Social Determinants and EHR Data: Analytic Decision Support

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The PaTH Clinical Data Research Network





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Patrick Ryan, Observational Health Data Sciences and Informatics (OHDSI) Overview, 5/14/14

Pioglitazone a	and I BMJ	
cancer: a pro	pens BMJ 2012;344:e3645 doi: 10.1136/bmj.e3645 (Published 31 May 2012)	Page 1 of 1
matched coh	ort s	
Li Wei, Thomas M. MacDonald Medicines Monitoring Unit (MEMO), Division of Medical School, Dundee, UK	BJCP May 2012: "In this study population, pioglitazone does not appear to be significantly associated with an increased risk of bladder cancer in patients with type 2 diabetes."	RESEARCH
WHAT IS ALREADY KNOW THIS SUBJECT	NABOUT The use of pioglitazone and the risl	of bladder cance
 Pioglitazone is mainly used in a with diet and exercise and oth anti-diabetic medications to tra- diabetes mellitus. Long term use of pioglitazone 	BMJ May 2012: "The use of pioglitazone is associated with an increased risk of incident bladder cancer among people with type 2	ted case-control
of therapy) may be associated increased risk of bladder cance	diabetes."	tian B Filion <i>assistant professor</i> ¹³ princlogist ⁴ . Michael N Pollak
WHAT THIS STUDY ADDS	oncologist and professor ² , Samy Suissa professor ⁵	te-Catherine, H-425.1, Montreal, Quebec, Canada

Patrick Ryan, Observational Health Data Sciences and Informatics (OHDSI) Overview, 5/14/14



Courtesy Kelly Gleason

Challenge

How do I convince hardboiled researchers that our results are as trustworthy and believable as the best epidemiological data? Dan Ford



DHS

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Where's the Population?

Sen A, et al. GIST 2.0: A scalable multitrait metric for quantifying population representativeness of individual clinical studies. J Biomed Inform. 2016 Oct;63:325-336.



What's the "diagnosis"? The case atrial fibrillation





Some Potential Biases



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Amateur Analysts

 Too many analysts to train them all at the level we want

► MACRA, eCQM, Pop Health, PMI, ...

- Analyses are the most complicated
- No funds for proper statistical analysis
- Statistical-analytic decision support is needed
- We need to convert methodological knowledge into computer-readable form



3. Amateur analysts



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Workforce

BHEF Issue Brief . 2014 http://files.eric.ed.gov/fulltext/ED55964 o.pdf



3. Amateur analysts

According to the McKinsey report, the United States will need an additional 140,000 to 190,000 data science experts with "deep analytical skills," plus 1.5 million managers capable of using data analytics in decision making.



Decision Support Cycle



Knowledge engineering

Knowledge use



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Decision Support Cycle



Knowledge engineering

Knowledge use



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Intelligent Assistance and Data Analysis

By 1995 or so, the largest single driving force in guiding general work on data analysis and statistics [will be] to understand and improve data-analytic expert systems..." ➤ John Tukey, 1986



Early History

- 1983: Nedler: Front-end system (for GLIM)
- 1984: Gale, Pregiborn: REX: Advise on linear regression
- 1985: Hahn defines levels of intelligence: simple computerized answering→automated statistical consulting
- 1988: Duijsens: PRINCE helps naïve users formulate analysis options
- 1988: Oldford & Peters: DINDE: graphical environment tracks steps
- 1989: Chowdury: MAXITAB for inexperienced users for data analysis and interpretation
- 1994: Silvers et al.: PROPHET: Beyond Anova
- Silvers, 1994

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Knowledge Cycle





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Desiderata for Computable Biomedical Knowledge for Learning Health Systems

Lehmann HP, Downs SM. Desiderata for Computable Biomedical Knowledge for Learning Health Systems. Learn Heal Syst. 2018;e10065:1–9.

Desiderata	Development Work to Be Done
1. Discrimination	 Measures that take clinical thresholds into account^{70,71} Elicitation and articulation of those thresholds Methods for recalculating local discrimination
2. Local Recalibration	 Application of calibration based on thresholds¹⁷
3. Thresholds & Local Preferences	Elicitation, articulation of preferencesLocal calculation of thresholds
4. Explanation	Deployment
5. Monitoring	 Choose variables based on value of information⁷²
6. Debiasing	 Creation and curation of debiasing models Application of debiasing models
7. Generalizability	 Calculation of distance⁶² Adding to the Knowledge Artifact the meta data required to choose the calculation
8. Semantic Uncertainty	 Derivation of the epistemic confidence interval
9. Findable	 Articulation of the full ontology required to index a Knowledge Artifact at all its multiple levels Tagging KO with that ontology
10. Other Commandments as necessary and proper	 Continuous monitoring and improvement of these desiderata

Ontology for Biases: Extensions to OCRe

H Lehmann, T Darden, G Williams. 2014. Unpublished.



To Do

- Methodology for the analysts
- Knowledge tools to store the knowledge
- Knowledge tools to apply the knowledge
- Combine JH/PaTH/Israeli expertise





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