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Experience with a Clinical Data Repository and Warehouse Adam Wilcox, PhD Columbia University March 24, 2009

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Outline

History Clinical Data Repository Clinical Data Warehouse

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Clinical Information Systems

Stage 1: Early computers calculated data in context
Stage 2: Client applications provided access to ancillary data
Stage 3: Systems began aggregating data from multiple sources
Stage 4: Data storage provided historical view
And analysis
Stage 5: Workflow applications formalize processes between clinical roles ation Systems

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Clinical Information System Technology Levels

Level 1: Departmental applications

Level 2: Internally-developed integrated systems

Level 3: Functional vendor-based systems

Level 4: Comprehensive clinical information systems

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Clinical Information Systems at Columbia University

Began at Stage 3 Pushing a Level 1 system to Level 2 Issues Vocabulary Data modeling Interfaces Decision support Data processing Recipient of first Nicholas Davies Award

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Image: Clinical Information Systems Architecture.

<u>Slide 7</u> Image: EMR environment <u>Slide 8</u>

Image: Architecture. Handling, Encoding, Routing, Monitoring, Access

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Other Level 2 Systems Intermountain VA Partners Regenstrief Vanderbilt

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Level 3 Systems Cerner Epic Eclipsys GE McKesson

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Challenges at Columbia

Moved from Stage 3 through Stage 4 to Stage 5 Purchased a vendor system (Level 3) How to get to Stage 5 and Level 4?

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Challenges at CPMC/CUMC/NYPH/WCMC

In 1998, merged two academic medical centers into NewYork Presbyterian Hospital Columbia Presbyterian campus became Columbia University Medical Center New York Hospital became Weill Cornell Medical Center Currently 4 different electronic health records Eclipsys (WCMC) Eclipsys (CUMC) Epic (WCMC) Allscripts (CUMC)

<u>Slide 13</u> Image: NYP Computing, Environment Sep 2007

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Image: Integrating Among Multiple EHRs Eclipsys (CUMC) Eclipsys (WCMC) Allscripts Epic

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Problems with Integrating to Application Databases

Must model each system multiple times Increased effort and complexity Overloading workflow databases Protecting external data consistency (no updates) Increased complexity of data protection Bringing in data for a new patient When to pull data in Interfaces don't naturally pull in historical data Increases complexity as move toward RHIOs

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Image: Repository Model Eclipsys (CUMC) Eclipsys (WCMC) Clinical Data Repository Allscripts Epic

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Benefits of CDR

Only model data from source systems once Common data store Data are read only Optimized for read Historical data included Web-based viewer adaptable to multiple applications Adaptable to future health information exchange efforts Platform of innovation

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Optimized for Retrieval

Relational structure can be difficult to query for both data and context Gathering multiple elements requires multiple table joins Good for data storage Good for aggregating across multiple patients Event-based model good for querying across data types Data organized according to patient Not good for querying across patients

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Retrieval optimization

Paradigm shift in how data are used Paper records mainly for primary use Electronic allows secondary use Secondary use can be multiple times

<u>Slide 20</u> Image: CDR View in Eclipsys

Slide 21 Graph: Proportion of CDR Viewer Access

<u>Slide 22</u> Graph: Increase in CDR View Access

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CUMC/NYP Clinical Data Warehouse History

1994: Created, sponsored by Columbia University Department of Medical Informatics and Office of Clinical Trials
Populated with data from existing clinical data repository
Supporting clinical research
1998: Columbia + Cornell = NewYork Presbyterian Hospital
Warehouse funded by NYPH
Goal to incorporate and provide data across whole system
2004: Formal analysis of CDW user needs by Clinical Quality and Information
Technology Committee (CQIT)
Creation of Data Warehousing Subgroup
Need to bring together disparate clinical data sources
Need to manage user requests for data

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Uses of the Warehouse Clinical research queries Management reports Clinical trial recruitment

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CDW Content Issues

Began as a copy of the repository Data already gathered Mainly for research queries Some data marts built for common queries Ability to query rapidly across patients increases security risk

<u>Slide 26</u> Screen Grab

<u>Slide 27</u> Screen Grab

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Goal of Access Policy

Provide broader access to data Central control is resource limited Allow collection of more data sources Reassure data stewards Three separate institutions Data ownership not completely defined for all data

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CDW Structure

Identifying data Patient identifying information Main data Event tables for clinical repository Lookup tables Vocabulary translation Contains no patient data Specialty data marts

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Access Policy Identifying data Most restricted Create a research identifier to replace the patient ID Allow access to only ResearchID, sex, birth date (month and year only), marital status, race, death status Specialty data Access policy defined by data steward Patient clinical data No access to text data Modified dates Lookup tables Full access (contain no patient data)

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Access Policy Specific patient information Sometimes needed to create initial queries Analysts get access only to a randomly selected subset Access request through supervisor De-identified patient data Test patients Full access given

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CUMC/NYP Clinical Data Warehouse History

1994: Created, sponsored by Columbia University Department of Medical Informatics and Office of Clinical Trials Populated with data from existing clinical data repository Supporting clinical research 1998: Columbia + Cornell = NewYork Presbyterian Hospital Warehouse funded by NYPH Goal to incorporate and provide data across whole system 2004: Formal analysis of CDW user needs by Clinical Quality and Information Technology Committee (CQIT) Creation of Data Warehousing Subgroup Need to bring together disparate clinical data sources Need to manage user requests for data

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Analysis of Challenges

Data in vendor-based transactional systems Could not query across transactional systems Users needed help in defining their needs Mature initiatives required more robust data solutions

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Graph: Pneumonia Core Measures

<u>Slide 35</u> Graph: Pneumonia Core Measures

<u>Slide 36</u> Image: VIRTUAL CLINICAL DATA WAREHOUSE

<u>Slide 37</u> Graph

<u>Slide 38</u> Image: VIRTUAL DATA WAREHOUSE

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Conclusion

Integrating clinical data repository view into workflow applications can improve use Access policies need to isolate data to reassure data use from different stakeholders Data access tools need to account for users' evolving data needs along the quality improvement life cycle