



FP7-ICT-2013-10
ICT-WP-2013.5.1

Personalized health, active ageing, and independent living



Capturing Scientific Knowledge of Medical Risk Factors

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1st International Workshop on Capturing Scientific Knowledge
Palisades, New York, 7th October 2015



What are we doing & why?

- A model of evidence-based medical risk
- Clinical research identifies *risk factors*
 - ↪ “If you smoke, your risk of X is increased by...”
- Used to advise patients on
 - ↪ Lifestyle changes
 - ↪ Risk mitigation
 - ↪ Chronic disease management

Existing risk prediction tools

- Hardcoded statistical models
 - ↪ Input age, height, weight, some lifestyle factors...
 - ↪ Output a single probability & some general recommendations
- Less informative
- Not extensible
 - ↪ New science: new model/tool



- **Cardiorenal patients**
 - ↳ Elevated risks of comorbidities

- **Risk calculation & decision support**
 - ↳ Based on at-home monitoring

- **Scientific contribution:**
 - ↳ *Generalised & extensible* risk model
 - ↳ Capable of fine-grained & hypothetical reasoning

A sample risk factor association

- Diabetes causes ischemic heart disease

- ↪ Diabetic and male: 2.82 x more likely to develop IHD

- Confidence interval 2.35 – 3.38

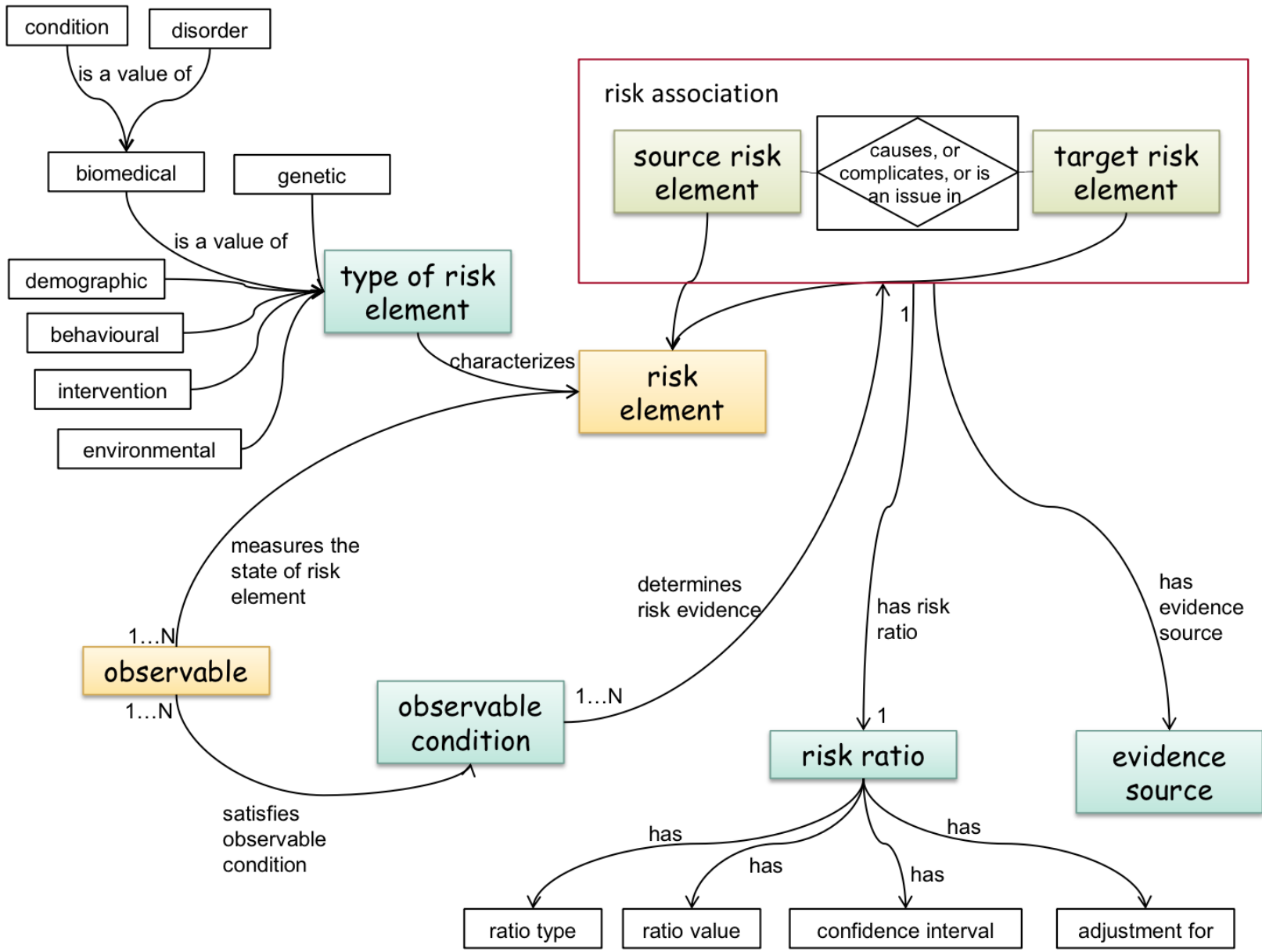
- Source: Pubmed ID 24859435

- ↪ Diabetic and female: 2.16 x more likely to develop IHD

- Confidence interval 1.82 – 2.56

- Source: Pubmed ID 24859435

- Combine en masse for extensible risk calculation



Types of risk factor

- Environmental
- Demographic
- Genetic
- Behavioural
- Biomedical

Observables

- How do we know when a risk factor applies to a patient?
 - ↪ Who fits the population criteria in the evidence?
- Observations of a patient
 - ↪ Manual or automatic
- Logical expression describing study population
 - ↪ E.g., “sex = “female” and diabetes = “diagnosed””

Capturing risk factors

- Clinician-defined literature search methodology
- Identify quantified risk factors & relevant population
- Custom (Drupal) forms reflecting model
 - ↳ Automatically convert into RDF and store
- Review
- Repeat..

How did it go?

- 93 risk factor associations
- Based on 45 different risk elements
- [Pause for valiant attempt at a demo]

How did it go?

- Relatively straightforward
- Identifying risk factors sometimes tricky
- Familiar language & natural model for clinicians
- Inconsistencies in clinical writing (“risk factor”)
- Biggest issue:
 - ↳ Logical expressions for grounding in observables

Observable expressions

- “Diagnosed AND/OR between 8% and 9%”
- In clinical literature but not easily made machine-readable
- OR not in clinical literature
 - ↳ Tacit knowledge in clinical process
- “?diabetes = “diagnosed” || (?HbA1c > 8 && ?HbA1c < 9)”

Lessons & future work

- “Hidden” clinical knowledge
- Standard clinical terminology hides some generalities
 - ↳ “Positive” risk factors
- Ongoing work on decision support/visualisation
- Easy to generalise to other areas of medicine
 - ↳ Outside medicine?
- Curation

acknowledgment

work funded under project CARRE: Personalized patient empowerment and shared decision support for cardiorenal disease and comorbidities

co-funded by the
European Commission under the
Information and Communication Technologies (ICT)
7th Framework Programme
Contract No. FP7-ICT-2013-611140

CARRE <http://www.carre-project.eu/>



cite as

A. Third, E. Kaldoudi, G. Gotsis, S. Roumeliotis, K. Pafili, J. Domingue, Capturing Scientific Knowledge on Medical Risk Factors, K-CAP2015: 8th International Conference on Knowledge Capture, ACM, Palisades, NY, USA, Oct. 7-10, 2015

<http://www.isi.edu/ikcap/sciknow2015/#papers>

