



Prof. Pascal Coorevits

**Certification and Labelling Adviser, EuroRec
Representative**

Assessing the data quality of hospital EHRs



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DATA

HEALTH

The European Institute For
Innovation Through Health Data

Assessing the data quality in hospital EHRs

Pascal Coorevits

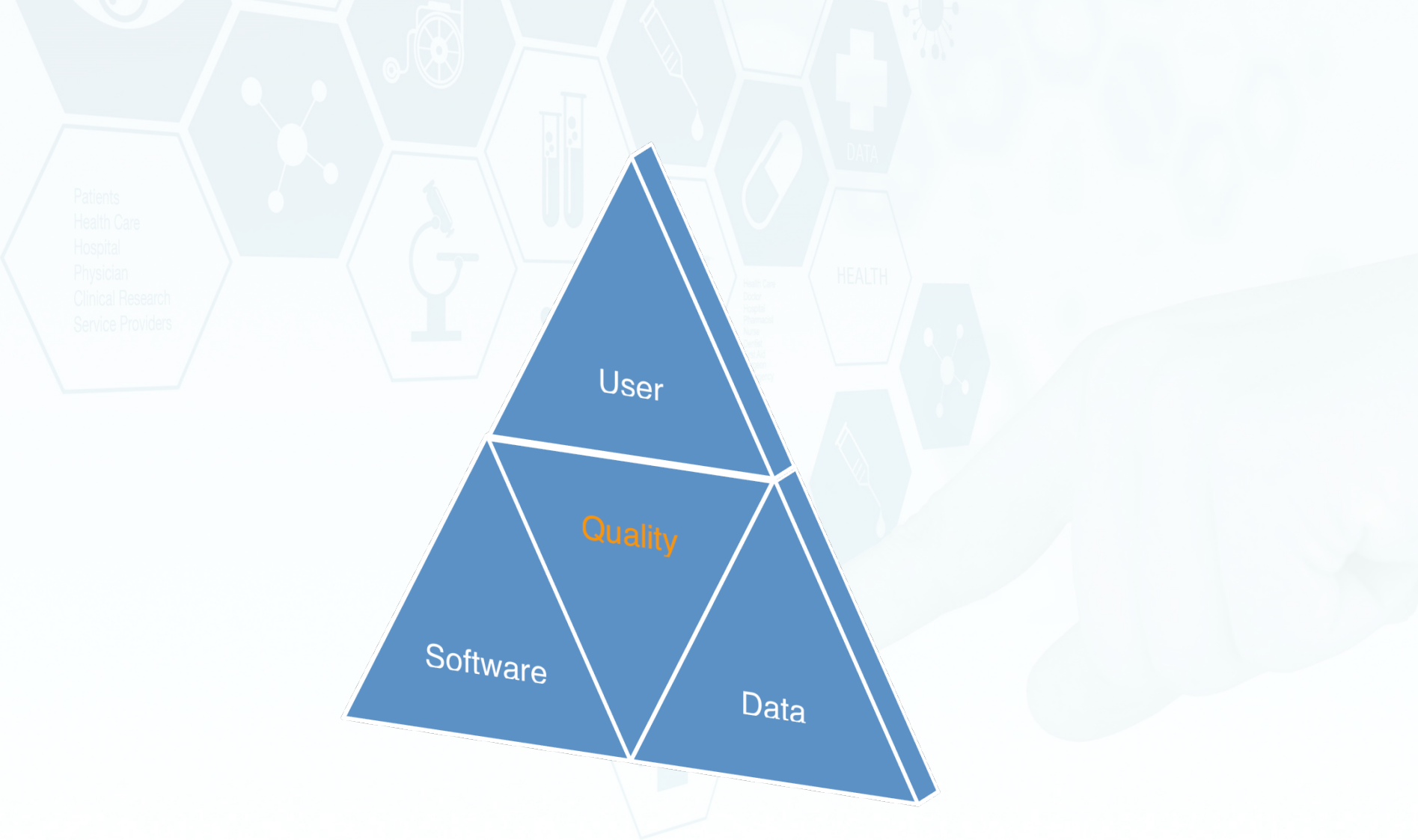
Professor of Medical Informatics and Statistics, Ghent University
Vice-President for Research, the EuroRec Institute
Data Quality Task Force Lead, i~HD



11 > 14 SEPTEMBER 2019

Quality of health data is vital

- ❑ Patients and clinicians want health data to be safe, rapid and evidence based
- ❑ Healthcare managers also want to use resources efficiently and need insights for strategic planning
- ❑ Public health agencies need reliable data to guide healthcare and prevention programmes
- ❑ Healthcare funders need good quality data to reward high-quality and value-based care
- ❑ Pharma wants to re-use EHRs to accelerate clinical research
- ❑ Regulators and HTA agencies want to be able to trust Real World Evidence in decision making
- ❑ Everyone wants to achieve the best patient outcomes and they all know that good data is a critical success factor



Electronic Health Records

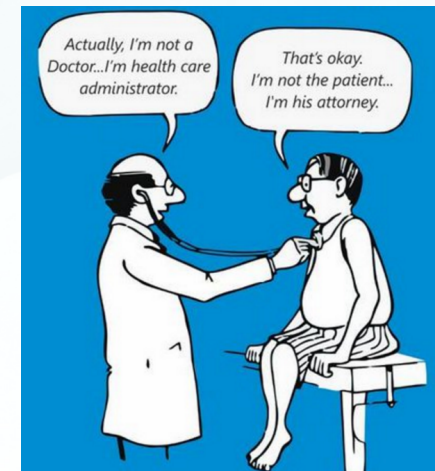
- ❑ Electronic Health Records offer enormous potential to improve the safety, quality and efficiency of healthcare
- ❑ Empirical evidence for the beneficial impacts of most eHealth technologies on the quality and safety of healthcare are often lacking, or at best, only modest
- ❑ Some of the reported issues: lacking of key EHR features, not using EHRs to their full potential, poor interoperability, low usability, low EHR (data) quality, etc...

Quality of EHRs and EHR data

- ❑ To use EHRs efficiently for daily routine care, for clinical research, for big data analytics, ... a number of functionalities are needed (e.g. security, confidentiality, trustworthiness, ...)
- ❑ Mechanisms are required to ensure e.g. data correctness, completeness, consistency, ...

Quality assurance is essential

Quality labelling & certification are needed



i~HD Data Quality Taskforce

Patients



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[Read short bio](#)



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Juan M Garcia-Gomez
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**Agustín Gómez de
la Cámara**
Hospital 12 de Octubre
Research Institute

[Read short bio](#)



Dipak Kalra
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Carlos Sáez
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**Bart
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Janssen Pharmaceuticals

[Read short bio](#)



Veli Stroetmann
Empirica

[Read short bio](#)



Diane Whitehouse
EHTEL

[Read short bio](#)

i~HD Data Quality Taskforce aims

- Develop data quality assessment methods, tools and improvement strategies to maximise quality of health data
- Promote the importance of data quality
- Guidance in assessing and improving data quality
- Scale up a multi-stakeholder understanding and commitment to increase data quality

Methods and dimensions of electronic health record data quality assessment: enabling reuse for clinical research

Nicole Gray Weiskopf, Chunhua Weng

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A Harmonized Data Quality Assessment Terminology and Framework for the Secondary Use of Electronic Health Record Data

Michael G. Kahn

Secondary Use of EHR: Data Quality Issues and Informatics Opportunities

Taxiarchis Botsis^{ab}, Gunnar Hartvigsen^{ac}, Fei Chen^b, Chunhua Weng^b

A practical framework for data management processes and their evaluation in population-based medical registries

M. SARIYAR¹, A. BORG¹, O. HEIDINGER² & K. POMMERENING¹

A Pragmatic Framework for Single-site and Multisite Data Quality Assessment in Electronic Health Record-based Clinical Research

Michael G. Kahn, MD, PhD,*† Marsha A. Raebel, PharmD,‡§ Jason M. Glanz, PhD, MS,‡¶
Karen Riedlinger, MPH, MT (ASCP), ¶ and John F. Steiner, MD, MPH,‡

A Data Quality Assessment Guideline for Electronic Health Record Data Reuse

Nicole G. Weiskopf, PhD¹ Suzanne Bakken, RN, PhD^{2,3,4} George Hripcsak, MD, MS⁵ Chunhua Weng, PhD⁶

Applying probabilistic temporal and multisite data quality control methods to a public health mortality registry in Spain: a systematic approach to quality control of repositories

Carlos Sáez^{1,2}, Oscar Zurriaga^{3,4,5}, Jordi Pérez-Panadés³, Inma Melchor³, Montserrat Robles¹ and Juan M García-Gómez^{1,6}

RECEIVED 30 August 2015
REVISED 21 December 2015
ACCEPTED 17 January 2016

AMIA
OXFORD
UNIVERSITY PRESS

Work on data quality dimensions and assessment methods

Work on data quality dimensions and assessment methods

- Several data quality indicators used
- Several definitions for the same concept
- No universal approach

→ mapping exercise

→ 9 quality dimensions



Completeness – “data values are present”

Data Group	Data Item	Avg. usage	Site 1	Site 2	Site 3	Site 4	Site 5	Site 6	Site 7	Site 9	Site 9
Demographics	Gender	100%	100,00%	100,00%	100,00%	100,00%	100,00%	100%	100%	100,0%	100,00%
Demographics	Case Status	96%	99,87%	100,00%	60,00%	100,00%	100,00%	100,00%	100%	100,0%	100,00%
Demographics	Date of Birth	89%	100,00%	100,00%	99,00%	NA	100,00%	100%	100%	100,0%	100,00%
Demographics	Admission date	84%	100,00%	100,00%	100,00%	NA	100,00%	99,53%	58%	100,0%	100,00%
Diagnosis	Diagnosis Text	81%	50,46%	84,02%	100,00%	100,00%	98,05%	100,00%	14%	100,0%	80,98%
Diagnosis	Diagnosis Code	81%	50,46%	84,02%	100,00%	100,00%	98,05%	100,00%	14%	100,0%	80,98%
Demographics	Discharge date	75%	100,00%	100,00%	100,00%	NA	100,00%	100,00%	58%	100,0%	14,18%
Diagnosis	Diagnosis Date	70%	50,46%	84,02%	100,00%	100,00%	100,00%	NA	13%	100,0%	80,98%
Medication	Dosage	25%	20,36%	0,00%	NA	NA	94,43%	95%	NA	NA	12,21%
Findings	Weight	25%	29,56%	18,24%	NA	NA	89,17%	27,20%	36%	7,5%	13,82%
Laboratory Findings	Platelets Blood	48%	52,78%	33,14%	63,73%	NA	100,00%	100%	45%	NA	33,88%
Laboratory Findings	SGPT (ALT) in serum	47%	33,61%	22,29%	100,00%	NA	100,00%	100%	47%	NA	21,86%
Laboratory Findings	Total Protein in serum	46%	52,37%	14,96%	86,53%	NA	100,00%	100%	47%	NA	16,34%
Laboratory Findings	Total Bilirubin in serum	46%	33,03%	16,99%	100,00%	NA	100,00%	100%	47%	NA	19,58%



Variables such as Weight are quite frequently not present

Doods et al. *Trials* 2014, 15:18
<http://www.trialsjournal.com/content/15/1/18>



RESEARCH

Open Access

A European inventory of common electronic health record data elements for clinical trial feasibility

Justin Doods¹, Florence Botteri², Martin Dugas¹, Fleur Fritz^{1*} and on behalf of EHR4CR WP7

- Data quality issues found in a survival analysis of pancreatic cancer patients (Columbia University Medical Center, New York)
- Information inconsistency between different EHR data sources:
 - In a few cases, pancreatitis was **diagnosed as being chronic in the pathology reports** but it was **reported as being only acute in the clinical notes**
- Information inconsistency within the same data sources :
 - Some patients received **simultaneously two different ICD-9-CM codes** for their diagnoses of diabetes, both **250.01 and 250.02 for type-1 and type-2 respectively**

[Summit on Translat Bioinforma](#). 2010; 2010: 1–5.
Published online 2010 Mar 1.

PMCID: PMC3041534

Consistency – “Data satisfies constraints”

Secondary Use of EHR: Data Quality Issues and Informatics Opportunities

[Taxiarchis Botsis](#),^{a,b} [Gunnar Hartvigsen](#),^{a,c} [Fei Chen](#),^b and [Chunhua Weng](#)^b

Clinical height measurements are unreliable: a call for improvement

A. L. Mikula¹ · S. J. Hetzel² · N. Binkley³ · P. A. Anderson⁴

“Fifty percent of staff reported they on occassion enter patient reported height into the EHR rather than performing a measurement”

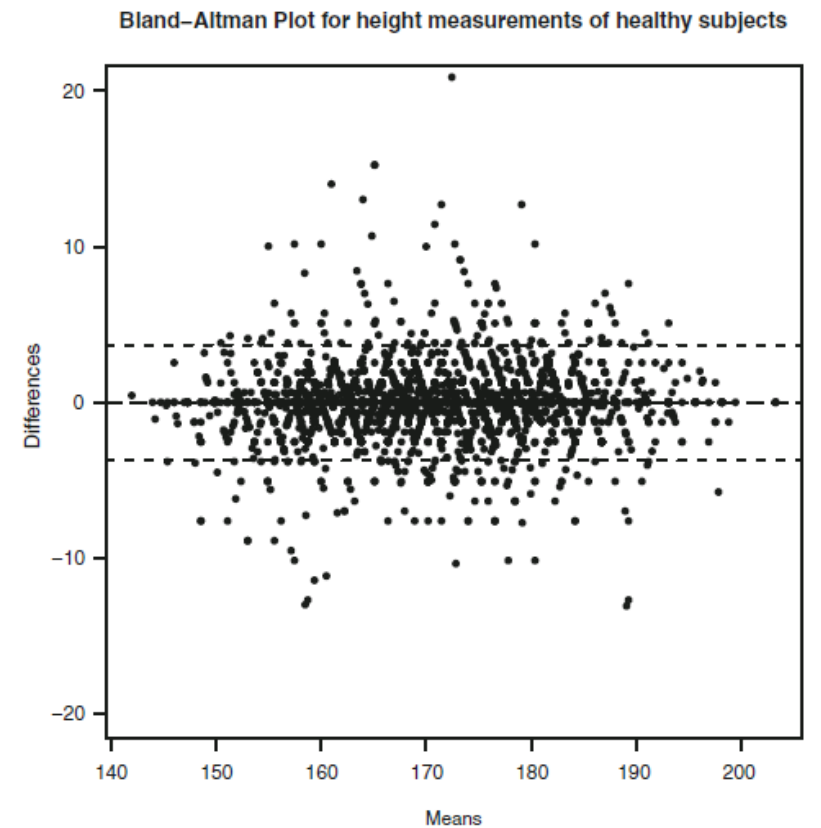


Fig. 4 Bland-Altman plot for height measurements of subjects. Each data point represents a single patient. *X axis* represent mean patient height in centimeters. *Y axis* represents difference between the first and the last height measurement for the individual patients in centimeters. The *dotted lines* represent 95 % CI



Hospital Network Workshop



The European Institute For Innovation Through Health Data

February 9th, 2017

Brussels, Belgium

www.i-hd.eu

enriching knowledge and enhancing care through health data

Service Providers

i~HD Hospital Network of Excellence Data Quality Workshop Towards better data quality in hospitals

Tuesday 23rd May 2017 - Wednesday 24th May 2017

Realising the Value from Health Data ~
Improving Care and Research

JOINT EVENT



The European Institute For Innovation Through Health Data



September 21-22, 2017

MADRID, SPAIN

i+12

Instituto de Investigación
Hospital 12 de Octubre

Annual Conference 2017

www.i-hd.eu/AC2017





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A two-day workshop run by the **i~HD**
Hospital Network of Excellence and Data Quality Task Force,
in collaboration with

Gaining the Benefits of Improved Health Data Quality



Realising the value of health data~
showcasing practical examples from research and healthcare

REPORT

i~HD Annual Conference
Gothenburg, Sweden 19 - 20 November 2018



i~HD@
LIFE SCIENCE
LIVE **MAY 15 - 16, 2019**
TURKU, FINLAND

Service Providers

i~HD @ VITALIS
21 - 23 May 2019
Gothenburg



CONGRESS '19
EAHM
BELGIUM - GHENT
11-14 September, 2019
Ghent, Belgium



**i~HD Session
(Workshop 2)**
“The reuse of EHRs for Learning Health Systems”

When: Friday 13.09.2019
Timing: 4 PM – 5:30 PM
Location: Meet & Greet Center
in the Ghelamco Arena, Gent

*Uncover the insights
hidden in your EHRs*



Joining the Dots Conference - 27 & 28 November - Brussels



Save the date - registration opens soon

Assessing the Quality of Congestive Heart Failure outcomes data at the Hospital del Mar Barcelona

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Pascal Coorevits
Geert Byttebier
Dipak Kalra
Geert Thienpont
Carlos Sáez
Juan M. García Gómez
Marta Durá-Hernández
Juan-Manuel Ramírez Anguita
Miguel-Angel Mayer



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DE VALÈNCIA



Pilot DQA

- Scoping of the DQA
- Congestive Heart Failure
- 146.602 patient visits with diagnosis of CHF
- ICHOM “Heart Failure” outcome variables basis for selection of variables for DQA → 22 variables (out of 72 ICHOM variables) were selected
- Data Quality Dimensions: Completeness, Correctness, Consistency, Uniqueness & Stability

Congestive Heart Failure (CHF) – ICD9 codes

428 Heart failure

Code, if applicable, heart failure due to hypertension first (402.0-402.9, with fifth-digit 1 or 404.0-404.9 with fifth-digit 1 or 3)

Excludes:

rheumatic (398.91)

that complicating:

abortion (634-638 with .7, 639.8)

ectopic or molar pregnancy (639.8)

labor or delivery (668.1, 669.4)

428.0 Congestive heart failure, unspecified

Congestive heart disease

Right heart failure (secondary to left heart failure)

Excludes:

fluid overload NOS (276.6)

428.1 Left heart failure

Acute edema of lung with heart disease NOS or heart failure

Acute pulmonary edema with heart disease NOS or heart failure

Cardiac asthma

Left ventricular failure

428.2 Systolic heart failure

Excludes:

combined systolic and diastolic heart failure (428.40-428.43)

428.3 Diastolic heart failure

Excludes:

combined systolic and diastolic heart failure (428.40-428.43)

428.4 Combined systolic and diastolic heart failure

428.9 Heart failure, unspecified

Cardiac failure NOS

Heart failure NOS

Myocardial failure NOS

Weak heart



402 Hypertensive heart disease

402.0 Malignant


402.00 Without heart failure





402.01 With heart failure

R package: automated pdf & html DQ reporting

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Data and configuration files

 **main_config.csv**
DQ analysis parameters, including paths to:

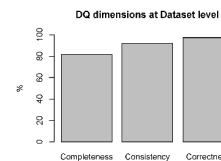
-  **dataset.csv**
The dataset to be analyzed
-  **dictionary_types.csv**
-  **dictionary_ranges.csv**
-  **dictionary_rules.R**

Data dictionary files with semantics about data types, ranges and multivariate rules

Data Quality Vector - R output

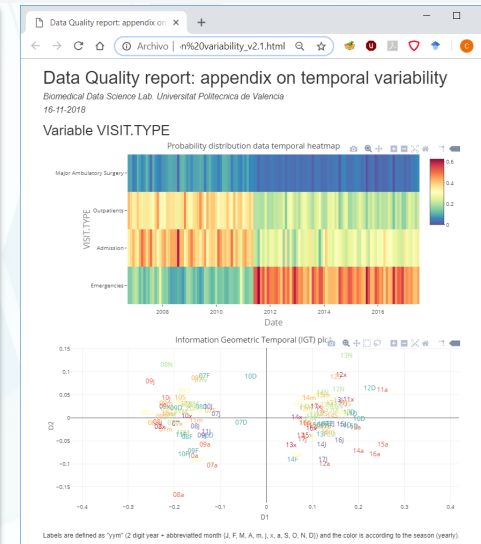
Data Quality report
Biomedical Data Science Lab. Universitat Politècnica de València
16-11-2018

Overall DQ analysis results of the Perinatal-Infant-Feeding dataset
The dataset contains 1893 rows and 15 columns. The following metrics were obtained: Completeness 91.31, Consistency 91.97 and Correctness 97.31.



Consistency detailed results
The following metrics were obtained by the Internal Consistency analysis: Consistency by type 8.89, Consistency by range 4.93 and Consistency by multivariate rules 0.05. In the next page, the detailed results for the Internal Consistency analysis are shown.

Page 1



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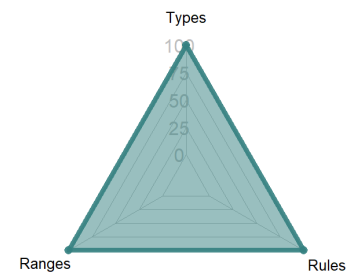
Types → dictionary_types.csv

Ranges → dictionary_ranges.csv

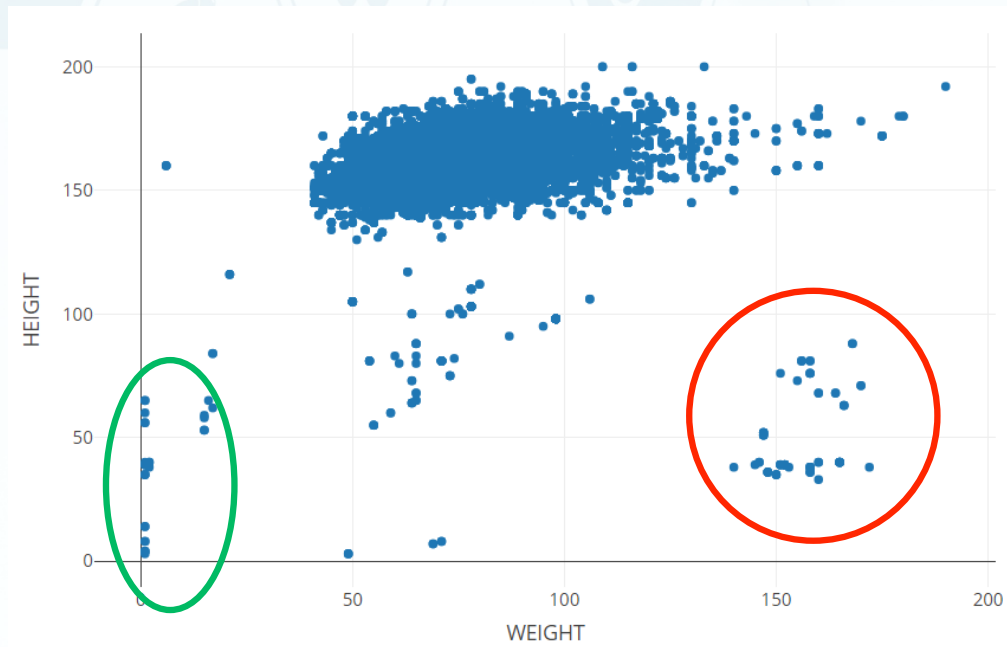
Multivariate rule	Comment	Error.su m
ARIDATE <= DISDATE	Arrival date <= Discharge date	0
DATEofBIRTH <= DEATHDATE	Birth date <= Death date	0
DISDATE <= DEATHDATE	Discharge date <= Death date	0
DATEofBIRTH <= ARIDATE	Birth date <= Arrival date	0
$(WEIGHT/(HEIGHT/100)^2) <= 70$	BMI under 70	180
$(WEIGHT/(HEIGHT/100)^2) >= 10$	BMI over 10	15

By Types: 99,99 %
By Ranges: 100 %
By Multivariate rules: 99,98 %

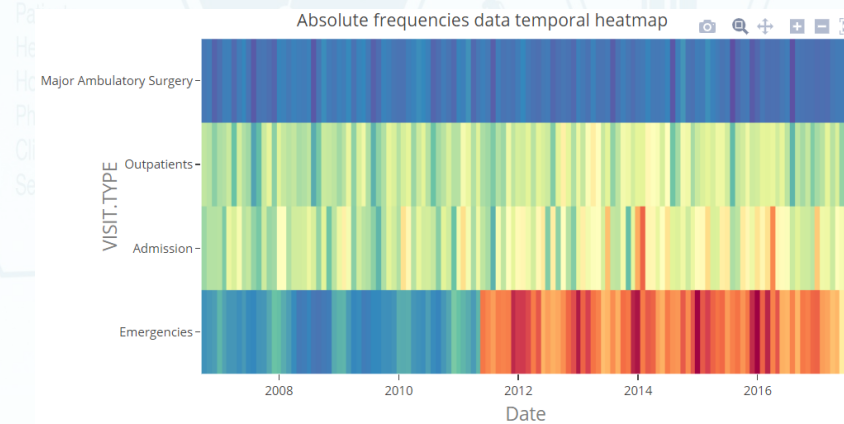
Consistency results by types, ranges and rules



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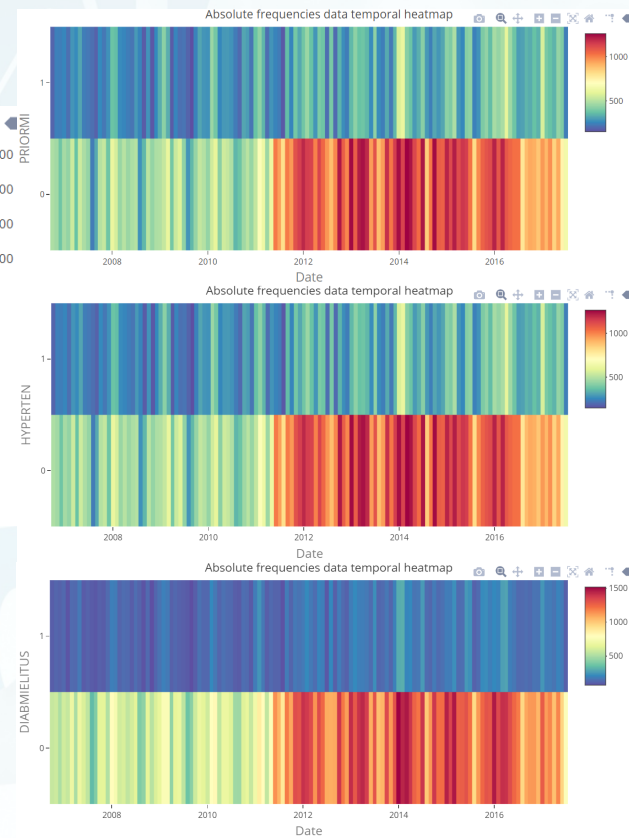
Slide courtesy of Carlos Saez and Juan M. García Gómez, Universitat Politècnica de València



Abrupt change in June 2011 in the frequencies of:

- Visit type
- Prior Myocardial Infarction
- Hypertension
- Diabetes Mellitus

New automated-coding system introduced in Emergencies at that date (before in text)





*Report - Pilot Project -
Hospital Parc de Salut Mar
Barcelona, Catalonia, Spain*

DQS4H
Data Quality Service
for Hospitals

Table of content

Executive Summary	4
Scope of the assessment	4
Method	4
Summary of findings	5
Conclusion and recommendations	6
List of Abbreviations	8
Introduction	9
Methodology	10
Preparatory steps	10
Description of the EHR data repository to be assessed	10
Population	12
ICHOM standard dataset	12
Construction of the requested dataset	12
Description of the format of the expected dataset	12
Description of the derived variables	15
Description of the Data Quality Dimensions	16
Tools	17
Results	18
Preliminary analysis of the received dataset	18
Descriptive statistics	19
Age at hospitalisation	20
Sex	22
Visit type	23
Congestive Heart Failure (CHF)	24
Atrial Fibrillation	24
Prior MI	25
Smoking Status	25
Alcohol use	26
Number of Hospital admissions during the past year	26
Number of Hospital appointments during the past year	28
Hypertension	29
Diabetes Melitus	30

• It is essential to have **considerable knowledge of the EHR** (types of data available, how the data were collected or who collected it)

• The **assessment** of the data is the very first step to improve the quality of your data

• Once you know about the quality of your data, it is important to **monitor it regularly**

• **Multidisciplinary** approach is highly recommended

• Thinking of using EHR for different purposes such as research, EHR data models would need to be **expanded and redesigned** and data quality assessment can assist in doing these tasks

• It is of value that an **external assessment** of the data quality is performed by an independent organization

• High-quality data **enhance the validity and reliability** of study findings

• It is critical to ensure that **the metrics** are feasible, valid, and meaningful for a specific EHR and purpose and its quality improvement

Prof. dr. Miguel-Angel Mayer – Hospital del Mar - Barcelona

**Do you want to make better use of your
electronic health record?**

Do you want to improve the quality of your health data?

DQS4H

***Data Quality Service
for Hospitals***

www.i-hd.eu

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Preparation and planning

- Webinar on data quality
- Needs and objectives
- Define the scope and domain(s)
- Scope the data sets
- Partnership and confidentiality agreements, GDPR
- Assign roles, timelines
- Contracts

Dataset generation

- Onsite visit
- Prioritise the DQ dimensions
- Select the EHR variables
- Prioritise and localise the i~HD DQ rule library
- Pre-assessment of the data set extract
- Fine tuning the data set
- Final data set for assessment

Assessment

- Validation of the data set
- Tools based analysis of the variables for agreed dimension
- Graphical outputs + descriptive interpretation by i~HD medical and statistical experts
- Preliminary findings discussed with the hospital, to exchange insights on causative factors

Outcome

- Final written report
- Presentation to the team
- Discussion of recommendations
- Improvement strategy planning
- Workshops, online tutorials
- Issue successful hospitals with a **Data Quality Seal**



Wat makes the i~HD DQS4H unique?

Pragmatic: minimally invasive to hospital operations

Evidence based: well researched, published, assessment methodology

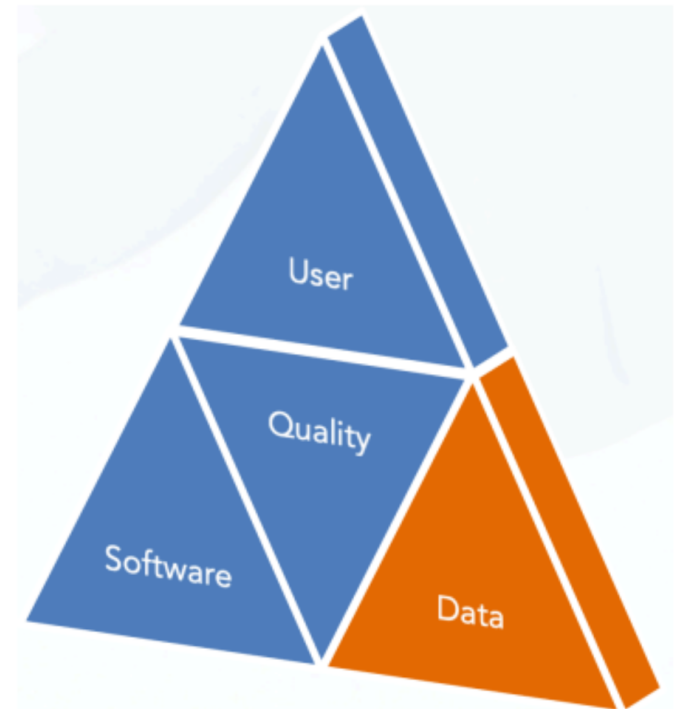
Flexible: can be tailored to your data quality drivers

Focussed: we can help you choose the most suitable dimensions and EHR data variables

Staged: clear sequence of steps with regular interactions and feedback loops

Holistic: we consider quality in the context of your user workflows and your EHR system

Extendable: data sets can be added incrementally, to chart out a data quality improvement journey





Frank Staelens
OLV Hospital Aalst

- [i~HD@iHD_HealthData](#) . Oct 15
- [@Joanxcomella](#), Director of [@VHIR_](#) : *Given #iHD mission in health data quality, it was quite obvious for us to select i~HD as a preferred partner for our hospital data quality strategy. It will make us an even more attractive partner for European research activities." #healthdata*



Miguel Angel Mayer
Hospital del Mar

[i~HD@iHD_HealthData](#) . Oct 9

i~HD member, [@fstaelens](#), OLV Hospital Aalst: My vision is that medical data is an important vehicle to measure and improve quality of care in our hospital. i~HD supports us in analysing and visualising our health data and its quality. This helps in providing high quality care."



Joan Comella
Director of Vall d'Hebron
Institut de Recerca

[Dipak Kalra@DipakKalra](#) . Oct 17

Dr Miguel Angel Mayer, from Hospital del MAR and i~HD Member, explains to the SCOPE audience how integrating clinical data, and assessing its quality, has enabled them to accelerate their bio-informatics research. #Datasaveslives

Thank you!

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