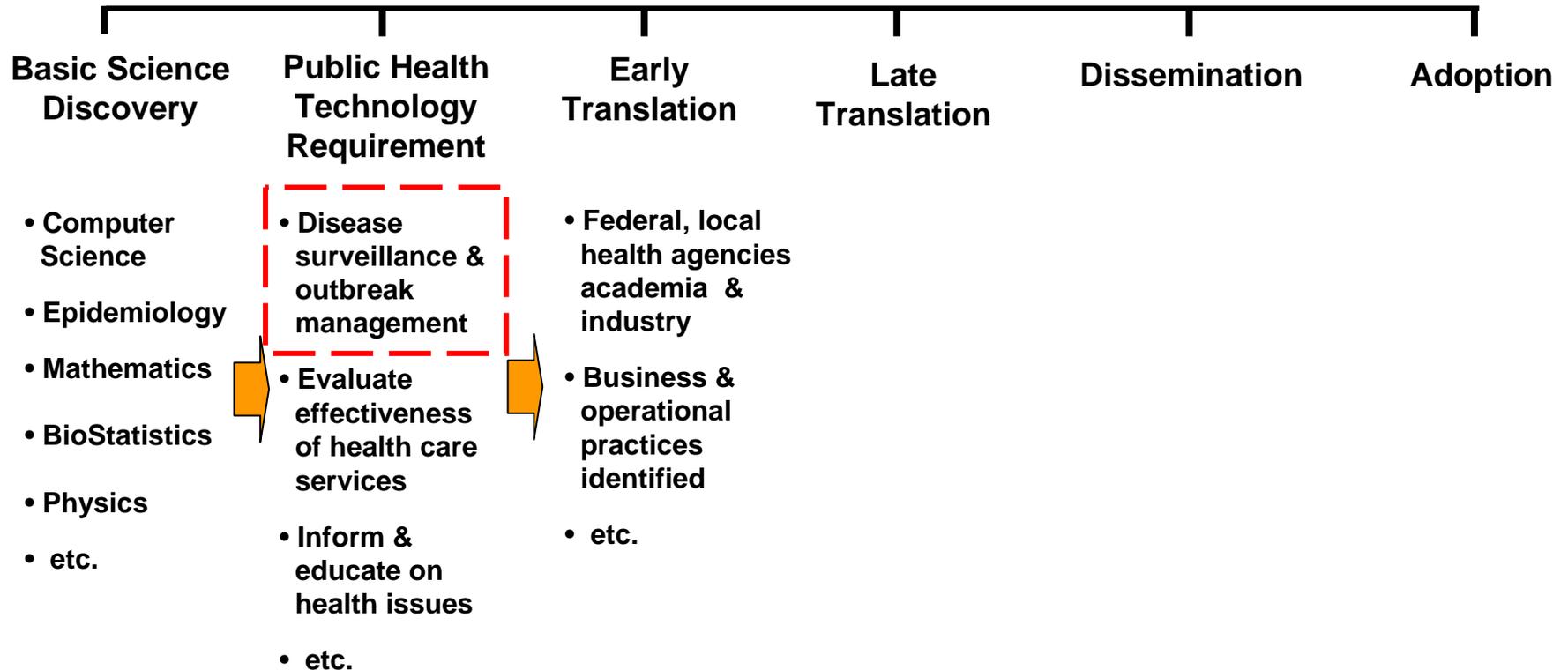


# JHU/APL COE Translational Research in Disease Surveillance

1. Background of the Surveillance Informatics Program at JHU/APL
2. Some Basis for Additional Surveillance Research & Development
3. Center of Excellence Sample Projects



# Surveillance Project Beginnings at JHU/APL





# Identification of a Requirements for a Public Health Informatics Solution for Automating Disease Surveillance

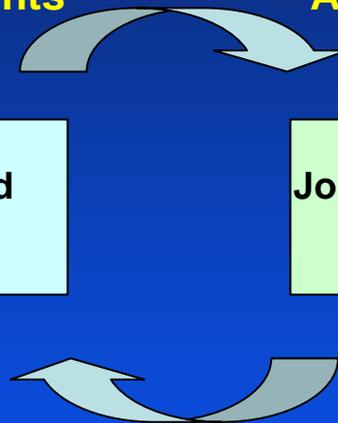
1997–1998

Surveillance  
Requirements

Technical  
Approaches

Maryland  
DHMH

Johns Hopkins  
APL





# Event Driven Requirement for an Operational Prototype

1997–1998–1999

Surveillance  
Requirements

Technical  
Approaches

Maryland  
DHMH

Johns Hopkins  
APL





# Early Indicators Used for Automated Surveillance

1997–1998–1999

Surveillance Requirements

Technical Approaches



Maryland  
DHMH

Johns Hopkins  
APL



**Y2K Surveillance**  
Electronic Billing ICD-9  
Hospital ICP Reports  
Military Data  
Over-the-Counter Meds  
Nursing Homes  
School Absentee Reports



# Y2K Sponsorship

1997–1998–1999

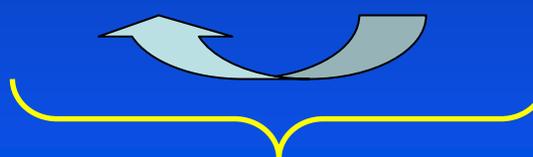
Surveillance  
Requirements

Technical  
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Maryland  
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Johns Hopkins  
APL

DARPA  
Seed  
Funding



**Y2K Surveillance**  
Electronic Billing ICD-9  
Hospital ICP Reports  
Military Data  
Over-the-Counter Meds  
Nursing Homes  
School Absentee Reports



# Expanded Collaborations

1997-1998-1999

Surveillance  
Requirements

Technical  
Approaches



Michael Lewis  
Prev. Med.  
Residency  
ESSENCE  
in the NCR

Maryland  
DHMH

Johns Hopkins  
APL

DARPA  
Seed  
Funding

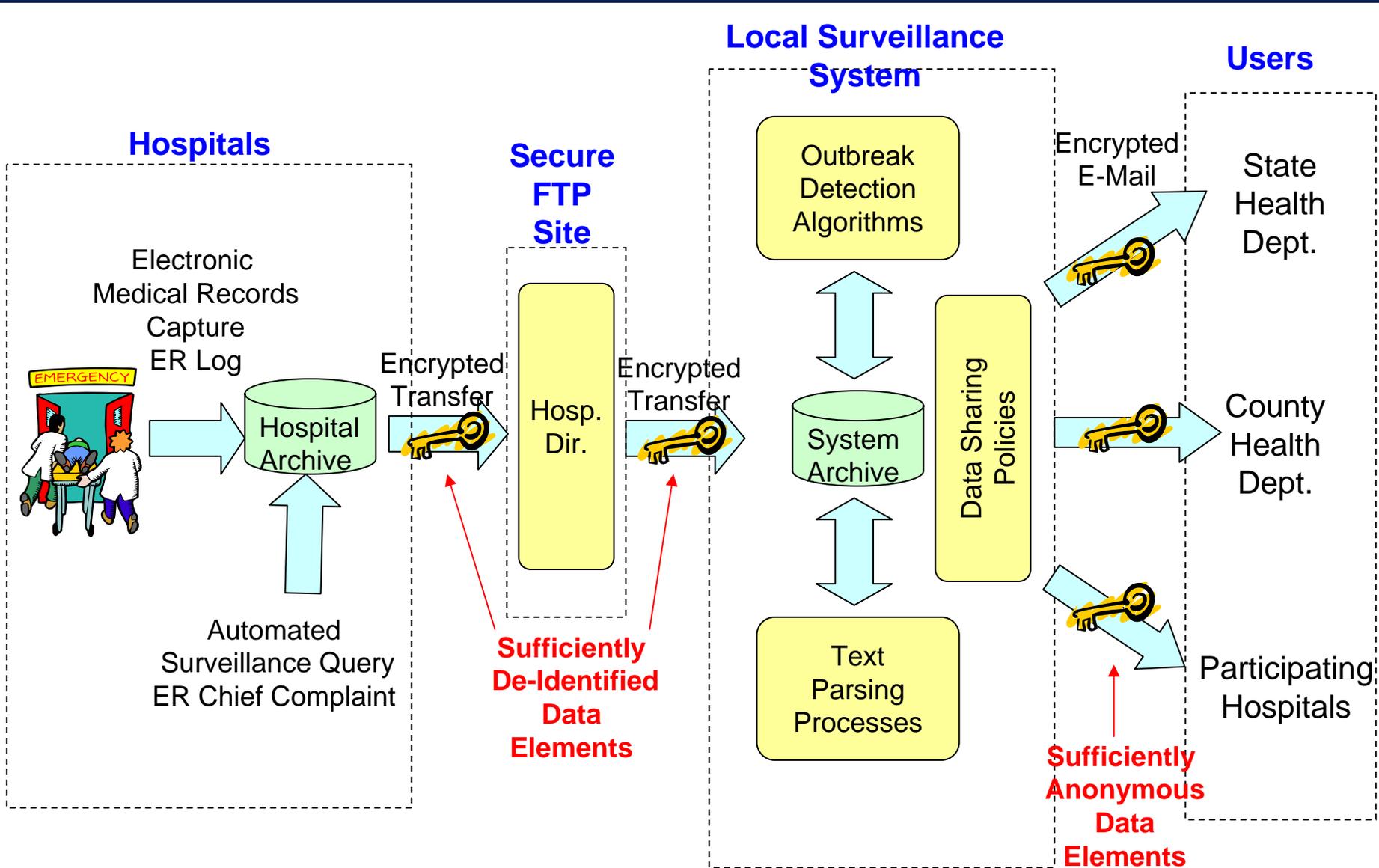


**Y2K Surveillance**  
Electronic Billing ICD-9  
Hospital ICP Reports  
Military Data  
Over-the-Counter Meds  
Nursing Homes  
School Absentee Reports



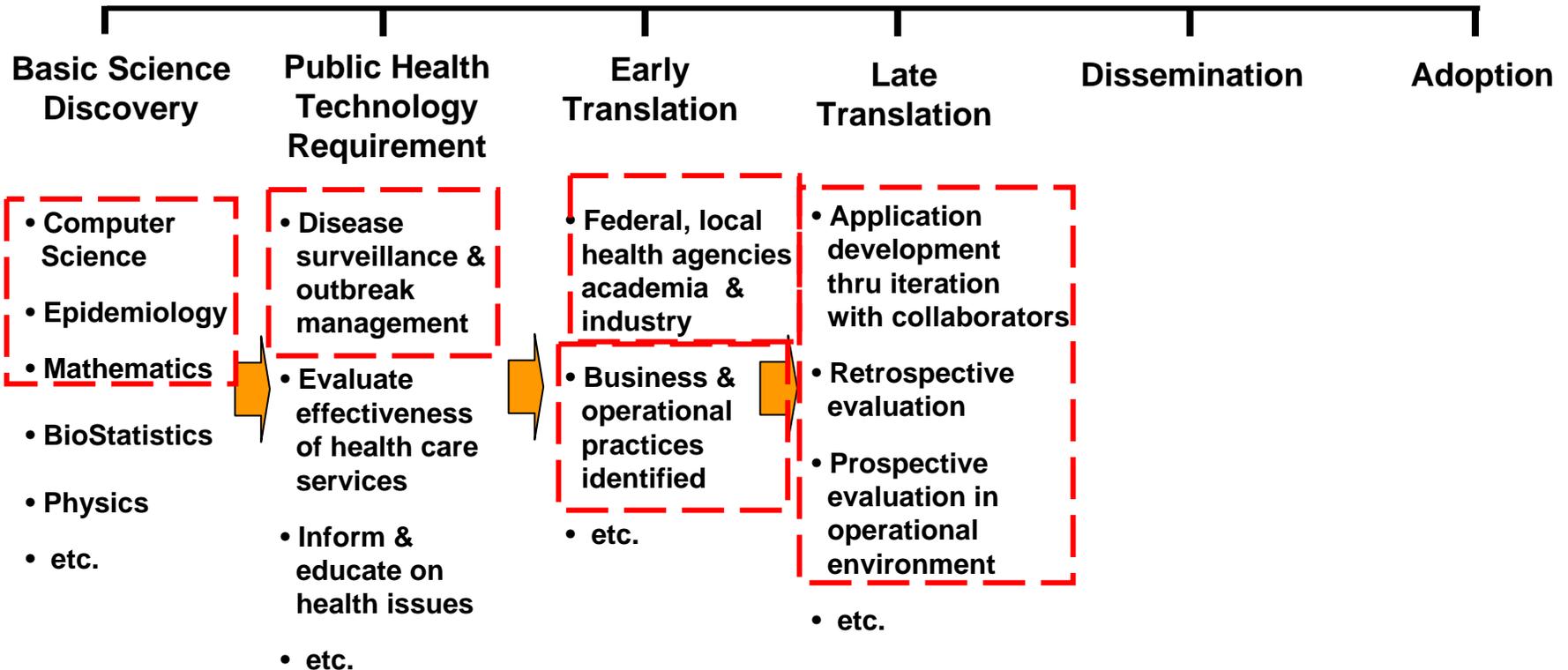
# Early ESSENCE Architecture

Electronic Surveillance System for the Early Notification of Community-based Epidemics





# A Public Health Informatics Translational Research Continuum





# Surveillance Collaborations Provide Adoption & Basis for New Surveillance Functionality

Seedling

BioAlert

Maryland  
DHMH

Maryland  
DHMH

Virginia  
DOH

DC  
DOH

DoD  
GEIS/WRAIR

Moving Across the Continuum  
Through Operational Experiences  
During 9/11 and the Anthrax Letters

Y2K

Regional  
Surveillance





# Surveillance Collaborations Provide Adoption & Basis for New Surveillance Functionality

Seedling

BioAirt

JSIPP

Camp Lejeune, NC

Pope AFB, NC

Barksdale AFB, LA

Ft. Campbell, KY

Ft. Gordon, GA

San Diego, CA

Ft. Lewis, WA

Dahlgren, VA

Robins AFB, GA

Maryland  
DHMH

Virginia  
DOH

DC  
DOH

DoD  
GEIS/WRAIR

Maryland  
DHMH

Y2K

Regional  
Surveillance

Regional Collaborations

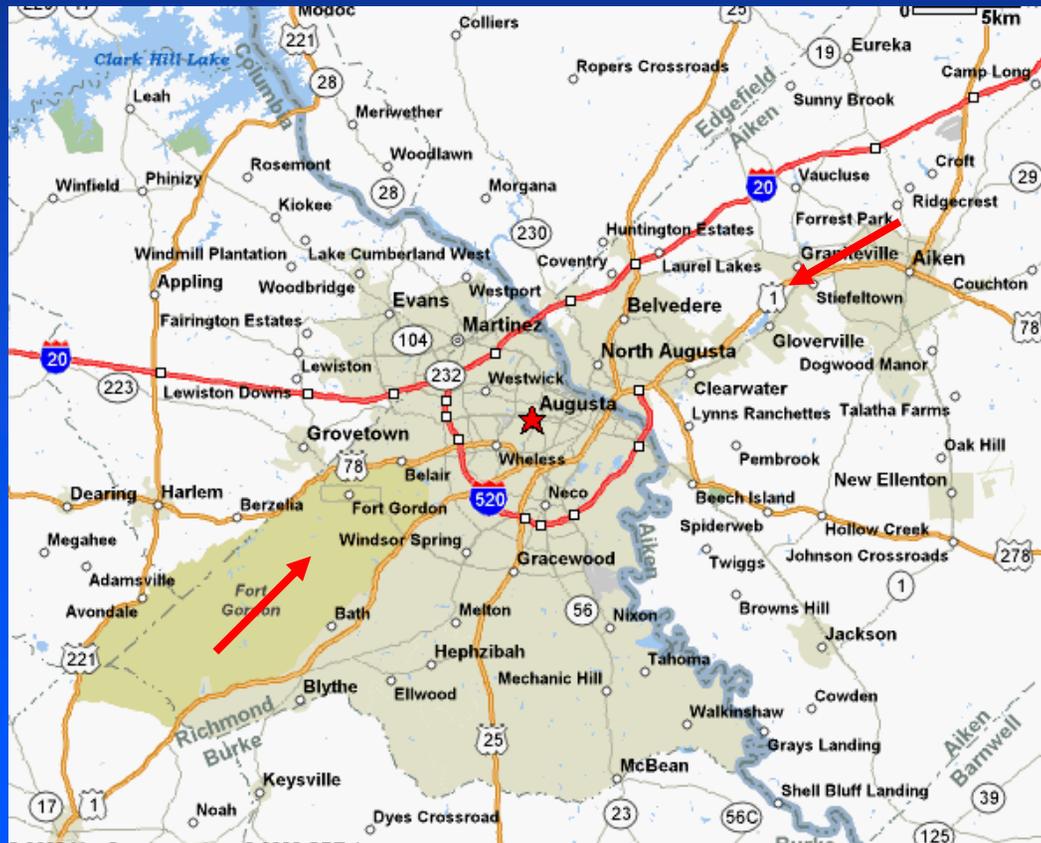
Time

Operational  
Experiences for the  
Continuum



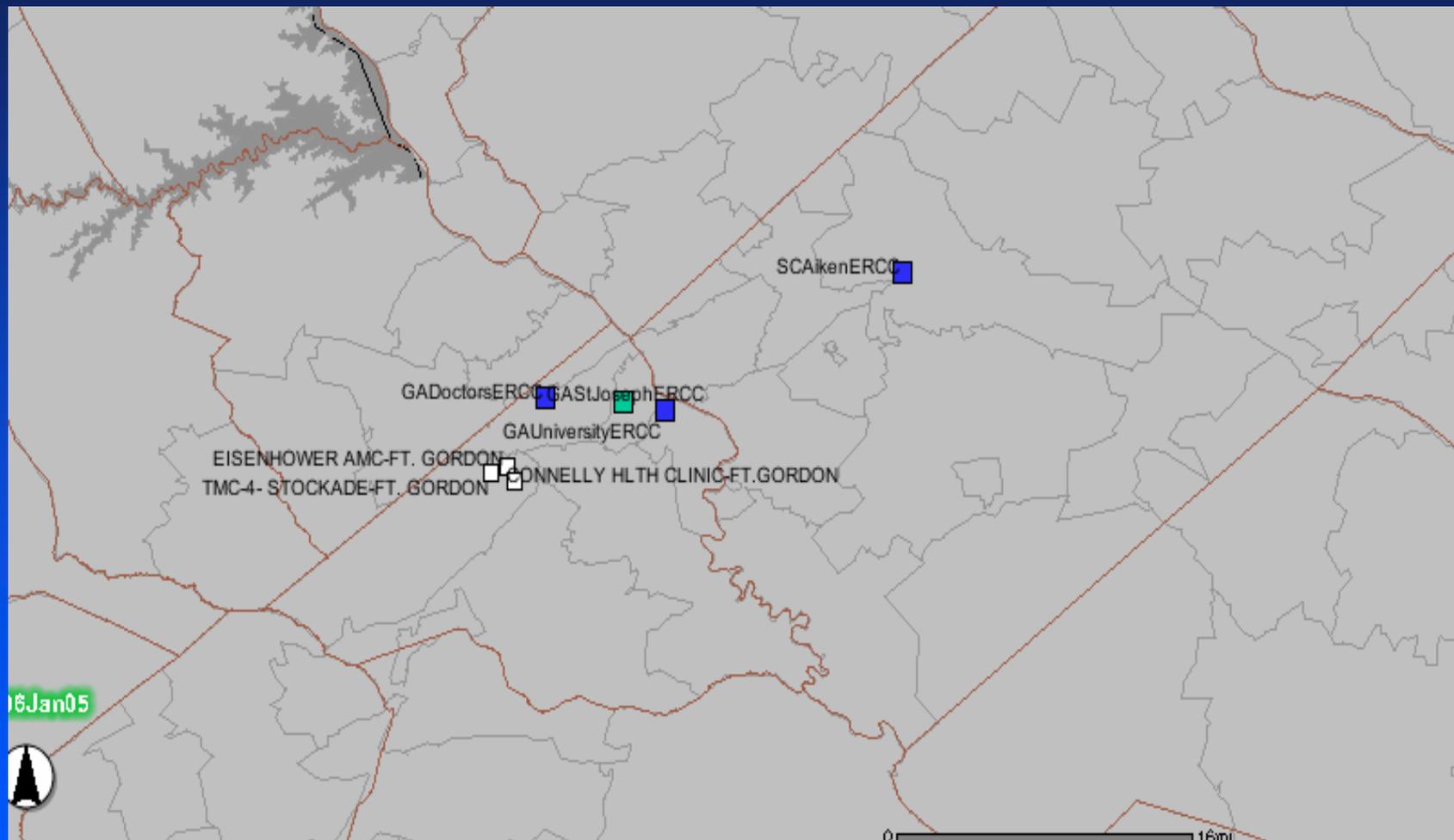
# Train Accident Near Ft. Gordon

On January 6, 2005, two freight trains collided in Graniteville, South Carolina (approximately 10 miles northeast of Augusta, Georgia), releasing an estimated 11,500 gallons of chlorine gas, which caused nine deaths and sent at least 529 persons seeking medical treatment for possible chlorine exposure.



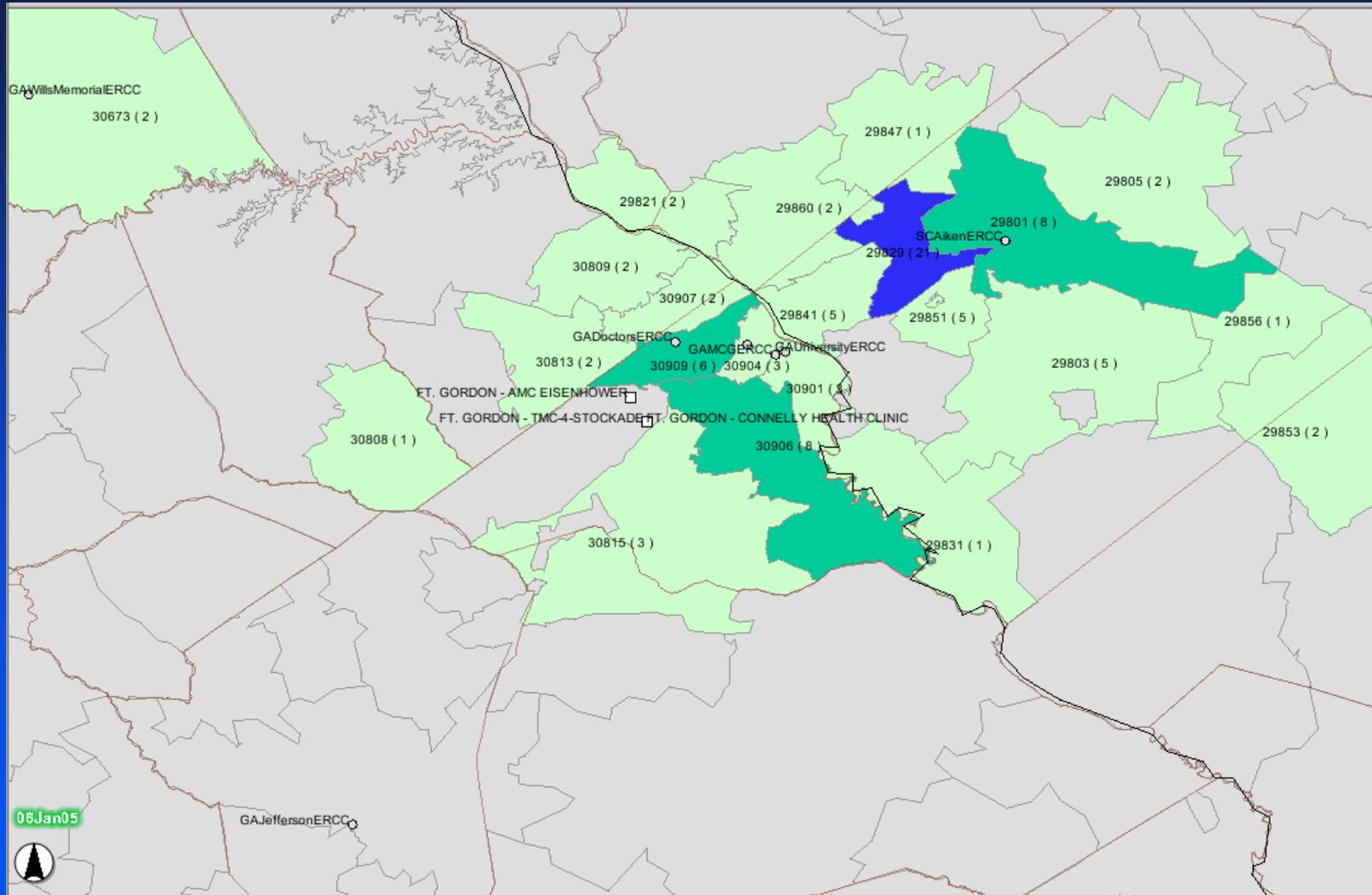


# Location of Medical Facilities Collecting Data for JSIPP During the Accident





# Residence of Patients Seen at Local Hospitals In the Respiratory Syndrome



Learning Opportunities and Feedback into the Continuum



# Surveillance Collaborations Provide Adoption & Basis for New Surveillance Functionality

**Seedling**

**BioAlert**

**JSIPP**

**BioWatch**

Maryland  
DHMH

Maryland  
DHMH

Virginia  
DOH

DC  
DOH

DoD  
GEISWRAIR

Camp Lejeune, NC

Pope AFB, NC

Barksdale AFB, LA

Ft. Campbell, KY

Ft. Gordon, GA

San Diego, CA

Ft. Lewis, WA

Dahlgren, VA

Robins AFB, GA

Veterans Health  
Admin.

Washington  
DoH

Santa Clara  
DoH

Missouri  
DoH

Marion Co.  
DoH

LA County  
DoH

Cook Co.  
DoH

Tarrant Co.  
DoH

Miami  
DoH

Milwaukee  
DoH

**Y2K**

**Regional  
Surveillance**

**Regional Collaborations**

**Wealth of Feedback**





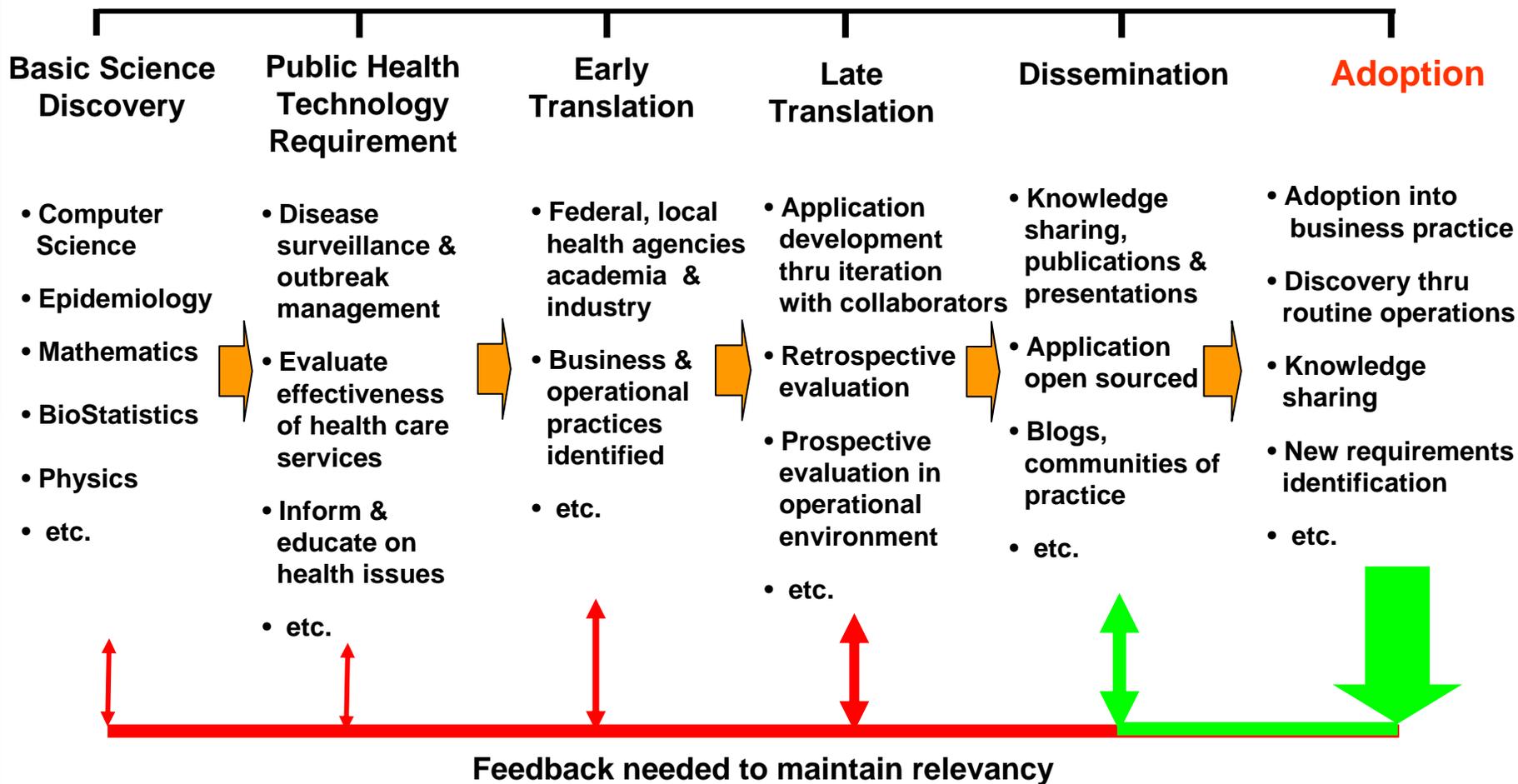
# Feedback into the Continuum from Presentations or Posters on Studies Supported by the ESSENCE Application

**Syndromic Surveillance Conference 2008, December 3-5, 2008**

1. Improvement in Performance of Ngram Classifiers with Frequency Updates, P. Brown et al.
2. Evaluation of Body Temperature to Classify Influenza-Like Illness (ILI) in a Syndromic Surveillance System, M. Atherton, et al.
3. Comparison of influenza-like Illness Syndrome Classification Between Two Syndromic Surveillance Systems, T. Azarian, et al.
4. Monitoring *Staphylococcus* Infection Trends with Biosurveillance Data, A. Baer, et al.
5. How Bad Is It? Using Biosurveillance Data to Monitor the Severity of Seasonal Flu, A. Baer, et al.
6. Socio-demographic and temporal patterns of Emergency Department patients who do not reside in Miami-Dade County, 2007, R. Borroto, et al.
7. Enhancing Syndromic Surveillance through Cross-border Data Sharing, B. Fowler, et al.
8. Early Identification of *Salmonella* Cases Using Syndromic Surveillance, H. Brown, et al.
9. Support Vector Machines for Syndromic Surveillance, A. Buczak, et al.
10. Evaluation of Alerting Sparse-Data Streams of Population Healthcare-Seeking Data, H. Burkom, et al.
11. Use of Syndromic Surveillance of Emergency Room Chief Complaints for Enhanced Situational Awareness during Wildfires, Florida, 2008, A. Kite-Powell et al.
12. North Texas School Health Surveillance: First-Year Progress and Next Steps, T. Powell, et al.
13. Utilizing Emergency Department Data to Evaluate Primary Care Clinic Hours, J. Lincoln, et al.
14. Application of Nonlinear Data Analysis Methods to Locating Disease Clusters, L. Moniz, et al.
15. Innovative Uses for ESSENCE to Improve Standard Communicable Disease Reporting Practices in Miami-Dade County, E. O'Connell, et al.
16. Substance Abuse Among Youth in Miami-Dade County, 2005-2007, E. O'Connell, et al.
17. ESSENCE Version 2.0: The Department of Defense's World-wide Syndromic Surveillance System Receives Several Enhancements, D. Pattie, et al.
18. Framework for the Development of Response Protocols for Public Health Syndromic Surveillance Systems, L. Uscher-Pines, et al.
19. A Survey of Usage and Response Protocols of Syndromic Surveillance Systems by State Public Health Departments in the United States, L. Uscher Pines, et al.
20. Amplification of Syndromic Surveillance's Role in Miami-Dade County, G. Zhang, et al.
21. Using ESSENCE to Track a Gastrointestinal Outbreak in a Homeless Shelter in Miami-Dade County, 2008, G. Zhang, et al.



# Feedback into the Continuum Through Operational Experiences





# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

**1) Syndromic groupings create performance constraints**



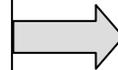
# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Are syndrome groupings the best way to perform surveillance?

#### ICD-9 Based

038.8 Septicemia NEC  
038.9 Septicemia NOS  
066.1 Fever, tick-borne  
066.3 Fever, mosquito-borne NEC  
066.8 Disease, anthrop-borne viral NEC  
066.9 Disease, anthrop-borne viral NOS  
078.2 Sweating fever  
079.89 Infection, viral NEC  
079.99 Infection, viral NOS  
780.31 Convulsions, febrile  
780.6 Fever  
790.7 Bacteremia  
790.8 Viremia NOS  
795.39 NONSP POSITIVE CULT NEC

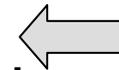


#### Syndrome

Botulism-like  
Febrile Disease  
Fever  
Gastrointestinal  
Hemorrhagic  
Neurological  
Rash  
Respiratory  
Shock / Coma

#### Chief Complaint Based

Chills  
Sepsis  
Body Aches  
Fatigue  
Malaise  
Fever Only





# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

**1) Syndromic groupings create performance limitations**



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Syndromic groupings create performance limitations

- Large groups create a noisy background level



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Syndromic groupings create performance limitations

- Large groups create a noisy background level
- **Signals must be strong enough to be distinguished above the background**



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Syndromic groupings create performance limitations

- Large groups create a noisy background level
- Signals must be strong enough to be distinguished above the background
- **Fixed number of predefined syndromes limit system usefulness for discovery of immediate health risk**



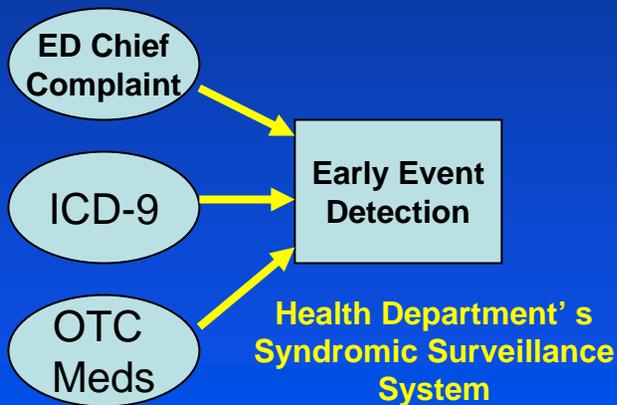
# Changing Environment for Public Health Surveillance & Its impact on Performance

## Existing Surveillance Focus



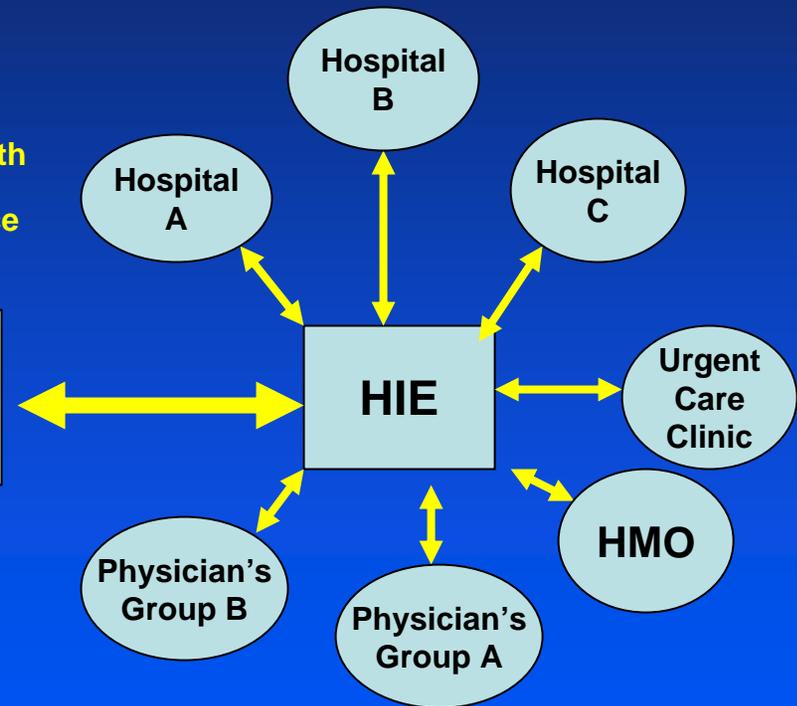
## Health Information Exchanges

Data Containing Health Risk Indications

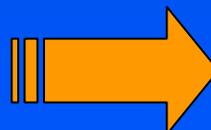


Public Health Agency Surveillance

Public Health Situational Awareness



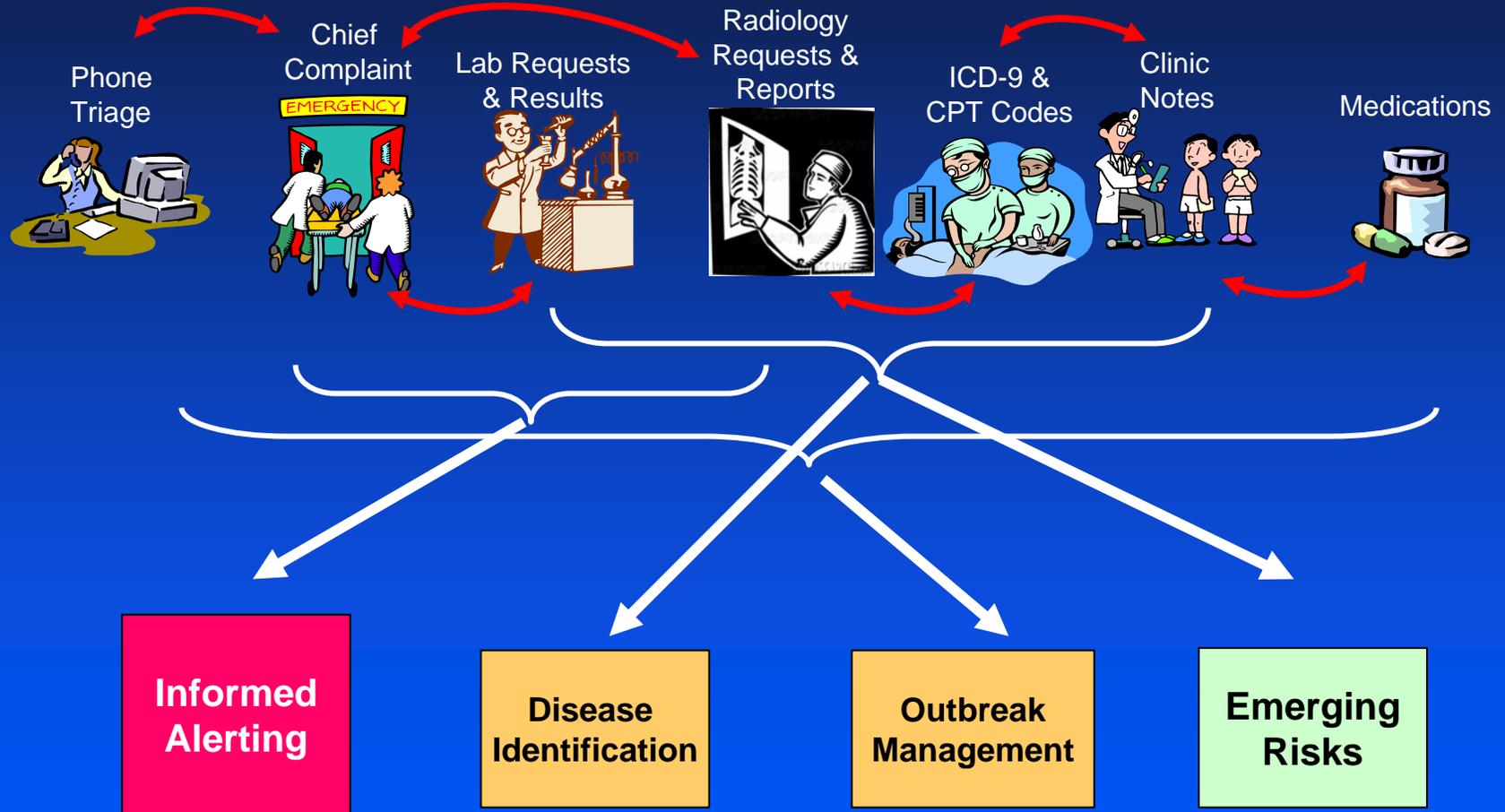
## Non Specific Data Sources



## Electronic Medical Records



# Accessing Linked Medical Records for Public Health Situational Awareness



**Effective use of the Electronic Medical Record Enables Situational Awareness**



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

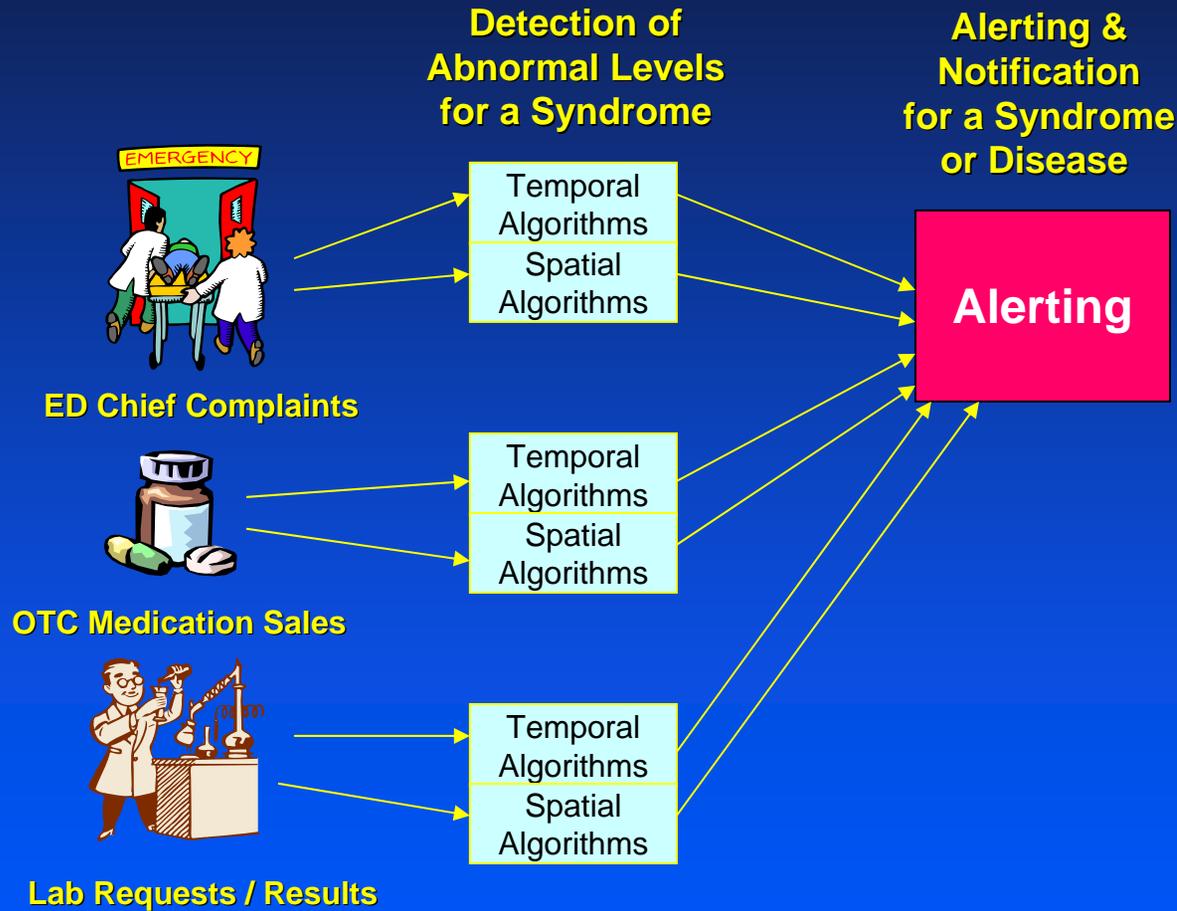
### 1) Syndromic groupings create performance limitations

- Large groups create a noisy background level
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- Fixed number of predefined syndromes limit system usefulness for discovery of immediate health risk

### 2) Effective utilization of multiple data streams



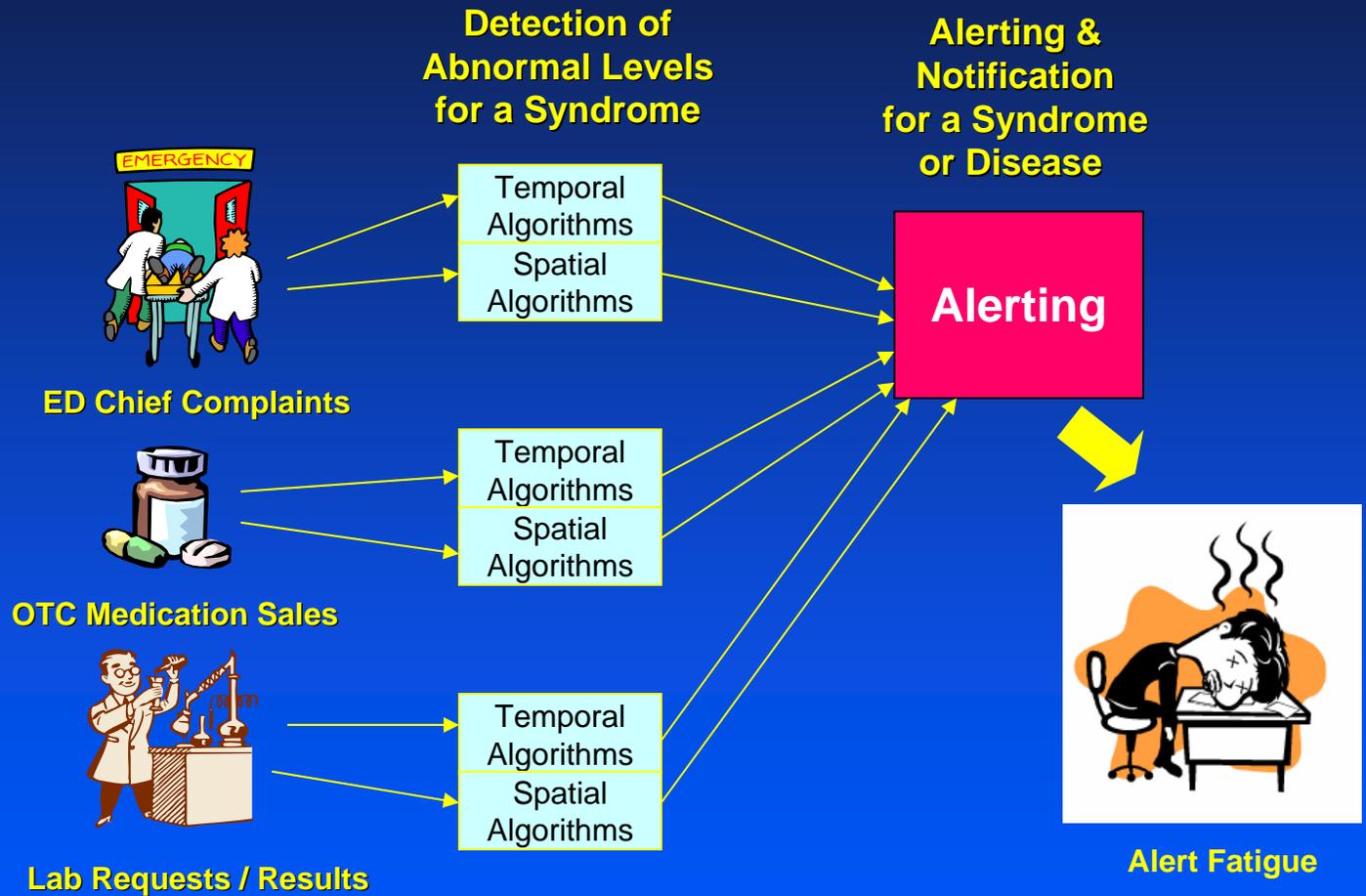
# Current Disease Surveillance Analytics Approach



Adding data sources increases the statistical false positives



# Current Disease Surveillance Analytics Approach



**Adding data sources increases the statistical false positives**



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Syndromic groupings create performance limitations

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### 2) Effective utilization of multiple data streams

- Clinical findings are most relevant on the individual patient level



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

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### 2) Effective utilization of multiple data streams

- Clinical findings are most relevant on the individual patient level
- **Creates additional false positives if the *relationships among the data streams aren't known and included in the algorithms***



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

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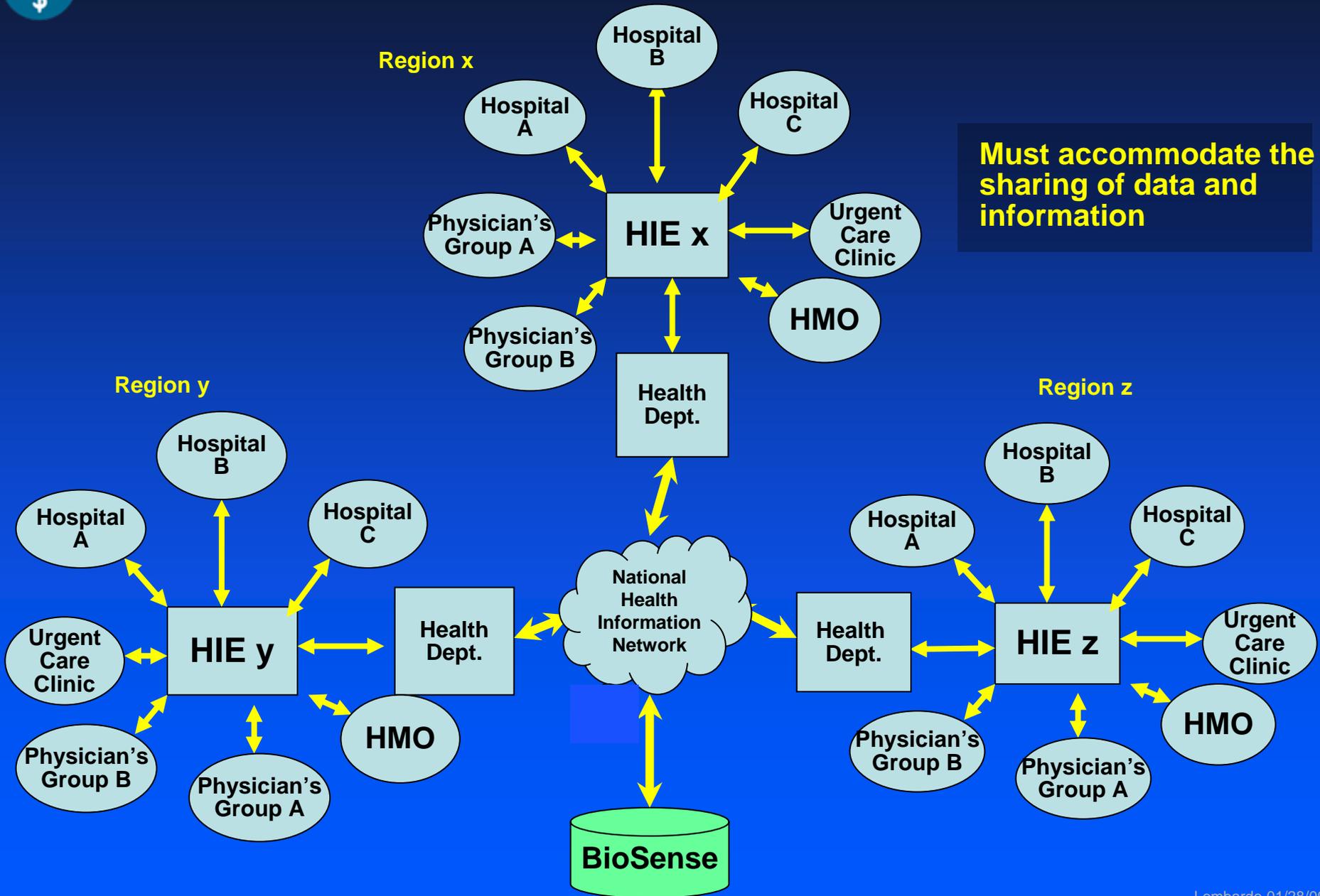
### 2) Effective utilization of multiple data streams

- Clinical findings are most relevant on the individual patient level
- Creates additional false positives if the relationships among the data streams aren't known and included in the algorithms

### 3) Data and information sharing



# National Health Information Sharing





# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

### 1) Syndromic groupings create performance limitations

- Large groups create a noisy background level
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### 2) Effective utilization of multiple data streams

- Clinical findings are most relevant on the individual patient level
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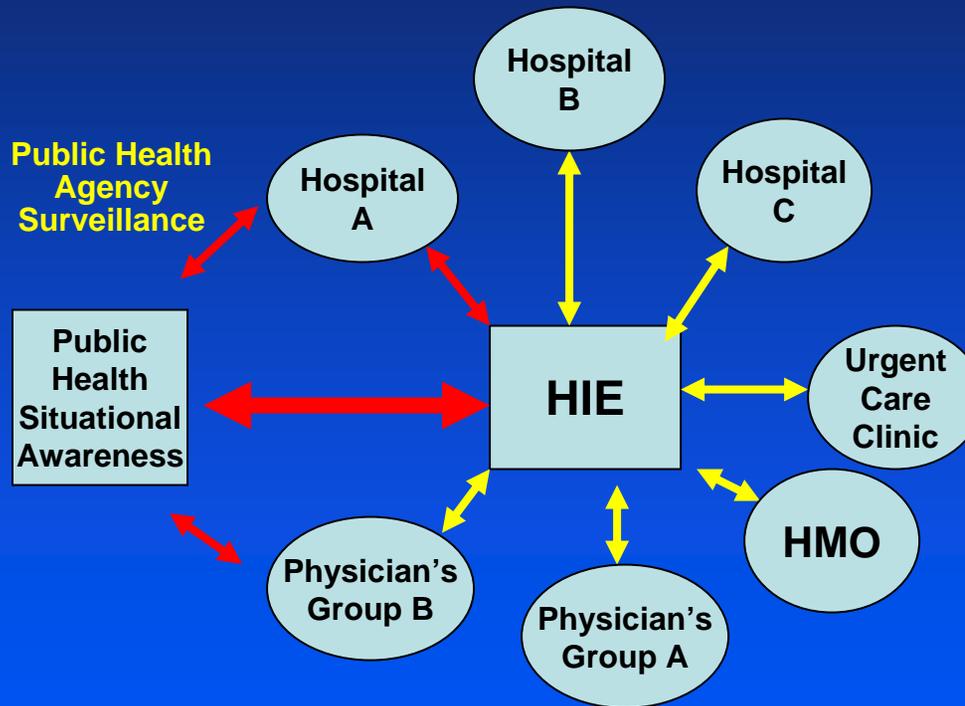
### 3) Data and information sharing

- HIPAA and identity theft have placed limitations on data sharing among public health agencies
- **State laws restrict sending data captured for surveillance purposes outside state boundaries**



# Information Must Be Shared Among Public Health and Health Care Systems

## Health Information Exchanges



## Electronic Medical Records



# Feedback into the Continuum

## Limitations of Existing Syndromic Surveillance

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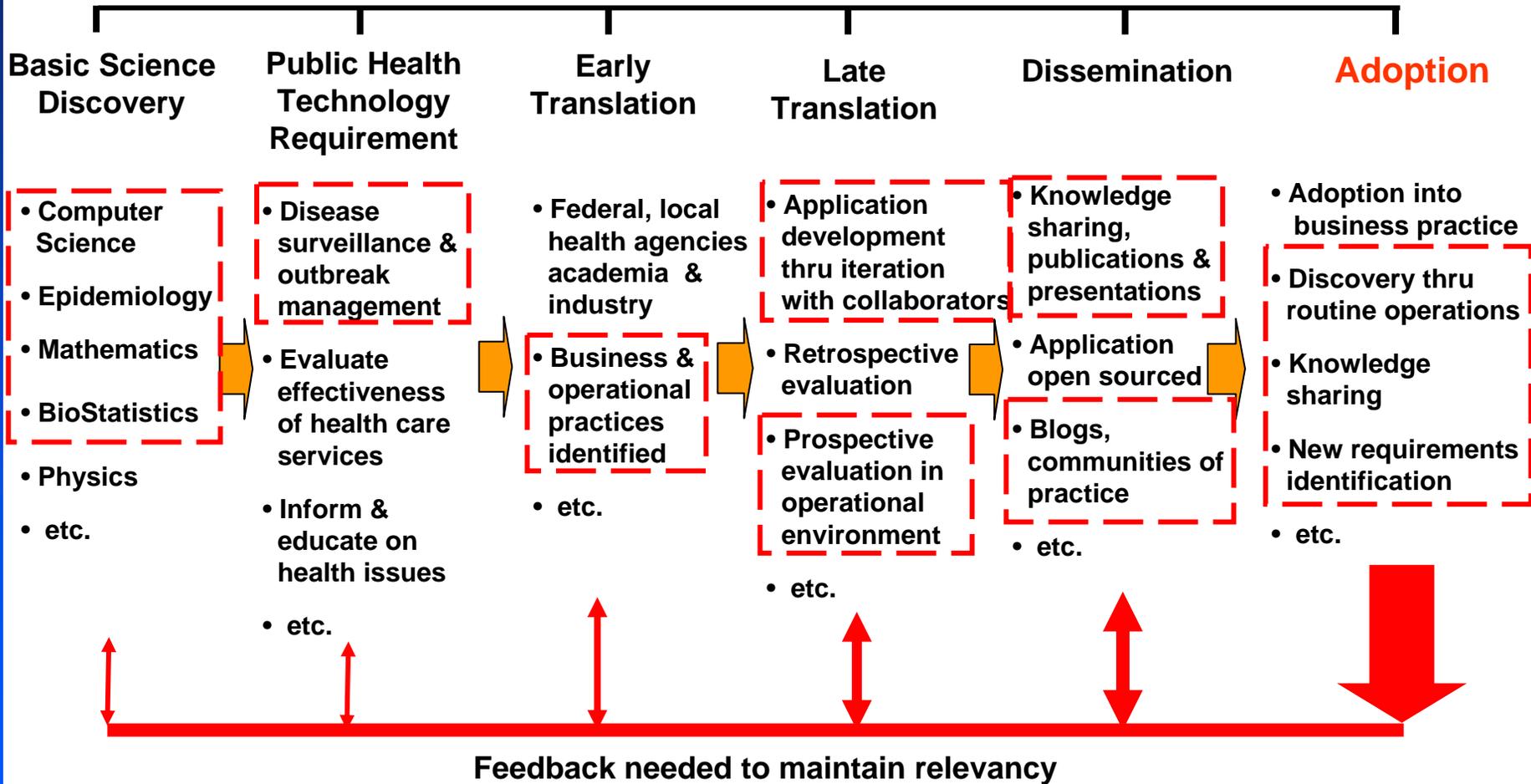
- Clinical findings are most relevant on the individual patient level
- Creates additional false positives if the relationships among the data streams aren't known and included in the algorithms

### 3) Data and information sharing

- HIPAA and identity theft have placed limitations on data sharing among public health agencies
- State laws restrict sending data captured for surveillance purposes outside state boundaries
- Healthcare delivery must be aware of public health concerns
- **Information to support public health surveillance information should be obtained during patient encounters**



# Current Feedback Paths Into the Translational Research Continuum





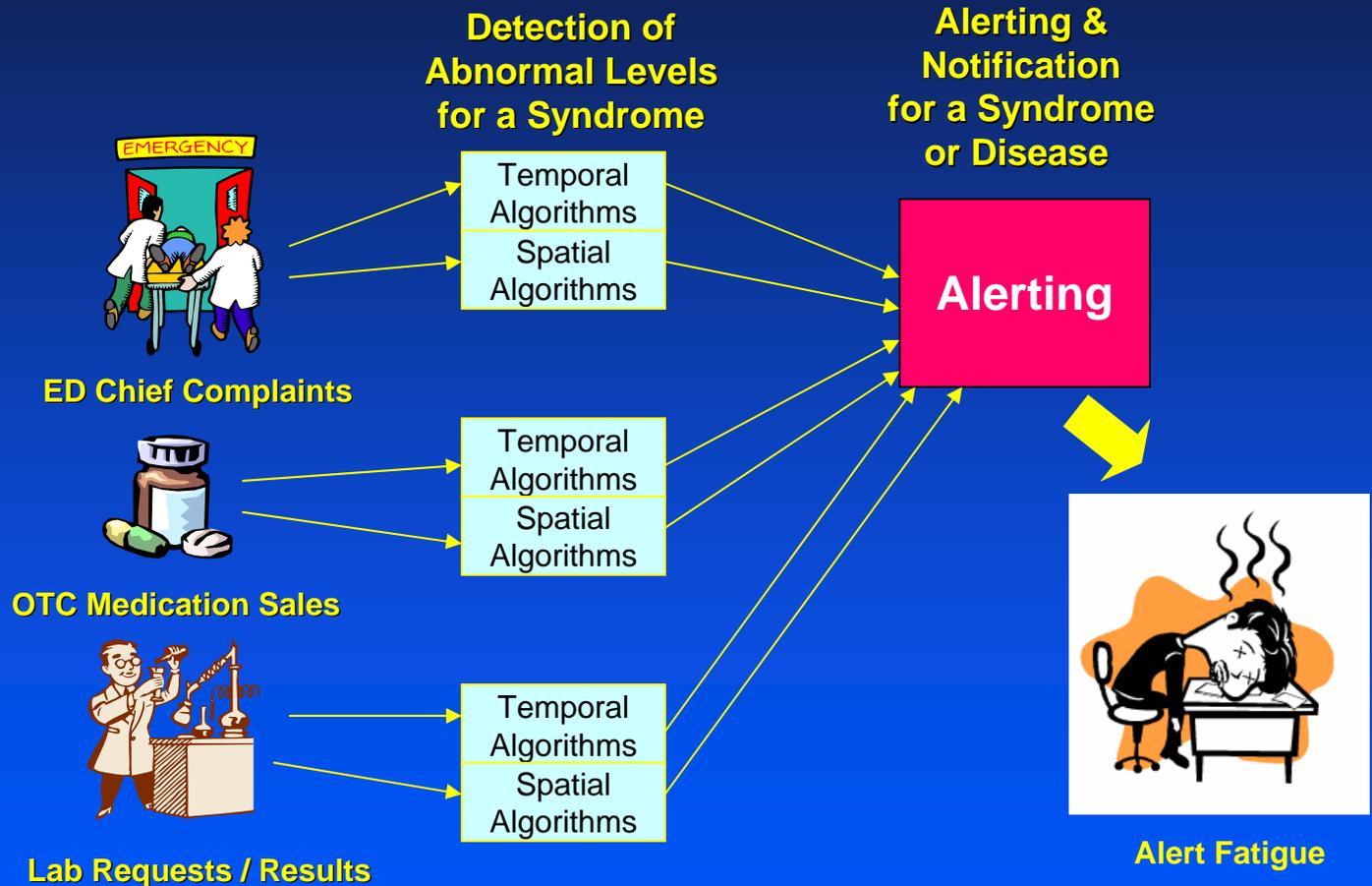
# Requirements Iteration Obtained Through Adoption

## Operationally Identified Surveillance Requirement:

- 1) Ability to perform surveillance for specific populations by performing advanced queries on linked clinical data from medical records
- 2) Ability to create rules to customize the detectors for the specific populations or events being monitored
- 3) Informatics tools are needed that permit epidemiologists and disease monitors to create new surveillance objects without enlisting the support of IT system specialists
- 4) Health risks must be shared between health care and other public health agencies



# Current Disease Surveillance Analytics Approach



**Adding data sources increases the statistical false positives**



# Moving from Syndromic to Case Specific Surveillance in a Collaborative Environment

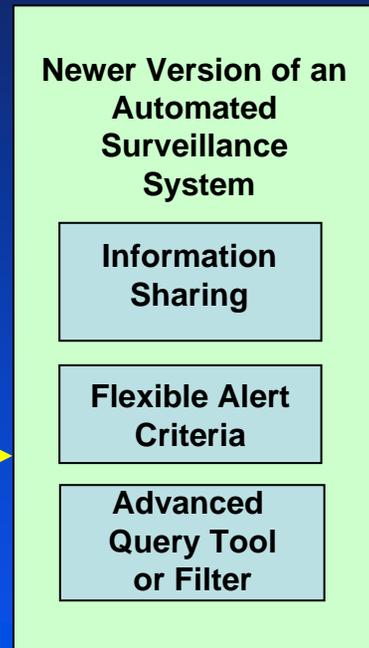
Health Care & Other PH Agencies

Collaborative Sharing of Surveillance Results



Population Specific Analysis

Local Public Health Agency



Epidemiologist Initiated Modifications

Medical Record



Health Information Exchange

Patient Care Delivery

Phone Triage



Lab Requests & Results



Chief Complaint

EMERGENCY



Radiology Requests & Reports



Clinic Notes



ICD-9 & CPT Codes



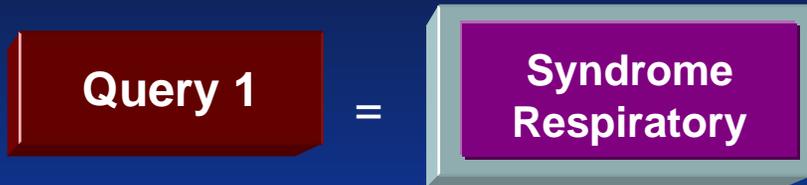
Medications





# Sample Project 1: *Advanced Query Tool, AQT* Analysis of user defined subpopulations

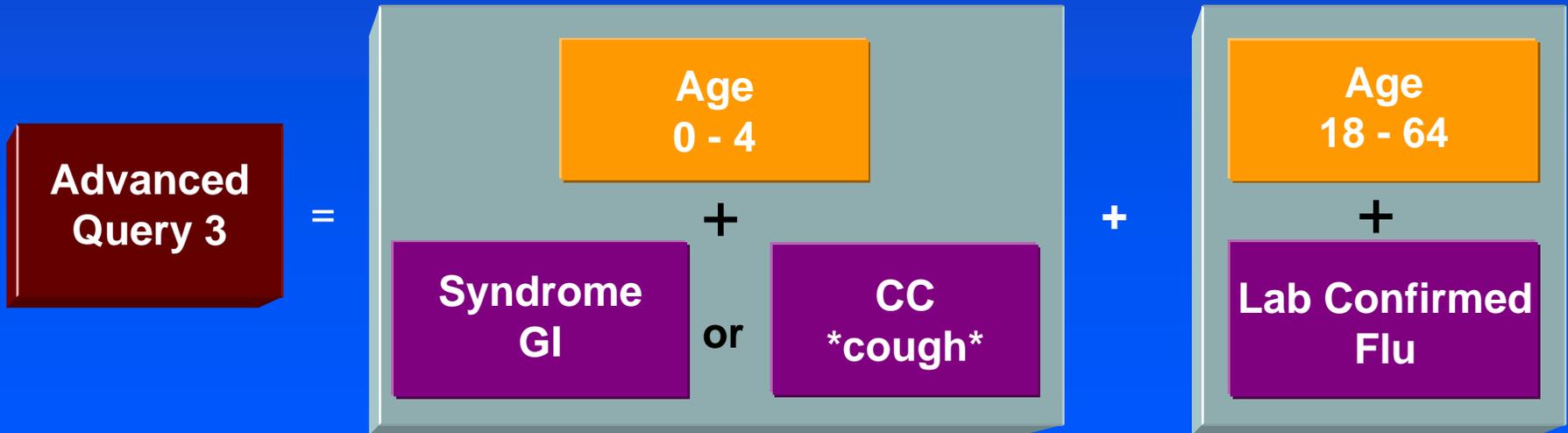
## Query 1: Syndrome based



## Query 2: Chief complaint (CC) free-text



## Query 3: Logical combinations that mix all stratifications (chief complaint, syndrome, etc)





# Advanced Query Tool Project

- Ability to include all the data elements that are available to the surveillance system in the query
- Hide the complexity of the underlying data models and query languages
- Allow users to build on-the-fly case definitions using any data element available.



# Advanced Query Tool

Data Source

## Message Area

## Query

```
([SUBSYNDROME="AbdominalCramps"] OR [SUBSYNDROME="AbdominalPain"])
```

**Example:** [ AGE > "35" ] OR ( [ SUBSYNDROME = "ACUTE BLOOD ABNORMALITIES" ] AND [ ZIPCODE = "21043" ] ) [More...](#)

## Query Builder

- REGION
- ZIPCODE

---

- SYNDROME
- SUBSYNDROME**
- CHIEF-COMPLAINT

---

- AGE
- SEX
- PRIVATE SAVED QUERIES

- =
- <>

---

- Contains
- Starts With
- Ends With

- AbdominalCramps**
- AbdominalPain
- AbdominalPainGroup
- AbdominalTenderness
- Abscess
- AcuteBloodAbnormalities
- AcuteBloodAbnormalitiesGroup
- AlcoholUse
- AllergicReaction
- AllOverBody

Group multiple selections with :  OR  AND



## Further Information on AQT

1. **Advanced Querying Features for Disease Surveillance Systems, M. Hashemian, et al., Spring 2007 AMIA Conf., May 2007.**
2. **Advanced Querying Features for Disease Surveillance Systems, M. Hashemian, et al., 2007 PHIN Conf., Aug. 2007.**
3. **Advanced Querying Features for Disease Surveillance Systems, M. Hashemian, *Advances in Disease Surveillance 2007*;4:97  
Available at: <http://www.isdsjournal.org/article/view/1993/1547>**



## Sample Project 2: My Alerts

- The ability for a user to generate on-the-fly case definitions lead to the need for those dynamic queries to become part of the health department's day-to-day detection system.
- In addition, because these very specific streams are well understood, specific detection criteria may be required for each individual query.
- myAlerts allows users to save any query, and define exact requirements for an alert to be generated. This may be temporal detection related (threshold for red/yellow alerts, minimum count, # of consecutive days alerting, etc), or can also be flagged as “Records of Interest” in which case any patient seen that matches the query will be alerted on.

The screenshot shows a configuration window for creating a 'myAlert'. The fields are as follows:

- Name of myAlert: gi in kids resp in adult myAlert
- Query: gi in kids resp in adult
- Enabled:
- This myAlert is being created for:  Records of Interest  Detection
- Detector: Regression/EWMA 1.1
- Red P-Value: .01
- Yellow P-Value: .05
- Minimum Count: 23
- 2 alerts in the past 7 days
- 2 consecutive alerts

Buttons at the bottom: Cancel, Save myAlert



# myAlert Results

## Detection based myAlerts

**Records of Interest Messages**

Manage Alert Definitions

Alerts | Records of Interest

Alert Definition	Date	Data Source	Detector	Level	Count	Expected	Timeseries
chills chiefcomplaint	03Oct06	ER by Patient	Regression/EWMA 1.2	0.049	16	9.86	<a href="#">Timeseries</a>
chills chiefcomplaint	04Oct06	ER by Patient	Regression/EWMA 1.2	0.039	13	9.64	<a href="#">Timeseries</a>
chills chiefcomplaint	06Oct06	ER by Patient	Regression/EWMA 1.2	0.044	16	9.86	<a href="#">Timeseries</a>
fever plus subsyndrome	04Oct06	ER by Patient	Regression/EWMA 1.2	0.039	571	515.46	<a href="#">Timeseries</a>
fever plus subsyndrome	05Oct06	ER by Patient	Regression/EWMA 1.2	0.016	568	516.89	<a href="#">Timeseries</a>
fever syndrome	04Oct06	ER by Patient	Regression/EWMA 1.2	0.005	420	363.36	<a href="#">Timeseries</a>
fever syndrome	05Oct06	ER by Patient	Regression/EWMA 1.2	0.000	424	363.96	<a href="#">Timeseries</a>
fever syndrome	07Oct06	ER by Patient	Regression/EWMA 1.2	0.013	415	366.39	<a href="#">Timeseries</a>
fever syndrome	08Oct06	ER by Patient	Regression/EWMA 1.2	0.006	419	366.32	<a href="#">Timeseries</a>
fever syndrome	10Oct06	ER by Patient	Regression/EWMA 1.2	0.029	415	371.36	<a href="#">Timeseries</a>

## Records of Interest based myAlerts

**Records of Interest Messages**

Manage Alert Definitions

Alerts | Records of Interest

Alert Definition	Date	Geography	Facility	Medical Grouping	Age Group	Sex	Data Details
fever and blood chiefcomplaint	10Oct06	OTHER_REGION		FEVER LOW BLOOD PRESSURE	45-64	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	09Oct06	PRINCE GEORGES		CHEST PAIN FEVER BLOOD VMT	18-44	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	08Oct06	PRINCE WILLIAM		COUGHING BLOOD FEVER	5-17	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	07Oct06	OTHER_REGION		FEVER LOW BLOOD PRESSURE	45-64	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	06Oct06	PRINCE GEORGES		CHEST PAIN FEVER BLOOD VMT	18-44	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	05Oct06	PRINCE WILLIAM		COUGHING BLOOD FEVER	5-17	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	05Oct06	WASHINGTON		FEVER SPITTING BLOOD	65+	Male	<a href="#">Data Details</a>
fever and blood chiefcomplaint	05Oct06	WASHINGTON		FEVER SPITTING BLOOD	65+	Male	<a href="#">Data Details</a>
fever and blood chiefcomplaint	04Oct06	OTHER_REGION		FEVER LOW BLOOD PRESSURE	45-64	Female	<a href="#">Data Details</a>
fever and blood chiefcomplaint	03Oct06	PRINCE GEORGES		CHEST PAIN FEVER BLOOD VMT	18-44	Female	<a href="#">Data Details</a>



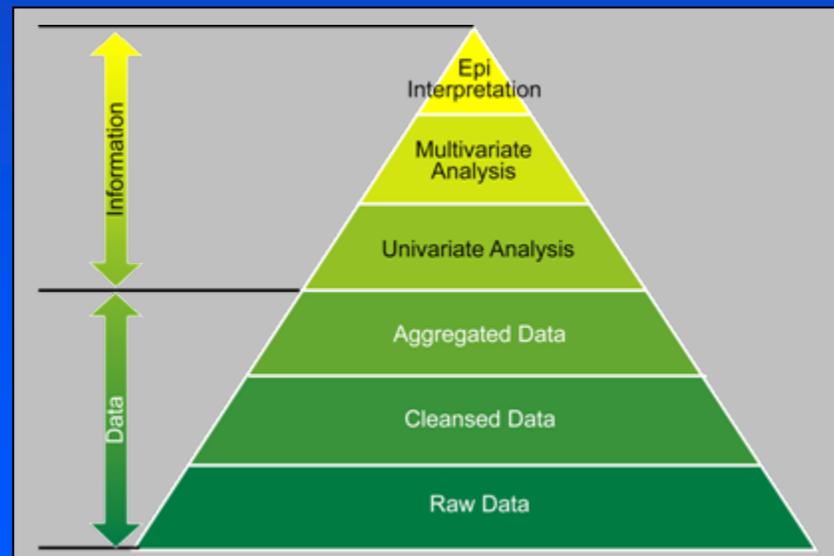
## **Additional information on *My Alerts***

- 1. Resolving the 'Boy Who Cried Wolf' Syndrome, M. Coletta, et al., 2006 ISDS Conference, Oct. 2006, *Advances in Disease Surveillance 2007*;2:99. Available at:  
<http://www.isdsjournal.org/article/view/2112/1668>**
- 2. *myAlerts: User-Defined Detection in Disease Surveillance Systems*, W. Loschen, et al., Submitted for Presentation at the 2009 Spring AMIA Conference.**



# Sample Project 3: *Infoshare* Overcoming Data Sharing Obstacles

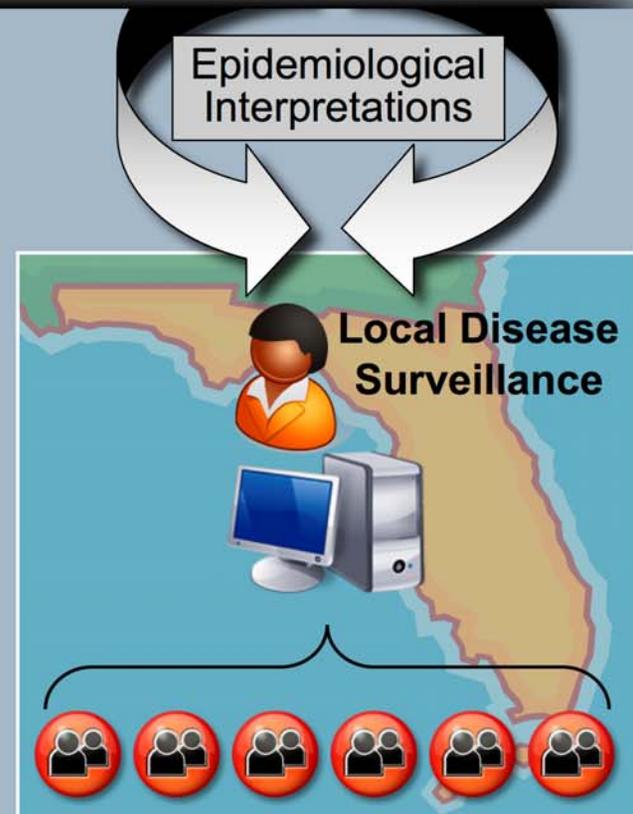
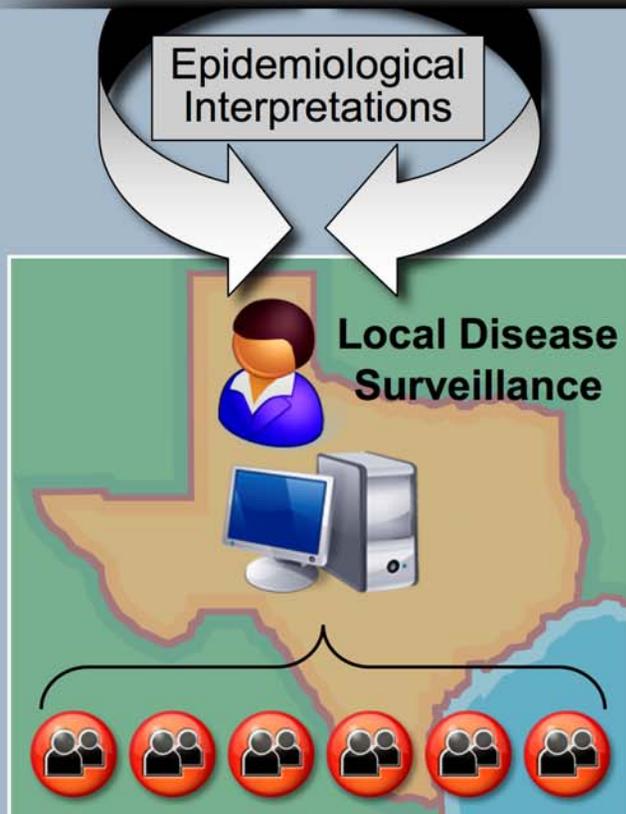
Epi Interpretation	Epi: “Respiratory outbreak we are currently investigating”
Multivariate Analysis	Fusion Detector: Respiratory has a Red Alert across 3 of 5 sources
Univariate Analysis	Detector: ER Respiratory visits are 4 times the normal rate
Aggregated Data	247 ER Respiratory visits, 1647 OTC Respiratory products sold
Cleansed Data	John Doe, 26, M, Sore Throat Jane Doe, 20, F, Shortness of Breath ...
Raw Data	John Doe, 26, M, Sore Throat John Doe, 26, M, Sore Throat Jane Doe, 1987, Female, SoB ...





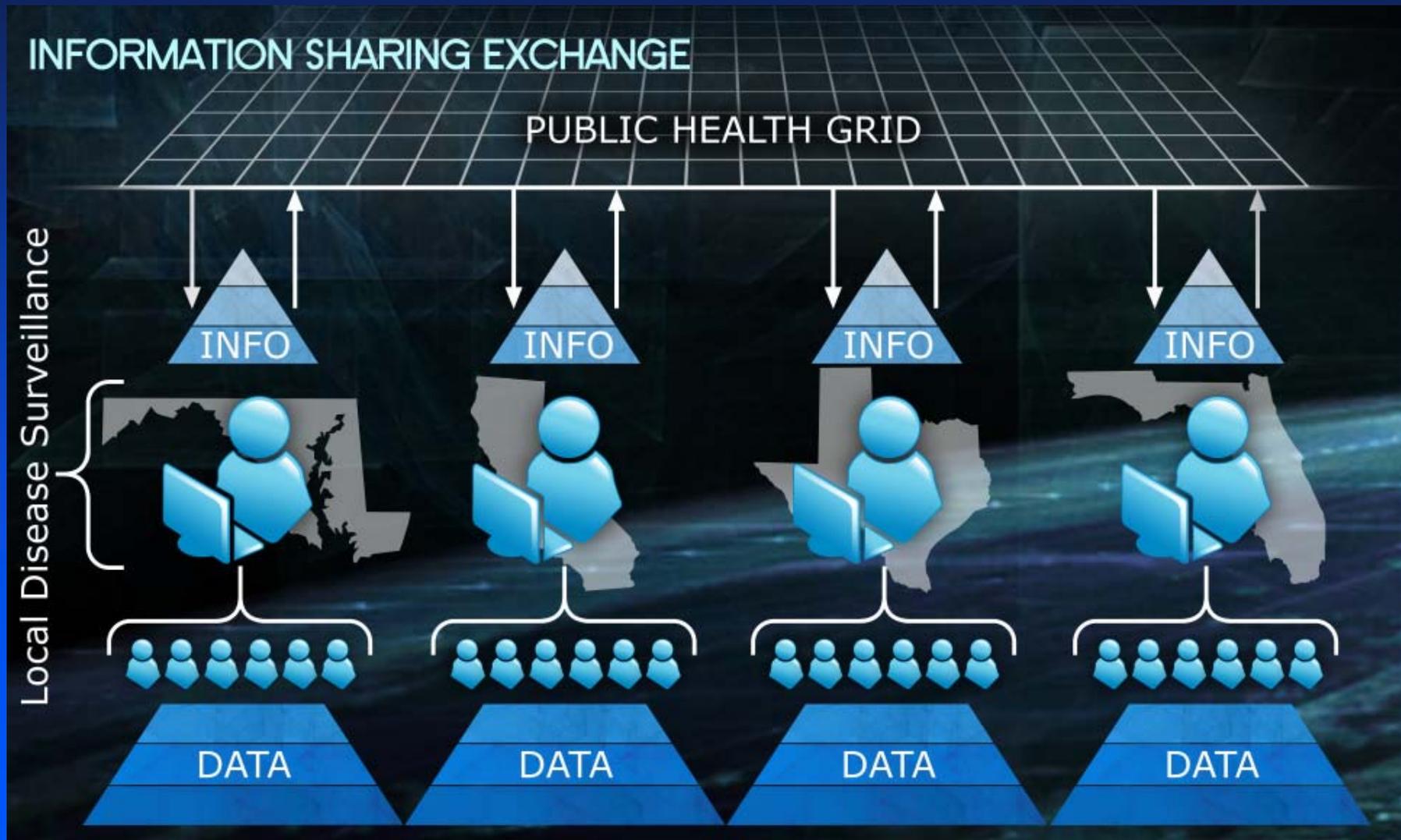
# Information Exchange Concept

## Information Sharing Exchange





# Information Sharing on the Grid

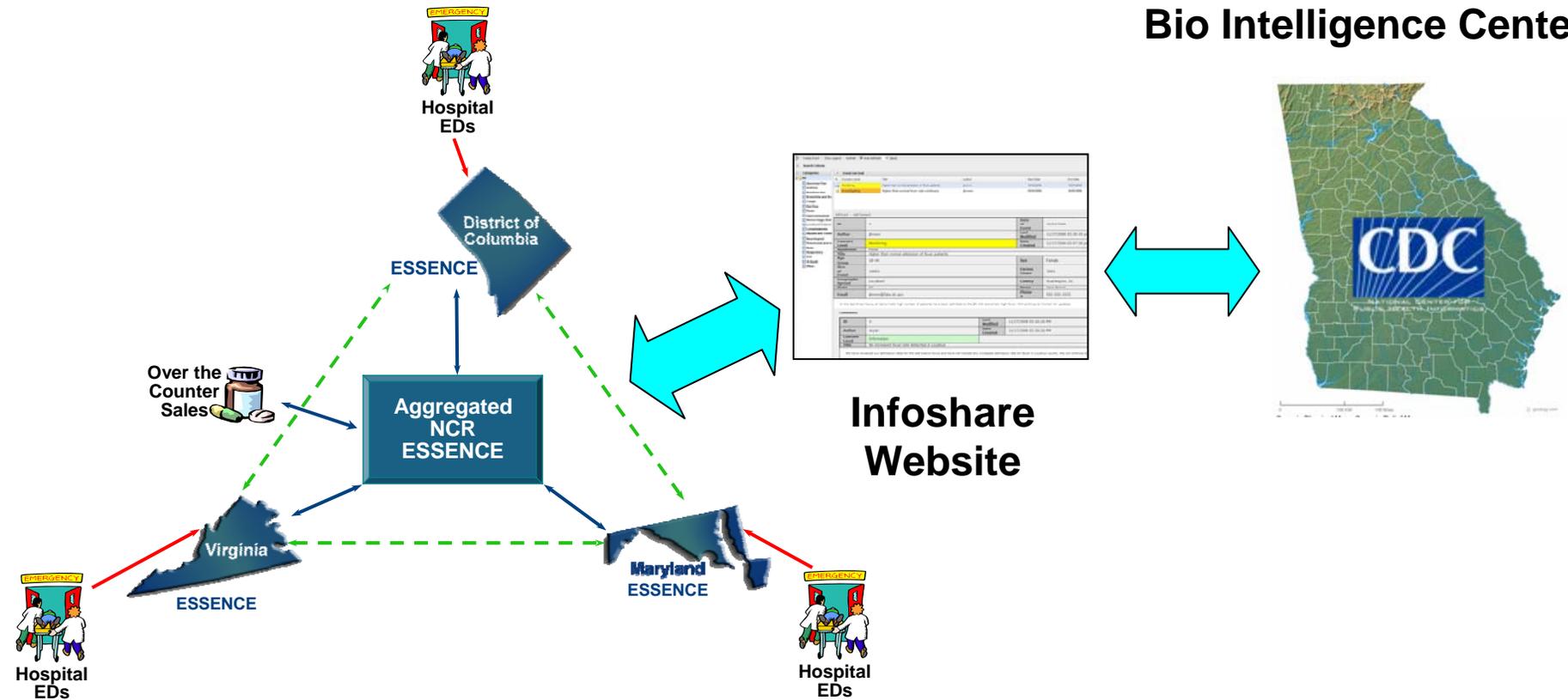




# Infoshare Used for the Inaugural NCR Regional Collaboration

## NCR Disease Surveillance Network

## Bio Intelligence Center





# Inaugural Infoshare Site

Create Event | Show Legend | Refresh |  Auto Refresh: 10 (secs)

- Search Criteria**
- Categories**
- All
  - Abominal Pain
  - Asthma
  - Botulism-like
  - Bronchitis and Bro
  - Cough
  - Diarrhea
  - Fever
  - Gastrointestinal
  - Hemorrhagic Illne
  - Localized Cutane
  - Lymphadenitis
  - Nausea and Vomit
  - Neurological
  - Pneumonia and lu
  - Rash
  - Respiratory
  - RSV
  - SI-Death
  - Other

**Event List Grid**

R	Concern Level	Title	Author	Start Date	End Date
	Monitoring	Higher than normal admission of fever patients	jbrown	10/03/2006	10/04/2006
	Investigating	Higher than normal fever rate continues	jbrown	10/04/2006	10/05/2006

Edit Event | Add Comment

<b>ID</b>	3	<b>Date of Event</b>	10/03/2006
<b>Author</b>	jbrown	<b>Last Modified</b>	11/17/2008 03:26:26 pm
<b>Concern Level</b>	Monitoring	<b>Date Created</b>	11/17/2008 03:07:56 pm
<b>Syndrome</b>	Fever		
<b>Title</b>	Higher than normal admission of fever patients		
<b>Age Group</b>	18-44	<b>Sex</b>	Female
<b>Size of Event</b>	1000's	<b>Excess Cases</b>	100's
<b>Geographic Spread</b>	Localized	<b>County</b>	Washington, DC
<b>State</b>	DC	<b>Name</b>	Jane Brown
<b>Email</b>	jbrown@fake.dc.gov	<b>Phone #</b>	555-555-5555

In the last three hours, an abnormally high number of patients have been admitted to the ER with extremely high fever. Will continue to monitor for updates.

**Comments:**

<b>ID</b>	4	<b>Last Modified</b>	11/17/2008 03:26:26 PM
<b>Author</b>	aryan	<b>Date Created</b>	11/17/2008 03:26:26 PM
<b>Concern Level</b>	Information		
<b>Title</b>	No increased fever rate detected in Loudoun		

We have reviewed our admission rates for the last twelve hours and have not noticed any increased admission rate for fever in Loudoun county. We will continue to



# Linkage within ESSENCE to *InfoShare*

The screenshot shows the ESSENCE web application interface. The top navigation bar includes tabs for 'History of ESSENCE', 'Syndrome Definitions', 'Detector Algorithms', 'Data Dictionary', and 'Help'. Below this, there are sub-tabs for 'Alert List', 'myAlerts', 'Event List', and 'Overview'. The 'myAlerts' tab is active, displaying a table of alerts. A yellow arrow points to a 'Share' button in the first row of the table. A 'Create Event' dialog box is open, showing pre-filled fields for an event, such as 'Age Group: 00-04', 'Sex: Male', 'Size of Event: 100's', and 'County: Alexandria'.

Share	Alert Definition	Date	Data
<a href="#">Share</a>	MontResp-Female	04Oct06	ER by
<a href="#">Share</a>	MontResp-Male	04Oct06	ER by
<a href="#">Share</a>	MontResp-Female	05Oct06	ER by
<a href="#">Share</a>	MontResp-Male	05Oct06	ER by
<a href="#">Share</a>	MontResp-Male	06Oct06	ER by

**Create Event**

Add Attachment

Message Type: [Dropdown]

Age Group: 00-04, 05-17, 18-44, 45-64, 65+ [Dropdown]

Sex: Male, Female, Unknown [Dropdown]

Size of Event: 100's [Dropdown]

Excess Cases: 80 [Dropdown]

Latitude: [Text]

Longitude: [Text]

Geographic Spread: Localized, Widespread [Dropdown]

County: Alexandria, Arlington, Fairfax, Loudoun, Montgomery [Dropdown]

City: [Text]

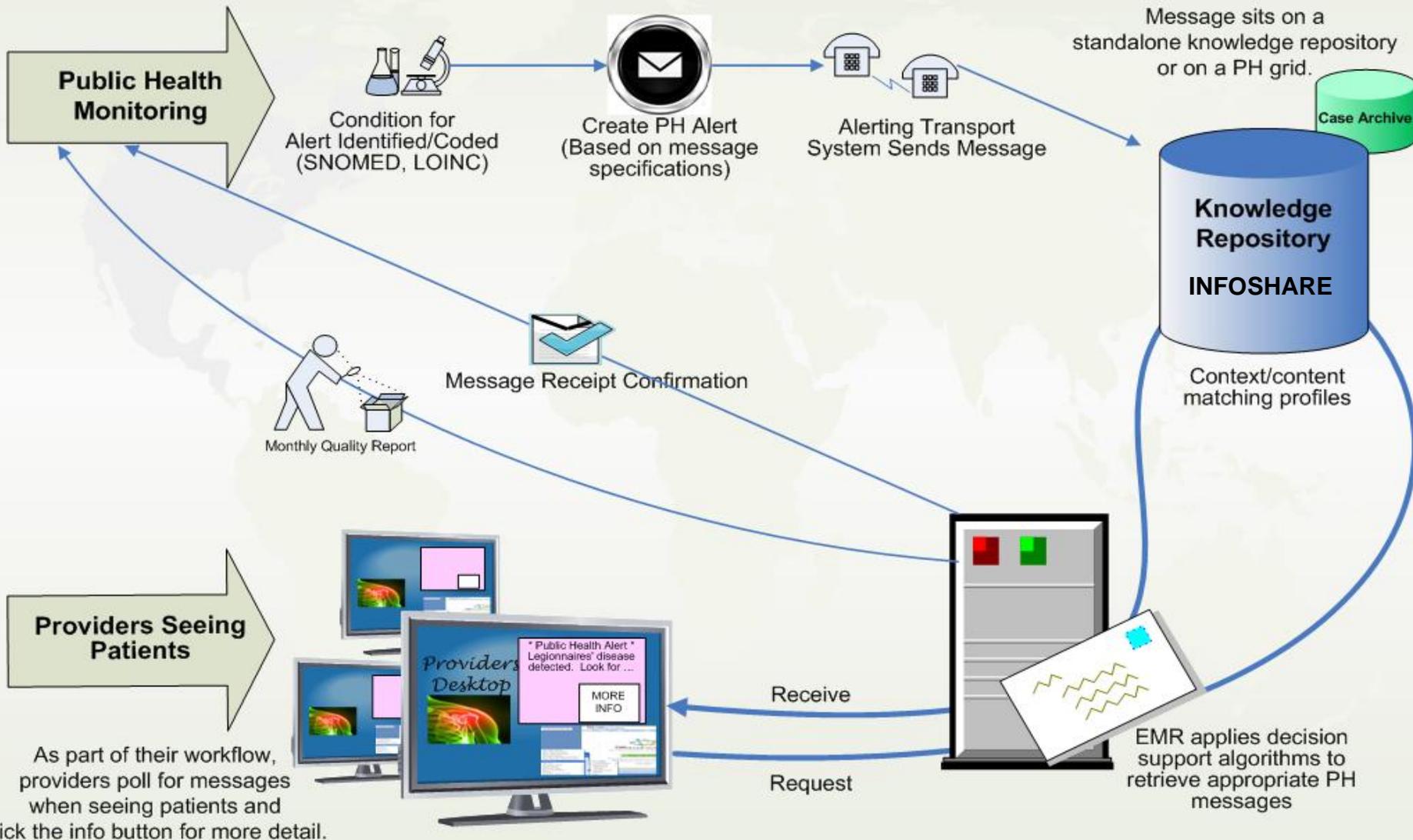
State: VA [Dropdown]

Preview Commit Cancel

**New Share Button added to myAlerts.**

**This allows users to create InfoShare messages directly from ESSENCE with most message fields pre-filled out.**

# CDC EMR Alerting Concept Diagram Alerting Data Flow v15 DRAFT



Courtesy Nedra Garrett, CDC

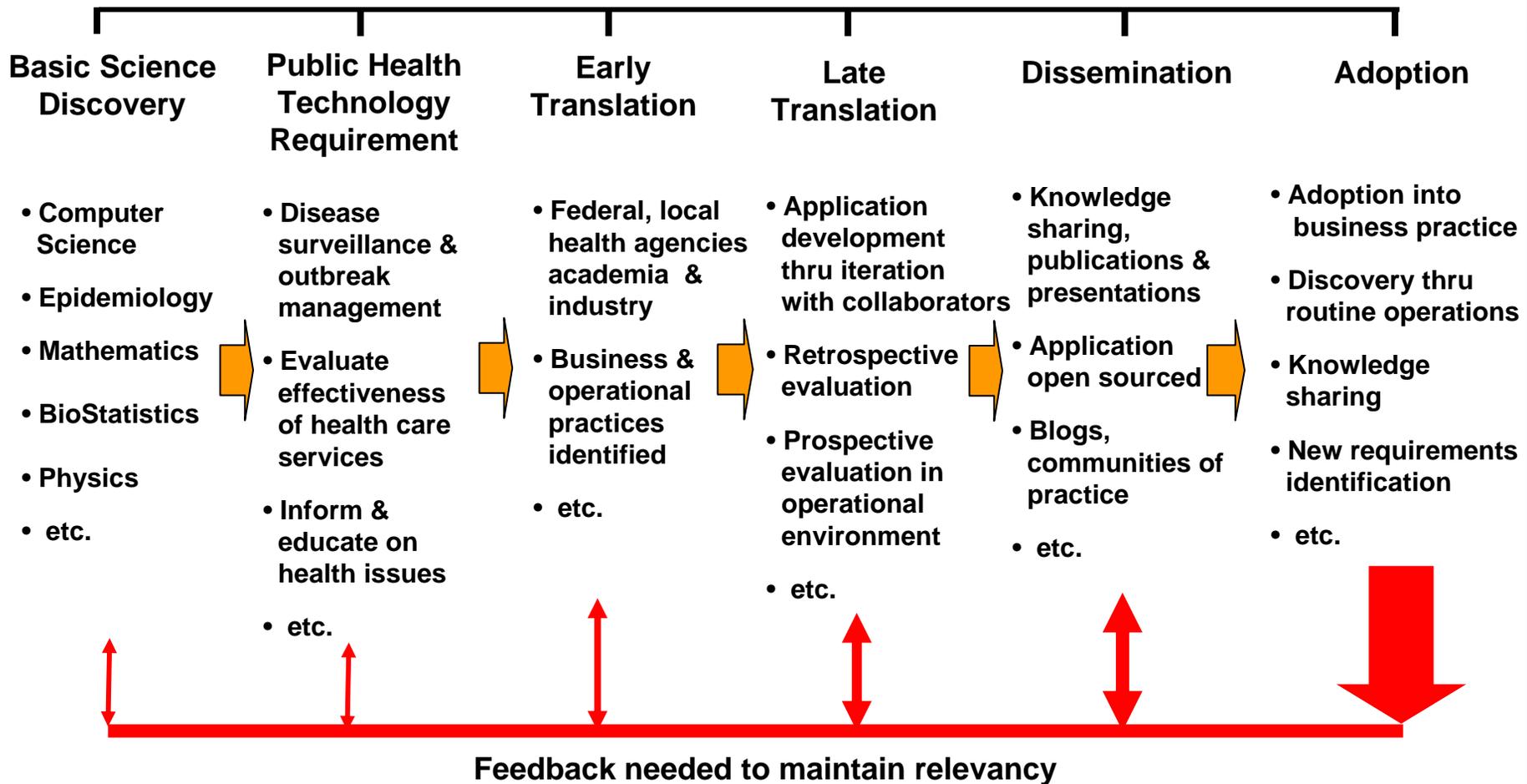


## Additional Information on *InfoShare*

1. Moving Data to Information Sharing in Disease Surveillance Systems, W. Loschen, et al., Spring 2007 AMIA Conference, May 2007.
2. Event Communications in a Regional Disease Surveillance System, W. Loschen, et al. AMIA 2007 Annual Symposium, Nov. 12, 2007.
3. Enhancing Event Communication in Disease Surveillance: ECC 2.0, N. Taberner, et al., 2007 PHIN Conference, Aug. 2007.
4. Enhancing Event Communication in Disease Surveillance: ECC 2.0, N. Taberner, et al., *Advances in Disease Surveillance* 2007;4:197. Available online at: <http://www.isdsjournal.org/article/view/2112/1668>
5. Super Bowl Surveillance: A Practical Exercise in Inter-Jurisdictional Public Health Information Sharing, C. Sniegowski, *Advances in Disease Surveillance* 2007;4:195. Available online at: <http://www.isdsjournal.org/article/view/2106/1666>
6. Structured Information Sharing in Disease Surveillance Systems, W. Loschen, et al., *Advances in Disease Surveillance* 2007;4:101. Available online at: <http://www.isdsjournal.org/article/view/1997/1552>
7. Disease Surveillance Information Sharing, N. Taberner, et al., 2008 PHIN Conference, Aug. 2008.
8. Methods for Information Sharing to Support Health Monitoring, W. Loschen, *APL Technical Digest*. 2008; 27(4): 340-346. Available online at: <http://techdigest.jhuapl.edu/td2704/loschen.pdf>



# Is this the Correct Translational Research Continuum for Public Health Informatics?





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### **Prince George's County Department of Health**

**Ms. Joan Wright-Andoh**

### **District of Columbia Department of Health**

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**Dr. Charles Konigsberg**

### **Consultant**

**Ms. Kathy Hurt-Mullen**



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### **University of Utah**

**Catherine Staes**

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### **General Electric Healthcare**

**Keith Boone**

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# Discussion

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