

Estimating Neighbourhood-Level Behavioural Risk Factor Prevalence from Large Population-Based Surveys: A Bayesian Approach

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"Estimating micro area behavioural risk factor prevalence from large population-based surveys: a full Bayesian approach"

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Outline

- Background
- Rationale & Objectives
- Study Area
- Methods
- Results
- Strengths & Limitations
- Conclusions
- Acknowledgements



Background

- Meeting local health needs a key objective of public health units (PHUs)
- Localized risk factor prevalence estimates:
 - Provide information for targeted public health programs
 - ii. Inform neighbourhood-level models of related chronic diseases



Background

- Complex surveys provide high quality data at a regional level, but not neighbourhood level
- Small numerator and denominators are a challenge in neighbourhood-level analyses
- Bayesian methods can overcome these challenges





Study Rationale & Objectives



Study Rationale

To generate neighbourhood-level covariateadjusted estimates of behavioural risk factors:

- i. to help public health planning
- ii. to inform future neighbourhood-level models of chronic disease

Objectives

- Estimate current smoking and excess
 bodyweight prevalence with acceptable
 precision and accuracy, accounting for spatial
 correlation and potential confounders
- ii. Identify areas of unusually high prevalence
- iii. Describe the spatial distribution over the entire study area by sex



Study Area



Study Area

CCO



Study Area







Methods



CCHS Data

- Risk factor data from Canadian Community Health Survey (CCHS), 5 cycles (2000-2001 to 2009-2010). Similar to BFRSS
 - <u>Current Smoking</u>: smoke daily or occasionally <u>Excess Bodyweight</u>: body mass index (BMI) <u>> 25 kg/m²</u>
- Postal code conversion file used to identify respondents' neighbourhoods



Geographical Data

- Geographical unit ("neighbourhood") = 2006
 Census Dissemination Area (DA)
 - Smallest geographical unit with full set of census data
 - Population: 400-700 people
 - Size varies: urban area =~ city block, rural
 DAs are larger, defined by rivers, roads, etc.

Bayesian Analysis

- Problems:
 - i. Small numbers of cases (0, 1, 2,...)
 - ii. Spatial dependence
- <u>Solution</u>: Bayesian modeling with hierarchical random effects
 - Allows for uncertainty due to low counts
 - Uses spatial dependence to pool information from adjacent areas



Fixed vs. Random Effects

Deterministic Model

Stochastic Uncertainty

Uncertainty Propagation





Model Specification

- Outcome: binary individual-level risk factor (yes/no) for current smoking and excess bodyweight
- Logistic regression: log(p/(1-p))
- **Covariates:** CCHS cycle, age group (10 yr groups), median neighbourhood-level income

<u>Model 1:</u> CCHS cycle & age group only <u>Model 2:</u> CCHS cycle, age group, income

Model Specification

- Bayesian analysis using Besag, York & Mollié (BYM) model
- Post-stratification weighted findings based on neighbourhood demographics due to complex CCHS sampling
- SaTScan used to corroborate findings using raw CCHS data





Results



Current Smoking: Model 1



Current Smoking: Model 2



Excess Bodyweight: Model 1



Excess Bodyweight: Model 2



Validity & Precision

Validity: Bayesian model-based estimates corresponded to CCHS design-based results

Precision: Coefficient of variation (CV; std error/ mean). Statistics Canada's CV thresholds (low, marginal, acceptable) for each neighbourhood

<u>Current Smoking:</u>

Model 1: mostly marginal (M: 90.1%, F: 96.5%), Model 2: mostly acceptable (M: 89.1%, F: 62.1%)

Excess Bodyweight: Mostly acceptable CVs in both sexes and models (>99%)

Strengths & Limitations

Strengths:

- High resolution heterogeneity vs. surveys
- Consistent Bayesian & SaTScan findings
- Assessed validity & precision

Limitations

- Self-reported survey data
- CCHS not designed to be pooled
- Assume neighbourhood-level household income stable over time



Conclusions

- Estimating high resolution risk factor prevalence from large survey data is feasible
- Better precision for Model 2, suggesting that inclusion of key covariates is important
- Applications for public health planning and studies of related chronic disease outcomes (e.g. cancer) at neighbourhood level

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