

International collation and sharing of information on data standards, quality & definitions

What can we do together?

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Dec 5, 2008, London

Outline

- International Methodology Consortium for Coded Health Information (IMECCHI)
- Datasets holding
- Data structure/elements
- Validity/comparability
- What can do together?

**International Methodology
Consortium for
Coded Health Information
(IMECCHI)**

**Optimal methodologies for deriving knowledge
and wisdom from health data.**

Correspondence

Open Access

Identifying priorities in methodological research using ICD-9-CM and ICD-10 administrative data: report from an international consortium

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Research Priorities

Some examples...

- International meta-data documentation
- ICD-9 to 10 translation of PSI's
- Implementation and evaluation of diagnosis-type indicators
- Documentation of health system coding standards and training programs
- International development and validation of internal consistency systems for data validity

Second Meeting – 2006

Montreux, Switzerland

- Refinement of research agenda
- Identification of core initiatives
 - Patient safety indicators
 - International meta-data project
- Development of Terms of Reference

Third meeting - Calgary 2008

Theme 1: Administrative Data Features

A compendium of meta-data

International disease coding systems

Validity of administrative data for clinical conditions

Theme 2: Optimization of Administrative Data

Shadow billing and data completeness

Measurement error in diagnostic codes

Interventions for physicians, billing staff and coders

Theme 3: Emerging Methods for EHR Data

Risk adjustment using laboratory and vital sign EHR data

Assessment of symptom collection methods in primary care

Ethical and legal issues in EHR research

International collaboration

- Many advantages
 - Consensus on methodologies
 - Access to regional networks and resources
 - International comparisons
 - Harmonization of strategies
 - Potential bridging of national and international agendas
 - Opportunity for linkages with WHO, OECD

www.IMECCHI.org

For more information, please contact

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Examples of Canadian Data Holdings

Where Can you Get Data in Canada?

- National Organizations
 - Canadian Institute of Health Information
 - Statistics Canada
 - Health Canada
- Provincial Organizations
 - Ministries of health
- Research Institutes/Universities
- Regional Organizations
 - Hospitals, etc.



Canadian Institute of Health Information Data Holdings

- Discharge Abstract Database
- Hospital Morbidity
- Ambulatory Care
- Hospital Mental Health
- Therapeutic Abortions
- Organ Registry
- Trauma Registries
- Joint Replacement Registry
- Continuing Care Reporting System
- Home Care Reporting System
- Rehabilitation Reporting System
- Ontario Mental Health Reporting System
- Medical Imaging Technology
- Prescribed Drug

Provincial Databases

Health insurance registry

Hospital discharges

Emergency/day surgery

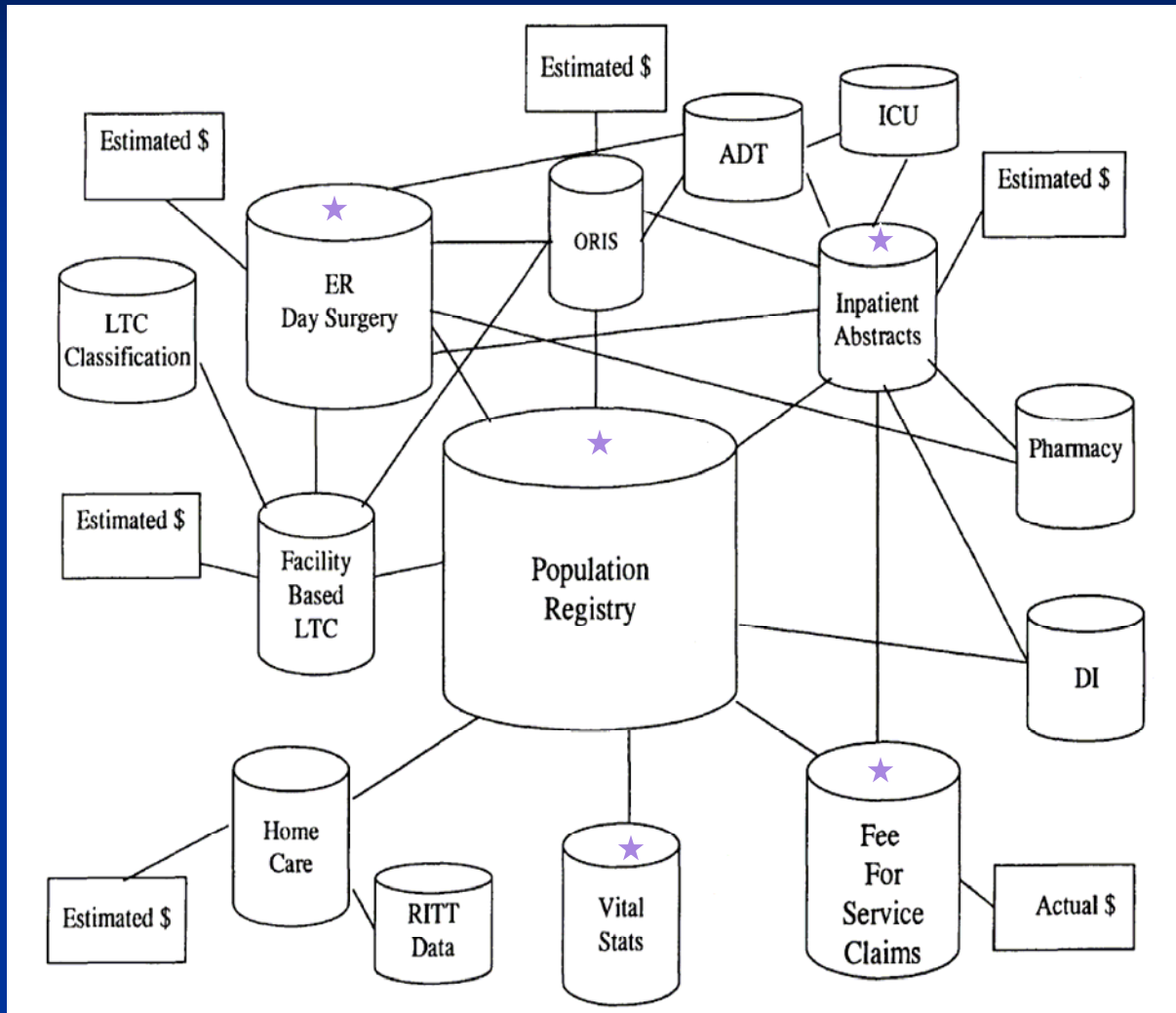
Physician claims

Home care

Long term care

Drugs

Calgary Health Region Data Holdings



ER/Day Surgery = ACCS or Ambulatory Care Classification System

Inpatient Abstracts = DAD or Discharge Abstract Database from CIHI

Fee For Service Claims = PC claims or physician claims

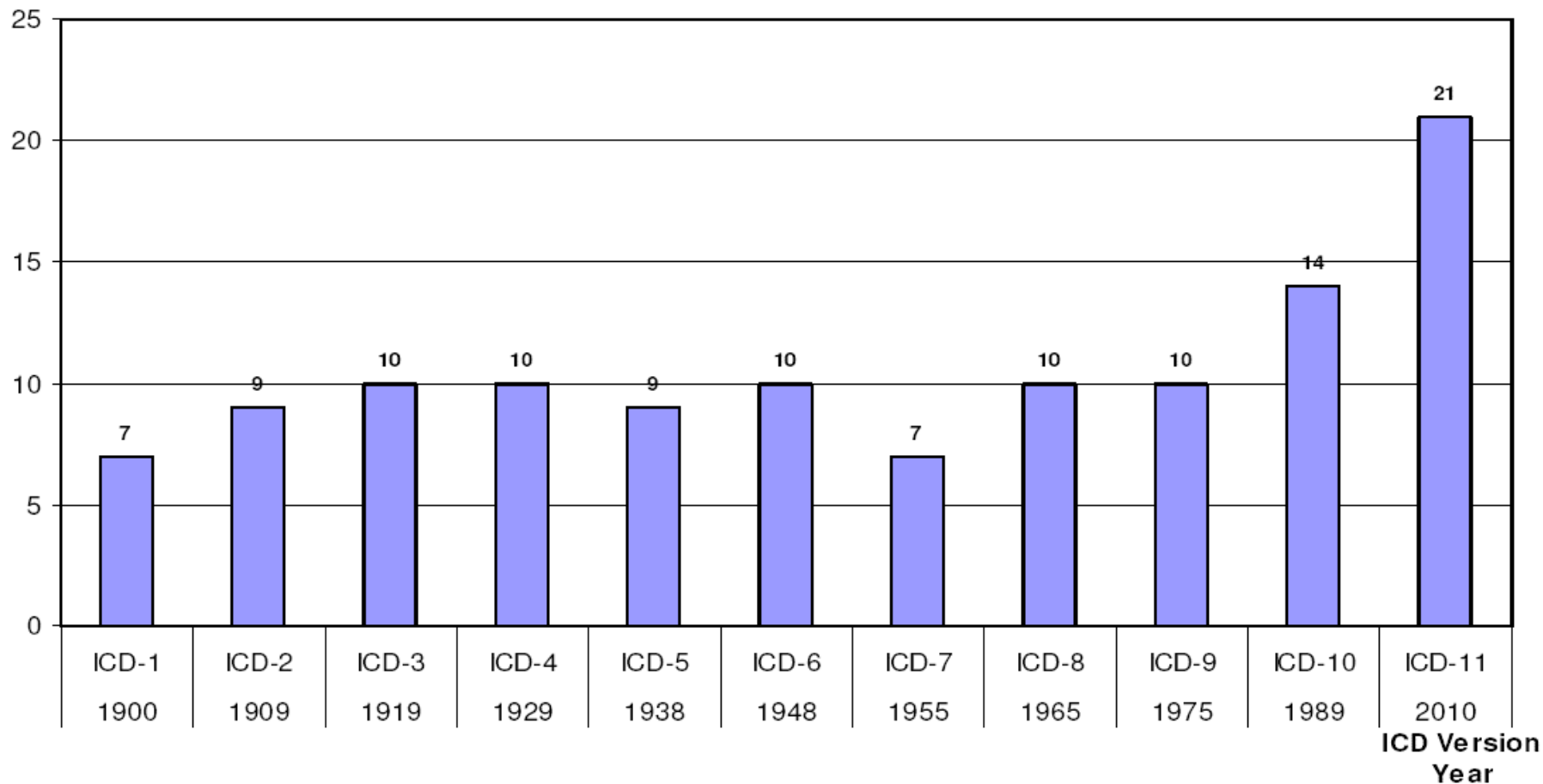
Variables in Health Databases

Hospital Discharge Data

Key Issues

- Coding system
- Number of coding fields for diagnosis and procedure
- Major diagnosis
- Diagnosis timing/type
- Disease grouping system

Interval (years)



Diversity

- USA ICD-9 **CM** and ICD 10 **CM**
- Australia ICD-10 **AM**
- Canada ICD-10 **CA**
- Germany ICD-10 **GM**
- Thailand ICD-10 **TM**
- ...
- Combined ICD *XM*

Comparison

Codes	ICD-10 CA 2006 (Canada)	ICD-10 GM 2007 (Germany)	ICD-10 AM 2006 (Australia)
A4151	Septicemia due to Pseudomonas	Septicemia due to Escherichia coli [E. coli]	Sepsis due to Escherichia coli [E. Coli]
A4152	Septicemia due to Serratia	Septicemia due to Pseudomonas	Sepsis due to Pseudomonas
G8201	Flaccid paraplegia, complete	Flaccid paraparesis and paraplegia, incomplete, acute paresis of non traumatic genesis	Flaccid paraplegia, unspecified, acute
G8202	Flaccid paraplegia, incomplete	Flaccid paraparesis and paraplegia, complete, chronic	Flaccid paraplegia, unspecified, chronic
L891	Decubitus ulcer limited to breakdown of skin (stage 2)	Decubitus ulcer stage 1	Decubitus [pressure] ulcer, stage II
L892	Decubitus ulcer with fat layer exposed (stage 3)	Decubitus ulcer stage 2	Decubitus [pressure] ulcer, stage III
L893	Decubitus ulcer with depth involving muscle (stage 4)	Decubitus ulcer stage 3	Decubitus [pressure] ulcer, stage IV
L894	Decubitus ulcer with depth involving bone (stage 5)	Decubitus ulcer stage 4	
S0620	Diffuse brain injury without loss of consciousness	Diffuse brain and cerebellum injury, unspecified	Diffuse cerebral and cerebellar brain injury, unspecified
S0621	Diffuse brain injury with brief loss of consciousness	Diffuse brain injury	Diffuse cerebral contusions
S0622	Diffuse brain injury with moderate loss of consciousness	Diffuse cerebellum injury	Diffuse cerebellar contusions
S0623	Diffuse brain injury with prolonged loss of consciousness with return to pre-existing level of consciousness	Multiple intracerebral and cerebral hematomas	Multiple intracerebral and cerebellar hematomas

Caution!!

- Common Codes But Different Descriptions

V codes

E codes

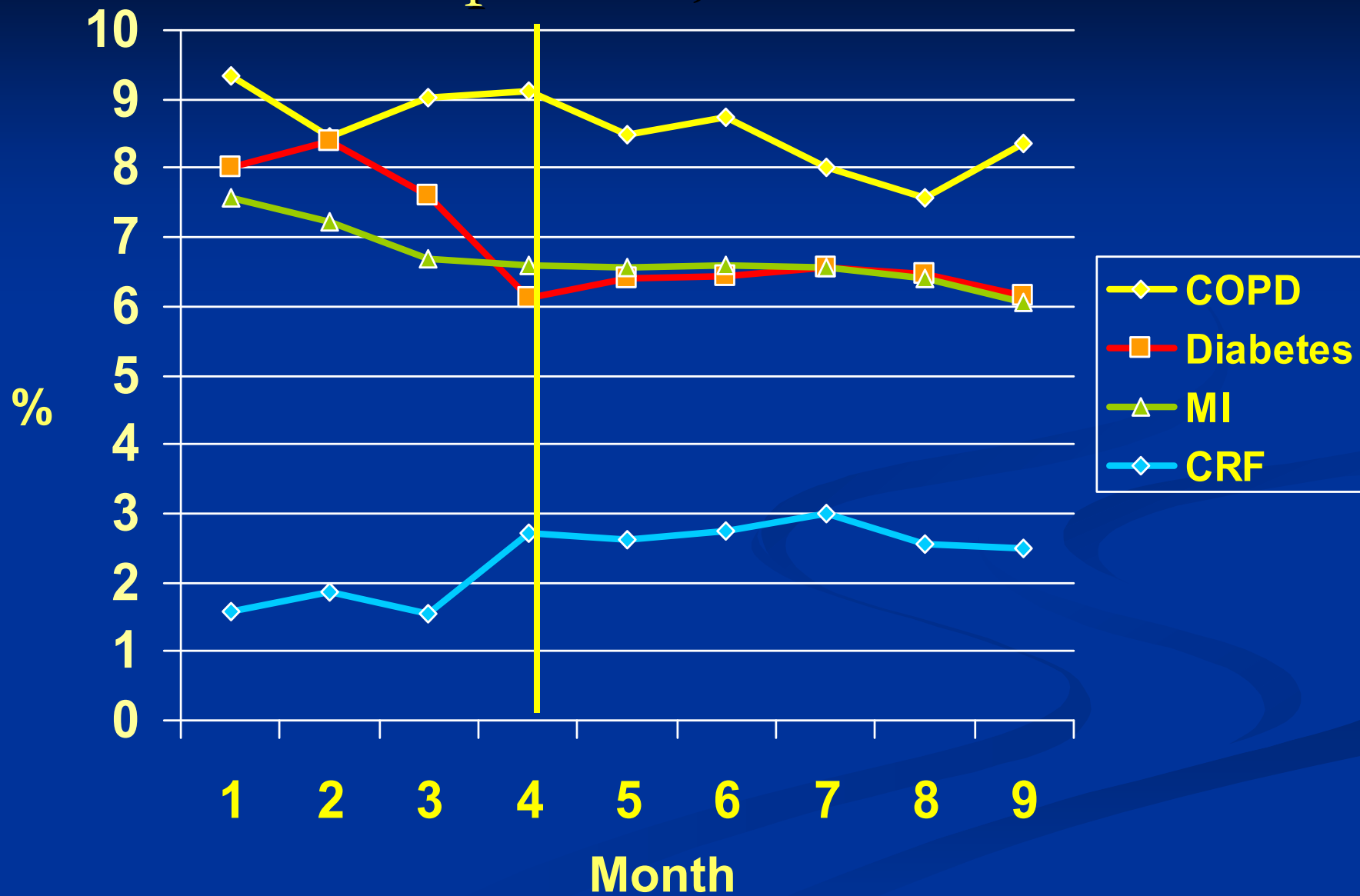
Disease and Procedure Classification

- ICD-9
- ICD-9-CM
- ICD-10-CA

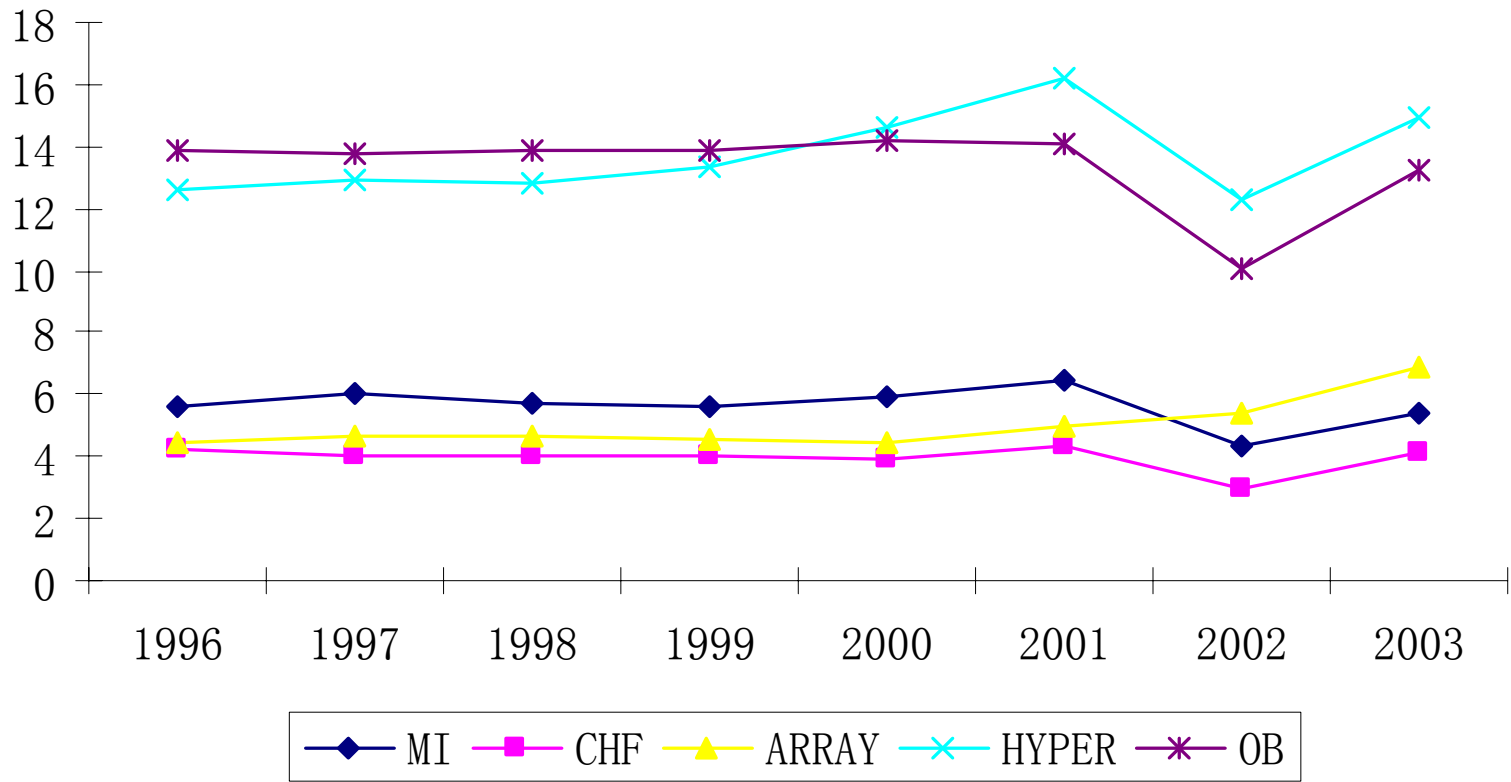
- CCP
- ICD-9-CM
- CCI



% Comorbidity in Calgary Health Region inpatients, 2002



Alberta ICD-9-CM and ICD-10



Coding system for dx and
procedure?

Administrative data

Hospital discharge data

Dx1 Dxtype1, Dx1 Dxtype1... Dx16 Dxtype16

Pr1 Pr1date, Pr2 Pr2date Pr10 Pr10date

Administrative data

Hospital discharge data

Dx1 Dxtype1, Dx1 Dxtype1,,, Dx16 Dxtype16

M

M = Most responsible diagnosis - the condition associated with the longest length of stay or most resource intense.

Administrative data

Hospital discharge data

Dx1 Dxtype1, Dx1 Dxtype1,,, Dx16 Dxtype16
M 1

1 = Pre-Admit comorbidity - A diagnosis or condition that existed pre-admission and satisfies the requirements for determining comorbidity.

Administrative data

Hospital discharge data

Dx1 Dxtype1, Dx1 Dxtype1,,, Dx16 Dxtype16

M

1

2

2 = Post-admit comorbidity - A diagnosis or condition that arises post-admission and satisfies the requirements for determining comorbidity.

Administrative data

Hospital discharge data

Dx1 Dxtype1, Dx1 Dxtype1,,, Dx16 Dxtype16

M

1

2

3

3 = Secondary diagnosis - A diagnosis or condition for which a patient may or may not have received treatment and does not satisfy the requirements for determining comorbidity.

The development, evolution and modifications of ICD-10: challenges to the international comparability of morbidity data

N. Jetté, H. Quan, B. Hemmelgarn, S. Drosler, C. Maass, L. Moskal, W. Paoin, V. Sundararajan, S. Gao, R. Jakob, B. Üstün, W.A. Ghali

for the IMECCHI Investigators

Advances in Patient Safety



Title: Adaptation of AHRQ Patient Safety Indicators for Use in ICD-10 Administrative Data by an International Consortium

Hude Quan, Saskia Drösler, Vijaya Sundararajan, Eugene Wen, Bernard Burnand, Chantal Marie Couris, Patricia Halfon, Jean-Marie Januel, Edward Kelley, Niek Klazinga, Jean-Christophe Luthi, Lori Moskal, Eric Pradat, Patrick S. Romano, Jennie Shephard, Lawrence So, Lalitha Sundaresan, Linda Tournay-Lewis, Béatrice Trombert-Paviot, Greg Webster, William A. Ghali, for the IMECCHI Investigators

ICD-9-CM PSI

PSI 1: Complications of Anesthesia

PSI 3: Decubitus Ulcer

PSI 5: Foreign Body Left During Procedure

PSI 7: Selected Infections Due to Medical Care

PSI 9: Postoperative Haemorrhage or Haematoma

PSI 11: Postoperative Respiratory Failure

PSI 13: Postoperative Sepsis

PSI 15: Accidental Puncture or Laceration

PSI 17: Birth Trauma—Injury to Neonate

PSI 19: Obstetric Trauma—Vaginal Delivery
without Instrument

PSI 2: Death in Low-mortality

PSI 4: Failure to Rescue

PSI 6: Iatrogenic Pneumothorax

PSI 8: Postoperative Hip Fracture

PSI 10: Postoperative Physiologic and
Metabolic Derangement

PSI 12: Postoperative Pulmonary Embolism or
Deep Vein Thrombosis

PSI 14: Postoperative Wound Dehiscence

PSI 16: Transfusion Reaction

PSI 18: Obstetric Trauma—Vaginal Delivery
with Instrument

PSI 20: Obstetric Trauma—Cesarean Delivery



Agency for Healthcare Research and Quality

Advancing Excellence in Health Care

<http://www.ahrq.gov/>

**Need to harmonize AHRQ PSI
administrative data definition
between ICD-9-CM and ICD-10**

**What grouping methods have
been used in your datasets?**

Diagnosis-Related Group (DRG)

DRGs are assigned by a "grouper" program based on ICD diagnoses, procedures, age, sex, and the presence of complications or comorbidities.

- Further grouped into Major Diagnostic Categories (MDCs).
- Developed by 3M
- Classified about 500 groups
- Used as part of the prospective payment system
- Have been used since 1983

MDC Description

0 Ungroupable

1 Nervous System

2 Eye

3 Ear, Nose, Mouth And Throat

4 Respiratory System

5 Circulatory System

6 Digestive System

7 Hepatobiliary System And Pancreas

8 Musculoskeletal System And Connective Tissue

9 Skin, Subcutaneous Tissue And Breast

10 Endocrine, Nutritional And Metabolic System

11 Kidney And Urinary Tract

12 Male Reproductive System

DRG Versions

Medicare DRG

Refined DRGs (RDRG)

All Patient DRGs (APDRG)

Severity DRGs (SDRG)

All Patient Refined DRGs (APRDRG)

International-Refined DRGs (IRDRG)

Case Mix Groups (CMG)

- **Developed by CIHI**
- **To group and describe types of inpatients discharged from acute-care hospitals.**
- **Modeled after the (DRG's) based on four criteria:**
 1. **Patient groups had to make good clinical sense.**
 2. **They had to be based on routinely collected data.**
 3. **There had to be a manageable number of groups.**
 4. **They had to be statistically homogeneous with respect to length of hospital stay.**

Clinical Risk Groups Classification System (CRG)

**Classify diagnoses into episode diagnostic categories
(EDCs, n = 534)**



**Major diagnostic categories
(MDCs, n = 37)**



Select primary chronic disease (PCD)



Assign severity level to PCD



Assign Core Health Status Group

Clinical Risk Groups Classification System (CRG)

Core Health Status Groups:

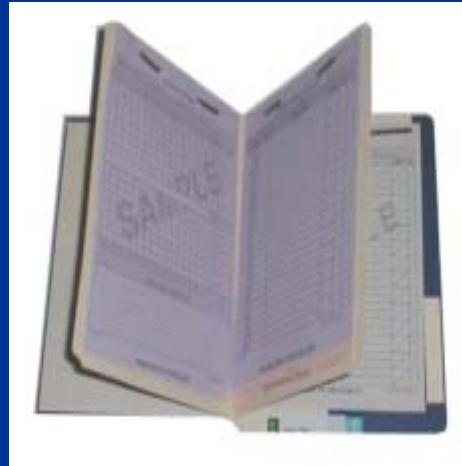
- 1. Healthy**
- 2. Significant acute diagnosis**
- 3. Single minor chronic**
- 4. Multiple minor chronic pairs**
- 5. Single dominant or moderate chronic**
- 6. Multiple significant chronic pairs**
- 7. Chronic triplets**
- 8. Dominant / metastasized / recurrent malignancies**
- 9. Catastrophic**

Validity and Comparability

Key Issues

- ‘Gold Standard’
- ICD coding definition for conditions

Hospital discharge data



Sources of Bias in Administrative Data

Disease Asymptomatic



**Health knowledge
Regular physical check**



Disease Symptomatic



**Access to services
-Predisposing factors
-Enabling factors
-Need factors**



Encounter with physicians

Sources of Bias in Administrative Data

Encounter with physicians



**Examination
History
Diagnostics**



Diagnosis



Interventions

Patient

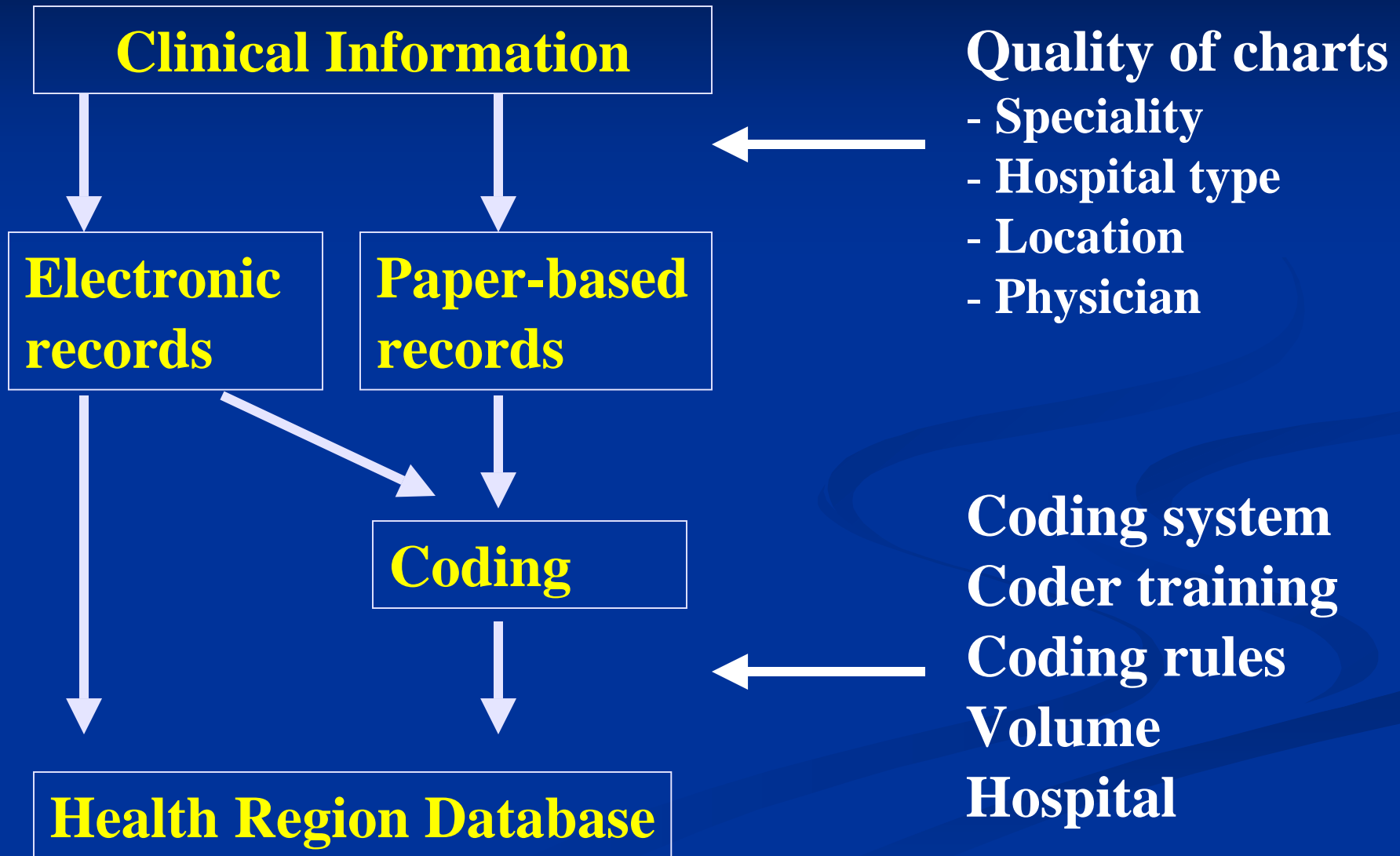
- Gender
- Urban/Rural, distance
- Race/ ethnicity
- Language

Physician

- Knowledge
- Experience
- Practice initiatives
- Attitudes on guidelines
- Follow-up
- Continuity of care



Sources of Bias in Administrative Data



Sources of Bias in Administrative Data

Health Region Database



Provincial Data



National Database



**Management
Cleaning data**



**Coding systems
ICD-9, ICD-9-CM,
ICD-10-CA, CCP,
CCI**

Concepts

Sensitivity

Specificity

Negative Predictive value

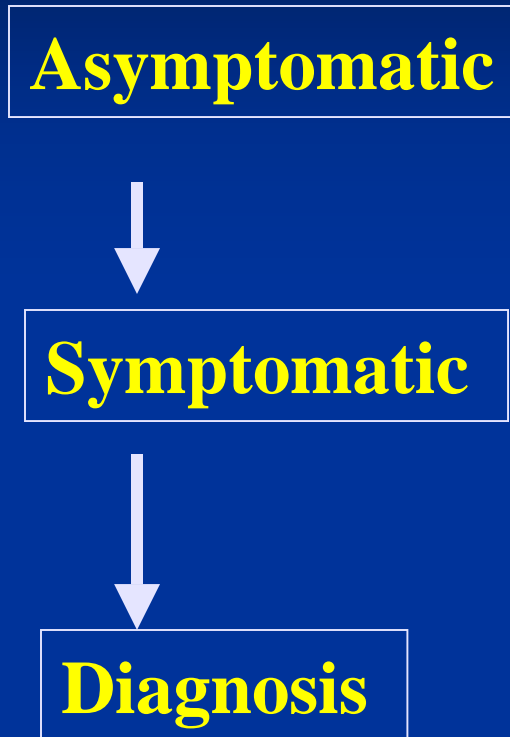
Positive predictive value

Kappa

		Gold Standard		
		Yes	No	
Adm data	Yes	a	b	a+b
	No	c	d	c+d
		a+c	b+d	N

Sensitivity, specificity, positive predictive value
 Negative predictive value, kappa

“Gold standard”



- **Diagnostics**
- **Survey – self-report**
- **Physical Measurement**
- **Disease registry**
- **Chart review**
- **Laboratory data**

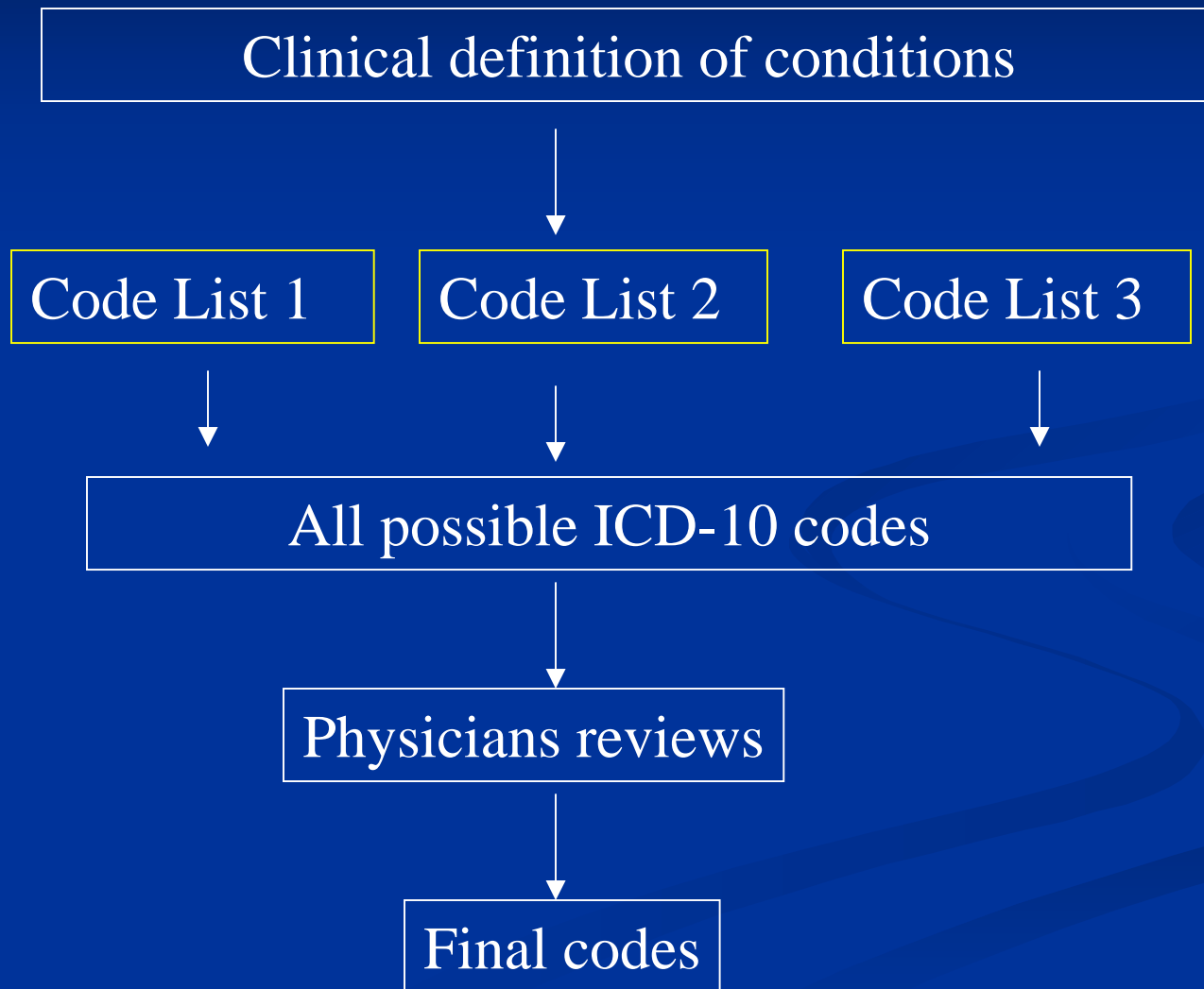
Defining Conditions

- **Documentations**
- **Diagnosis criteria**

Development ICD-9/10 coding definitions for conditions

Code	Definition
ICD 9 CM	
410	Acute myocardial infarction
411	Other acute and subacute forms of ischemic heart disease
412	Old myocardial infarction
413	Angina pectoris
414	Other forms of chronic ischemic heart disease
427	Cardiac dysrhythmias
428	Heart failure
429	Ill-defined descriptions and complications of heart disease
518	Other diseases of lung
780	General symptoms
780.2	Syncope and collaps
785	Symptoms involving cardiovascular system
785.51	Cardiogenic shock
786	Symptoms involving respiratory system and other chest symptoms
786.5	Chest pain
ICD 10 CA	
I20	Angina pectoris
I21	Acute myocardial infarction
I22	Subsequent myocardial infarction
I23	Certain current complications following acute myocardial infarction
I24	Other acute ischemic heart diseases
I25	Chronic ischemic heart disease
I25.2	Old myocardial infarction
I46	Cardiac arrest
I51	Complications and ill-defined descriptions of heart disease
I51.3	Intracardiac thrombosis, not elsewhere classified

ICD Definition Development



Coding Algorithms for Defining Comorbidities in ICD-9-CM and ICD-10 Administrative Data

Hude Quan, MD, PhD,† Vijaya Sankararajan, MD, MPH, FACP,‡ Patricia Halfon, MD,§ Andrew Fong, BCOMM,* Bernard Burnand, MD, MPH,§ Jean-Christophe Luthi, MD, PhD,§ L. Duncan Saunders, MBBCh, PhD,¶ Cynthia A. Beck, MD, MAsc,*|| Thomas E. Feasby, MD,** and William A. Ghali, MD, MPH,*†,††*

Objectives: Implementation of the International Statistical Classification of Disease and Related Health Problems, 10th Revision (ICD-10) coding system presents challenges for using administrative data. Recognizing this, we conducted a multistep process to develop ICD-10 coding algorithms to define Charlson and Elixhauser comorbidities in administrative data and assess the performance of the resulting algorithms.

Methods: ICD-10 coding algorithms were developed by ‘translation’ of the ICD-9-CM codes constituting Deyo’s (for Charlson comorbidities) and Elixhauser’s coding algorithms and by physicians’ assessment of the face-validity of selected ICD-10 codes. The process of carefully developing ICD-10 algorithms also produced modified and enhanced ICD-9-CM coding algorithms for the Char-

Results: Among 56,585 patients in the ICD-9-CM data and 58,805 patients in the ICD-10 data, frequencies of the 17 Charlson comorbidities and the 30 Elixhauser comorbidities remained generally similar across algorithms. The new ICD-10 and enhanced ICD-9-CM coding algorithms either matched or outperformed the original Deyo and Elixhauser ICD-9-CM coding algorithms in predicting in-hospital mortality. The C-statistic was 0.842 for Deyo’s ICD-9-CM coding algorithm, 0.860 for the ICD-10 coding algorithm, and 0.859 for the enhanced ICD-9-CM coding algorithm, 0.868 for the original Elixhauser ICD-9-CM coding algorithm, 0.870 for the ICD-10 coding algorithm and 0.878 for the enhanced ICD-9-CM coding algorithm.

Conclusions: These newly developed ICD-10 and ICD-9-CM comorbidity coding algorithms produce similar estimates of comor-

Mortality in 4 country data

Score	Canada	Switzerland	Australia	Japan
0	0.36	0.51	0.37	1.53
1	2.43	2.83	2.81	2.60
2	4.24	3.09	5.80	3.75
3	8.51	5.68	8.24	6.25
4	10.26	7.57	11.05	6.55
5	16.54	12.30	15.96	8.13
6+	13.70	11.40	16.43	16.76
C- statistics	0.836	0.772	0.840	0.701

Cross-National Comparative Performance of Three Versions of the ICD-10 Charlson Index

Vijaya Sundararajan, MD, MPH,*† Hude Quan, MD, PhD,‡§ Patricia Halfon, MD,¶
Kiyohide Fushimi, MD, PhD,|| Jean-Christophe Luthi, MD, PhD,¶ Bernard Burnand, MD, MPH,¶ and
William A. Ghali, MD, MPH,‡§** for the International Methodology Consortium for Coded Health
Information (IMECCHI)

Objective: The Charlson comorbidity index has been widely used for risk adjustment in outcome studies using administrative health data. Recently, 3 International Statistical Classification of Diseases, Tenth Revision (ICD-10) translations have been published for the Charlson comorbidities. This study was conducted to compare the predictive performance of these versions (the Halfon, Sundararajan, and Quan versions) of the ICD-10 coding algorithms using data from 4 countries.

Methods: Data from Australia (N = 2000–2001, max 25 diagnosis codes), Canada (N = 2002–2003, max 16 diagnosis codes), Switzerland (N = 1999–2001, unlimited number of diagnosis codes), and Japan (N = 2003, max 11 diagnosis codes) were analyzed. Only the first admission for patients age 18 years and older, with a length of stay of ≥ 2 days was included. For each algorithm, 2 logistic regression models were fitted with hospital mortality as the outcome and the Charlson individual comorbidities or the Charlson index score as independent variables. The c-statistic (representing the area under the receiver operating characteristic curve) and its 95% probability bootstrap distribution were employed to evaluate model performance.

Results: Overall, within each population's data, the distribution of comorbidity level categories was similar across the 3 translations. The Quan version produced slightly higher median c-statistics than the Halfon or Sundararajan versions in all datasets. For example, in

Japanese data, the median c-statistics were 0.712 (Quan), 0.709 (Sundararajan), and 0.694 (Halfon) using individual comorbidity coefficients. In general, the probability distributions between the Quan and the Sundararajan versions overlapped, whereas those between the Quan and the Halfon version did not.

Conclusions: Our analyses show that all of the ICD-10 versions of the Charlson algorithm performed satisfactorily (c-statistics 0.70–0.86), with the Quan version showing a trend toward outperforming the other versions in all data sets.

(*Med Care* 2007;45: 1210–1215)

Risk adjustment is a vital component in health services utilization and outcome studies. The Charlson Comorbidity Index, well-validated in many international studies, has facilitated this process.¹ Several ICD-9-CM coding algorithms of the Charlson index were developed for use with administrative hospital discharge abstract data.^{2–4} After the recent implementation of International Statistical Classification of Diseases, Tenth Revision (ICD-10) 3 ICD-10 translations of the Charlson index were formulated in Switzerland (Halfon version),⁵ Australia (Sundararajan version),⁶ and Canada (Quan version)⁷ (Table 1). Of these 3 translations, the Halfon and Sundararajan versions were developed independently of one another, using different approaches: clinical judgment about the original Charlson definition of comorbidities in combination with the Deyo ICD-9-CM codes in the Halfon version, and a computerized mapping file to translate the Deyo ICD-9-CM to ICD-10-AM followed by coder and clinical review for the Sundararajan version (Table 1). In comparison, the Quan version combined the strengths of the Halfon and Sundararajan versions and, to further improve on this base, incorporated an intensive clinician and professional coder panel review.⁷

With 3 versions of the Charlson comorbidity algorithm and the country-specific nature of ICD-10 codes, the question arises of how these algorithms perform when applied to administrative data routinely collected in various countries. No empirical evidence is available to address this question. Thus, we aimed to compare the predictive ability of the Halfon, Sundararajan, and Quan versions of the ICD-10

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Dr. Quan is supported by a Population Health Investigator Award from the Alberta Heritage Foundation for Medical Research, Edmonton, Alberta, Canada and by a New Investigator Award from the Canadian Institutes of Health Research.

Dr. Ghali is supported by a Health Scholar Award from the Alberta Heritage Foundation for Medical Research, Alberta, Canada, and by a Government of Canada Research Chair in Health Services Research.

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ISSN: 0025-7079/07/4512-1210

4000 Records
July – Dec 2003



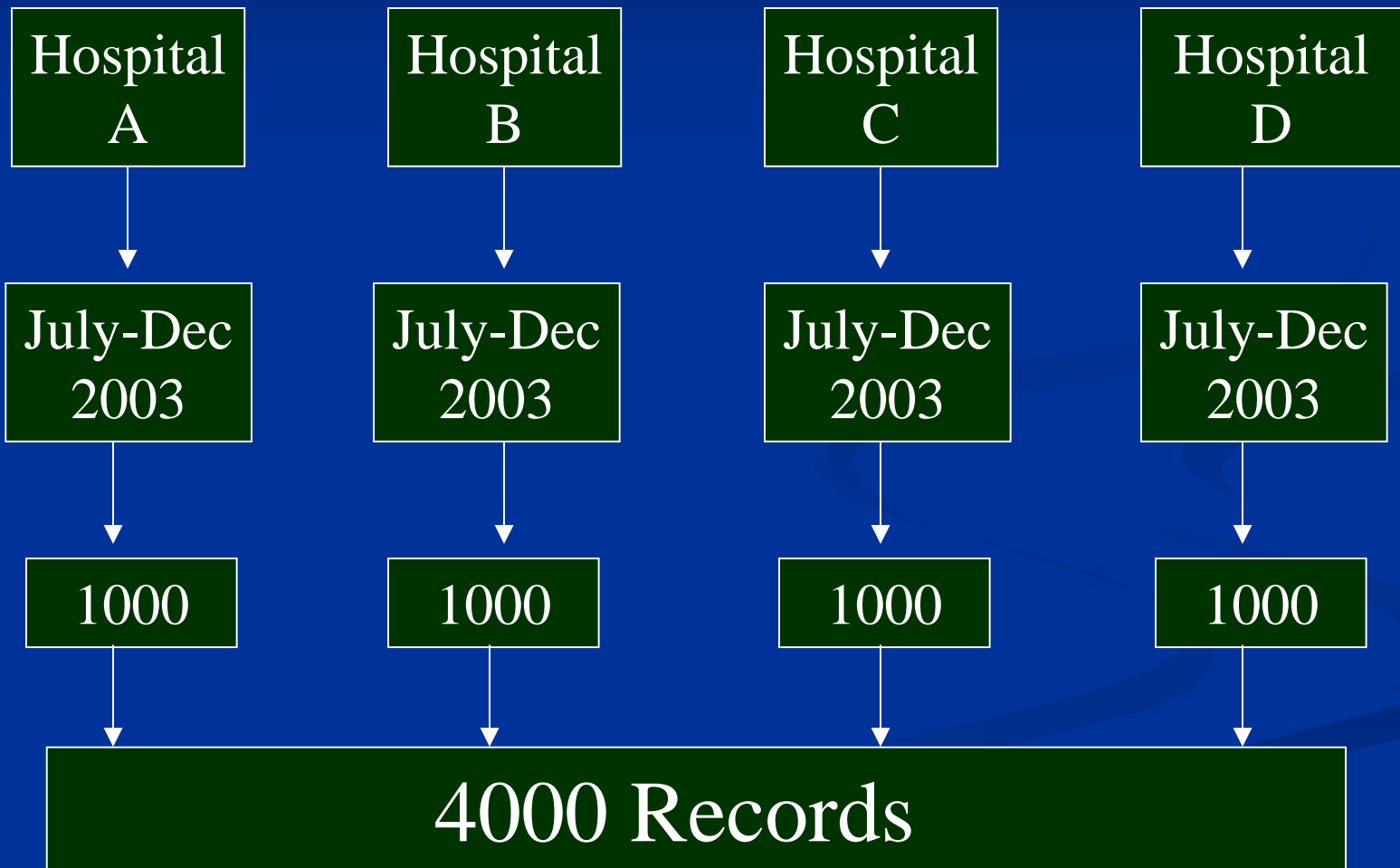
Data
ICD-10-CA

Data
Chart
Review

Data
ICD-9-CM

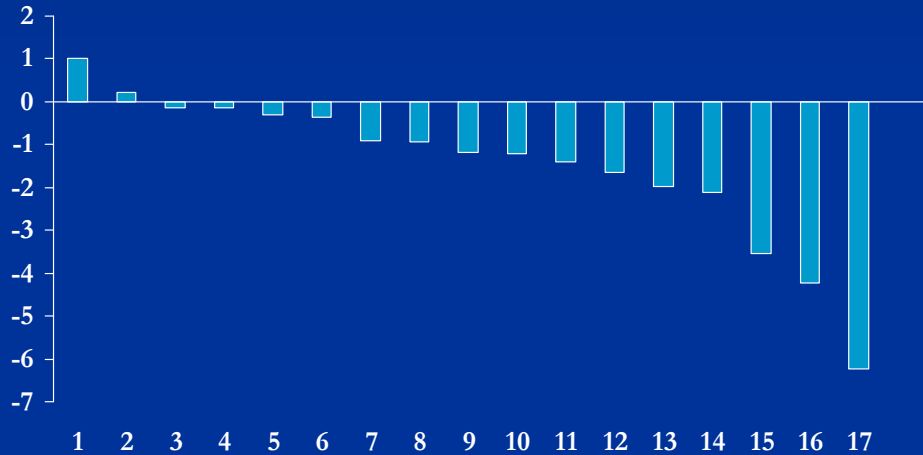


Alberta Teaching Hospitals

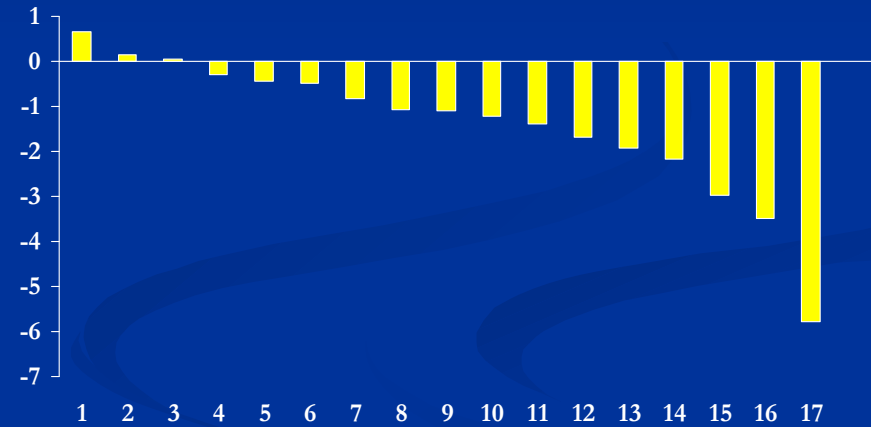


Difference in prevalence (%)

% in ICD 10 data - % in charts



% in ICD 9 CM data - % in charts



Kappa

● ICD-9-CM vs. Chart

□ ICD-10 vs. Chart

1.00

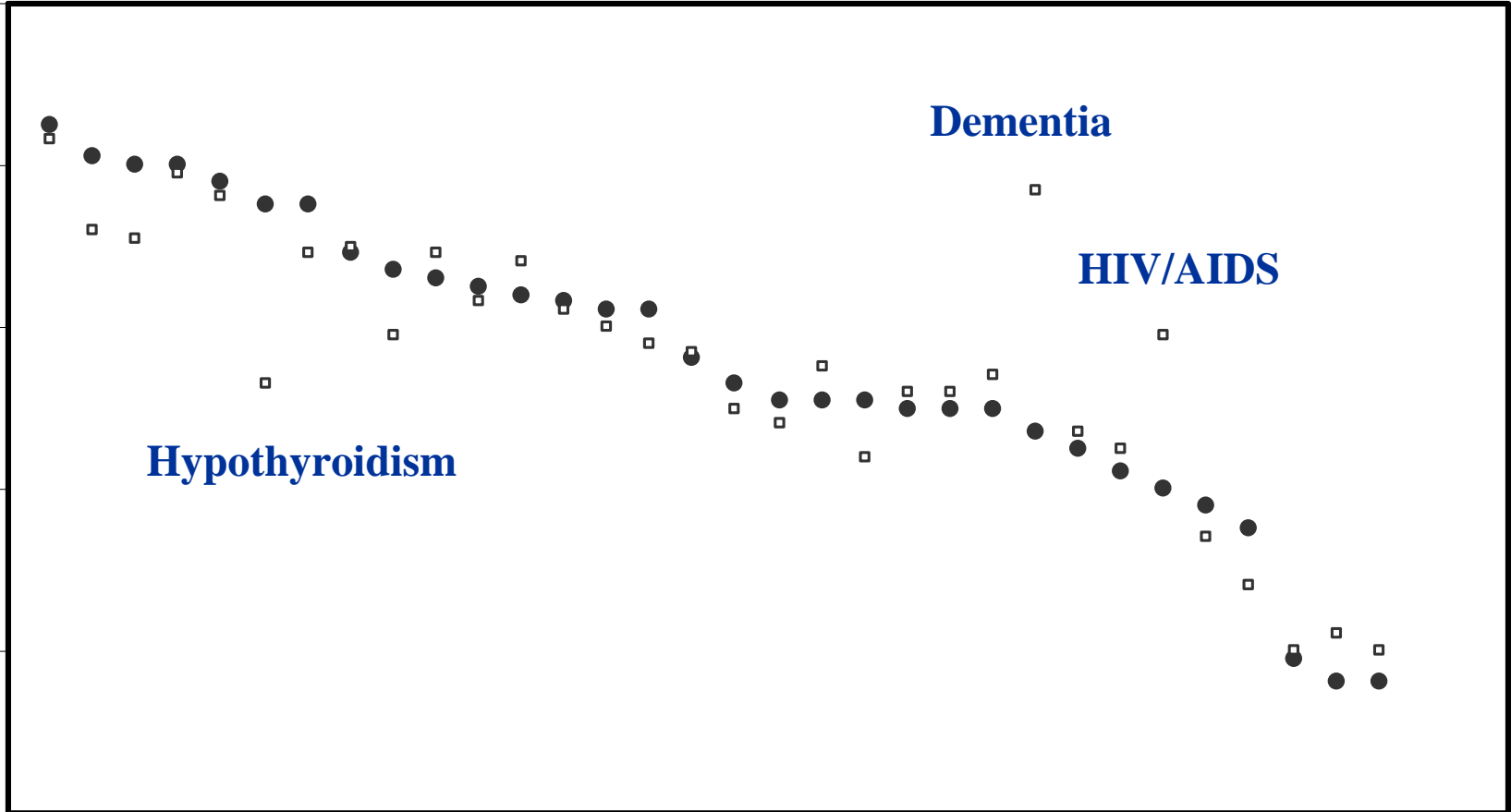
0.80

0.60

0.40

0.20

0.00



32 Conditions

Kappa= 0-0.20 (Poor)

Barium radiograph (0.08)

Computerized tomography scan (0.01)

Insertion of nasogastric tube (0.12)

Tracheal intubation (0.04)

Ultrasound (0.03)

Kappa 0.81 and over (near perfect)

Bone marrow aspirate (0.93)

Hemo or peritoneal dialysis (0.83)

Lumbar puncture (0.81)

Mechanical ventilation (0.9)

Percutaneous abdominal drainage (0.85)

Thoracentesis (0.85)

Table 29: Estimates of agreement, sensitivity, specificity, and predictive values for hypertension algorithms

# Years		Algorithm	κ	Sens. (%)	Spec. (%)	Youden	PPV (%)	NPV (%)
1	1	1+ P	0.65	68.4	94.7	0.63	74.6	93.0
	2	2+ P	0.54	48.4	97.5	0.46	81.4	89.3
	3	1+ H or 1+ P	0.66	70.1	94.4	0.65	73.9	93.3
	4	1+ H or 2+ P	0.56	51.1	97.1	0.48	79.9	89.7
	5	1+ H or 1+ P or 1+ Rx	0.68	89.9	88.7	0.79	64.4	97.5
	6	1+ H or 1+ P or 2+ Rx	0.70	89.0	89.9	0.79	66.6	97.3
2	7	1+ P	0.67	79.4	91.9	0.71	69.1	95.2
	8	2+ P	0.66	66.3	95.6	0.62	77.5	92.6
	9	1+ H or 1+ P	0.68	81.2	91.6	0.73	68.8	95.6
	10	1+ H or 2+ P	0.67	69.4	95.2	0.65	76.8	93.2
	11	1+ H or 1+ P or 1+ Rx	0.64	91.9	86.0	0.78	59.8	97.9
	12	1+H or 1+ P or 2+ Rx	0.66	91.2	87.3	0.78	62.0	97.8
3	13	1+ P	0.67	83.2	90.3	0.74	66.0	95.9
	14	2+ P	0.68	72.4	94.8	0.68	76.0	93.8
	15	1+ H or 1+ P	0.67	84.9	89.9	0.75	65.7	96.3
	16	1+ H or 2+ P	0.70	75.6	94.4	0.71	75.2	94.5
	17	1+ H or 1+ P or 1+ Rx	0.62	92.8	84.0	0.77	56.8	98.1
	18	1+H or 1+ P or 2+ Rx	0.64	92.2	85.7	0.78	59.4	98.0

Hypertension Definition

(2 claims or 1 hospitalization within 2 years)

Chart review and administrative data

Sensitivity: 75%

Specificity: 94%

PPV: 81%

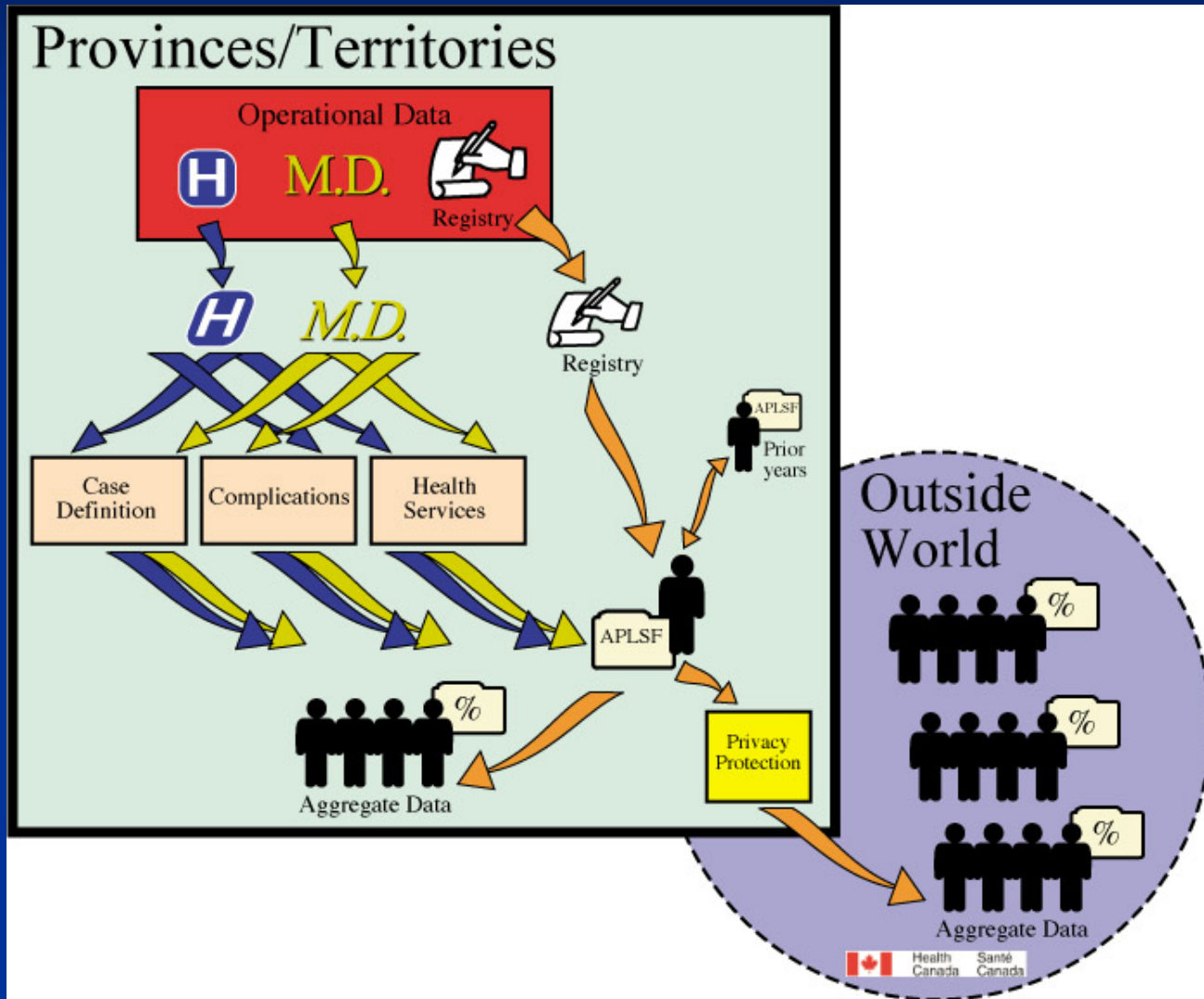
NPV: 92%

Kappa: 0.71

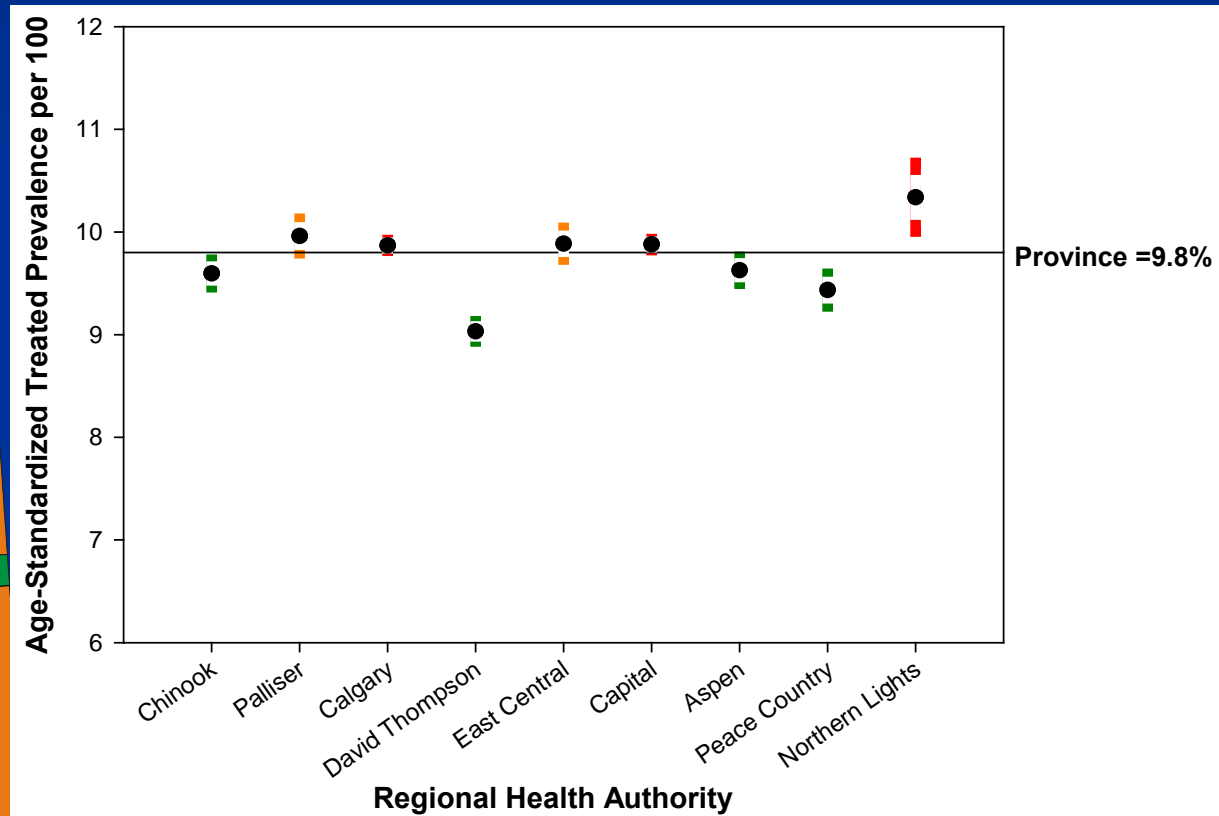
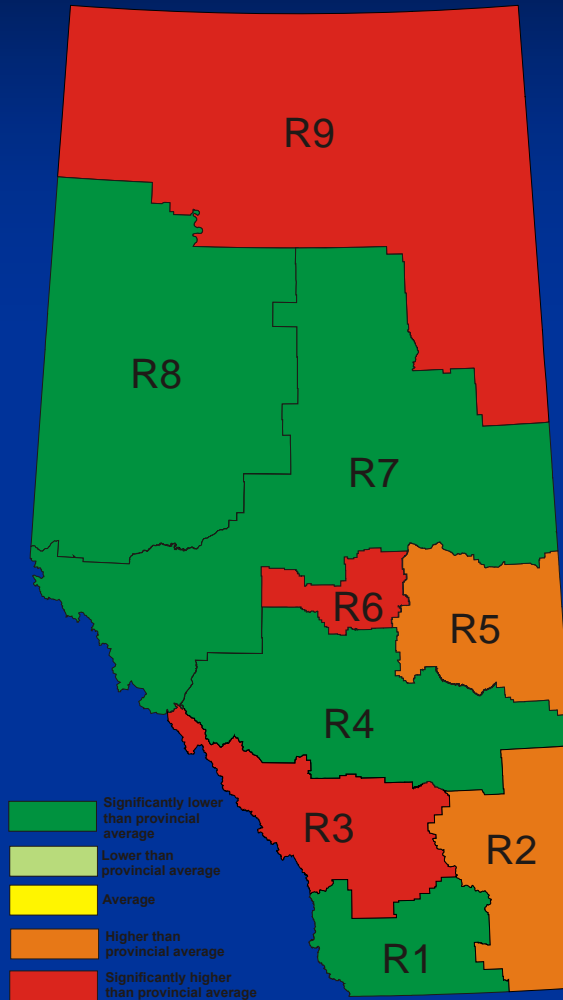
Diabetes Case Definition

Method	Sensitivity	PPV
1 claim or 1 hospital (2 yr)	0.91	0.61
2 claim or 1 hospital (2 yr)	0.86	0.80

Data Extraction and Analysis



Regional differences in the treated prevalence of hypertension, Alberta 2003



Validated Conditions

AMI

Hypertension

Diabetes

Epilepsy

Chronic renal disease

Charlson and Elixhauser comorbidities

AHRQ patient safety indicators

What can we do for structure, quality and definition ?

- Documentations
- Internal logical check of data
- Statistical methods
- Standardizing coding algorithms for conditions (Concept Dictionary)
- Training