

Smoothing rates and disease mapping

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Disease mapping

- **Analysis of the spatial variation of disease incidence or mortality**
- **Its subsequent representation on a map**

Population/public health
Health or health service outcomes
Health service utilization
Survival rates
Others

Disease mapping: retrospective study

- A fixed time frame (time window of 5 or 10 years)
- Disease risks of local areas of a region under study
- Underlying geographical distribution of disease incidence/mortality rates

Applications:

- Disease and health outcome epidemiology and surveillance (disease or injury incidence, mortality, etc)
- population health study (disease, injury, low birth-weight rate, etc)
- Health service research (health outcomes, service utilization, etc)

The Data: (simple spatial design)

- Y_j : number of disease incidence or mortality in health region j , $j = 1, \dots, J$ (J regions under study)
- N_j : the corresponding “population at risk” (often the mid-period population estimate)
- X_j : the covariates

Objective:

Explore spatial risk clusters; identify high risk regions and risk factors

The Data: (spatial design with age effects)

- Y_{jk} : number of disease incidence or mortality for age group k ($k = 1, \dots, K$) and in health region j , $j = 1, \dots, J$ (J regions under study)
- N_{jk} : the corresponding “population at risk” (often the mid-period population estimate)
- X_{jk} : the covariates

Objective:

Explore spatial risk clusters; identify high risk regions and risk factors (age specific or adjusted)

The Data: (spatio-temporal design)

- Y_{jt} : number of disease incidence or mortality in health region i and year t (eg. $t = 1991, \dots, 2000$); (For HSDA, for example, $j = 1, \dots, 16$)
- N_{jt} : the corresponding “population at risk” (often the mid-year population estimate)
- X_{jt} : the covariates

Objective:

Model non-linear rate trends; uncover evolution of spatial risk clusters

The Data: (spatio-temporal design with age effects)

- Y_{jkt} : number of disease incidence or mortality
- N_{jkt} : the corresponding “population at risk” (often the mid-year population estimate)
- X_{jkt} : the covariates

Focus

- Understand disease occurrence

Area-specific rates are assessed as an ensemble

- Assess spatial variation of disease risks
- Predict disease risk over space and time
- Capture spatial patterns evolving over time
- Evaluate health services outcome and resource utilization

Important Issues

- The areal units (Administrative health regions)
- The measures used for the construction of disease map
- The precision of measurement
- Trade-off between resolution of the data and precision of measurement

Map

- True underlying spatial variation?
- Chance variation?
- Artifacts of the “at-risk population” variation?

Simple Statistical Representation

- Crude rate: Y_j/N_j , $j = 1, 2, \dots, J$

Frequency of disease occurrence in the j -th local area

- 'Global' mean rate (BC rate for example): m

$$m = \text{sum}(Y_j) / \text{sum}(N_j)$$

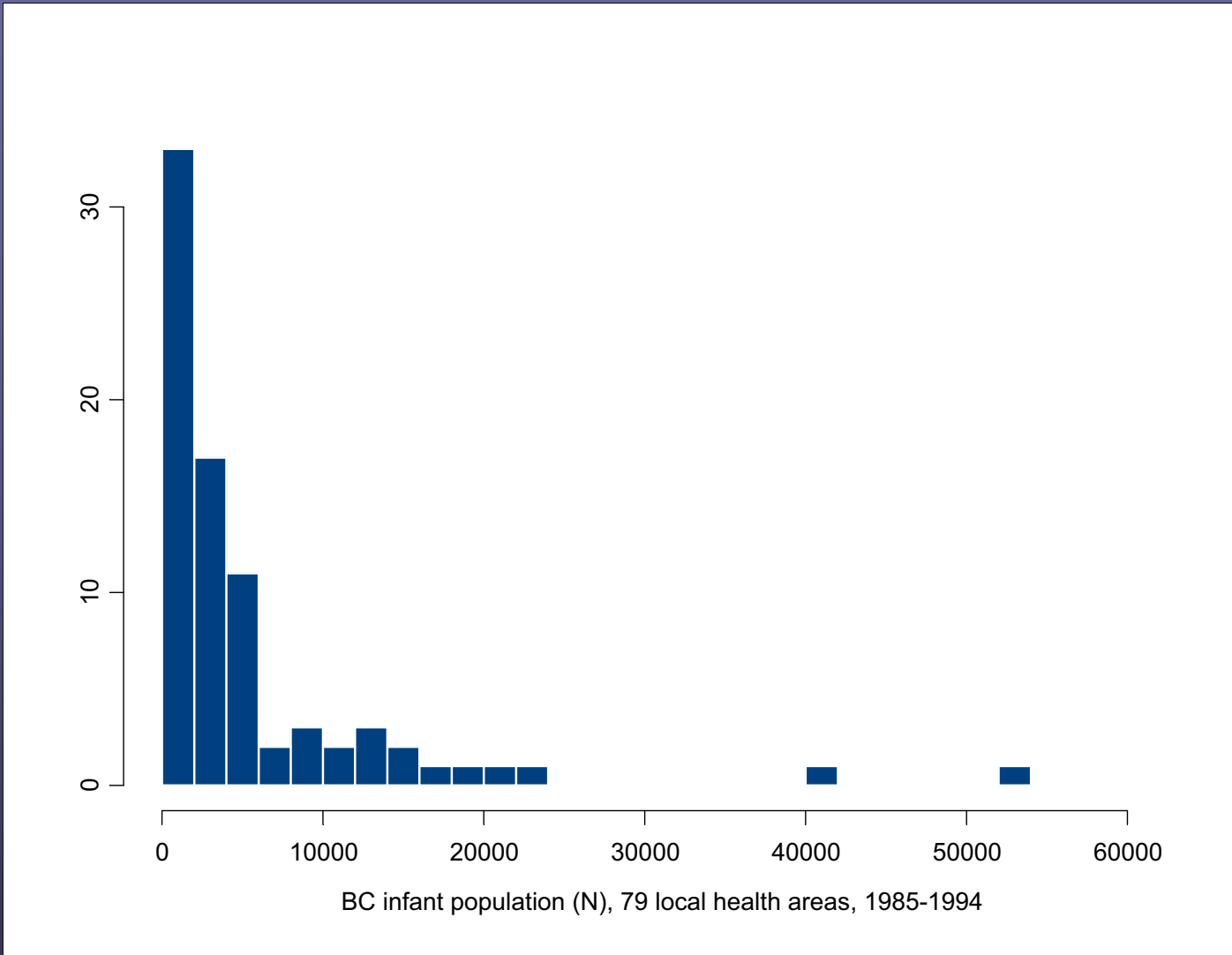
- Rate ratio (also called relative risk):

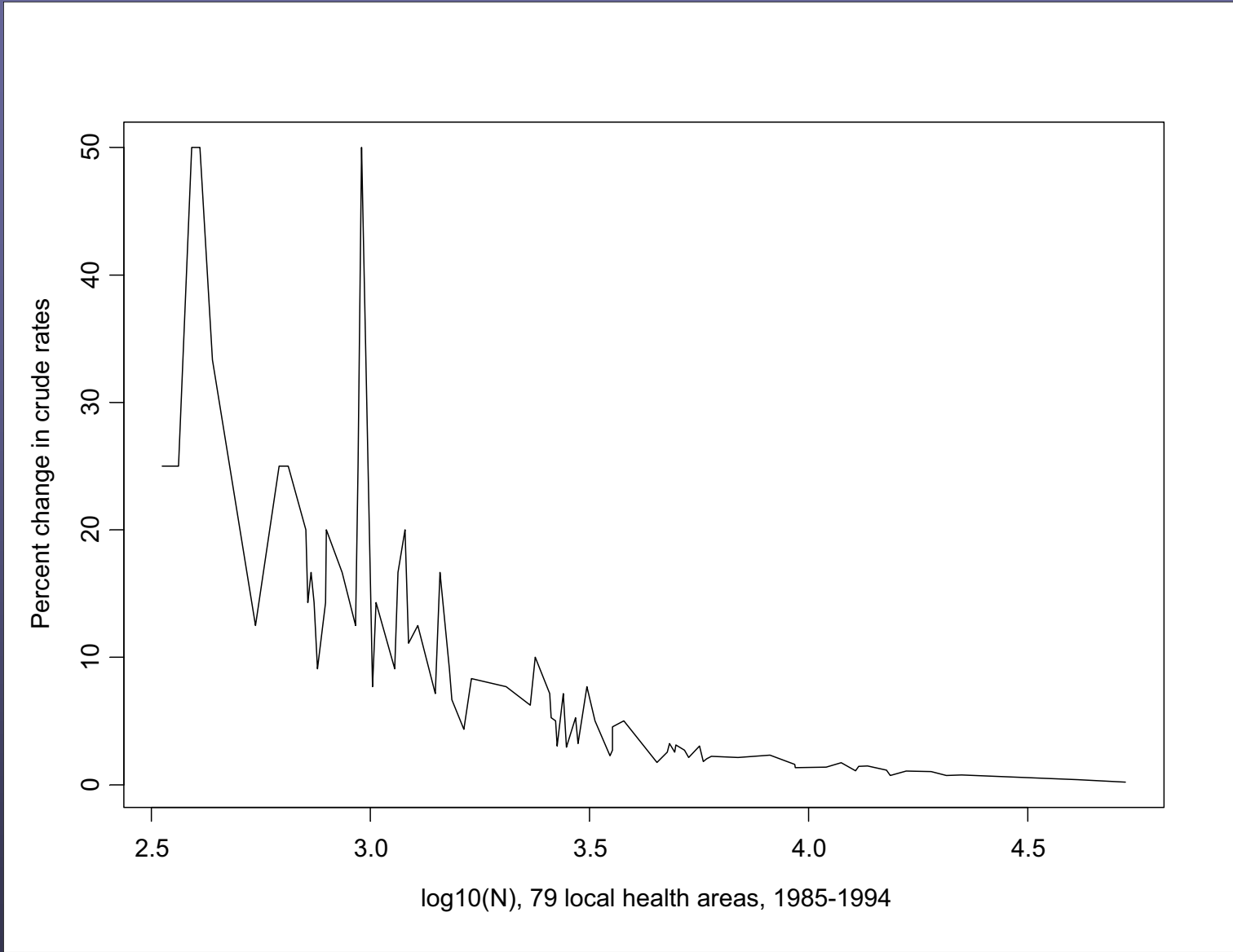
$$RR_j = (Y_j/N_j) / m$$

Ratio of local area rate over 'global' mean rate or ratio of observed cases/expected cases

Limitations of the Simple Statistics

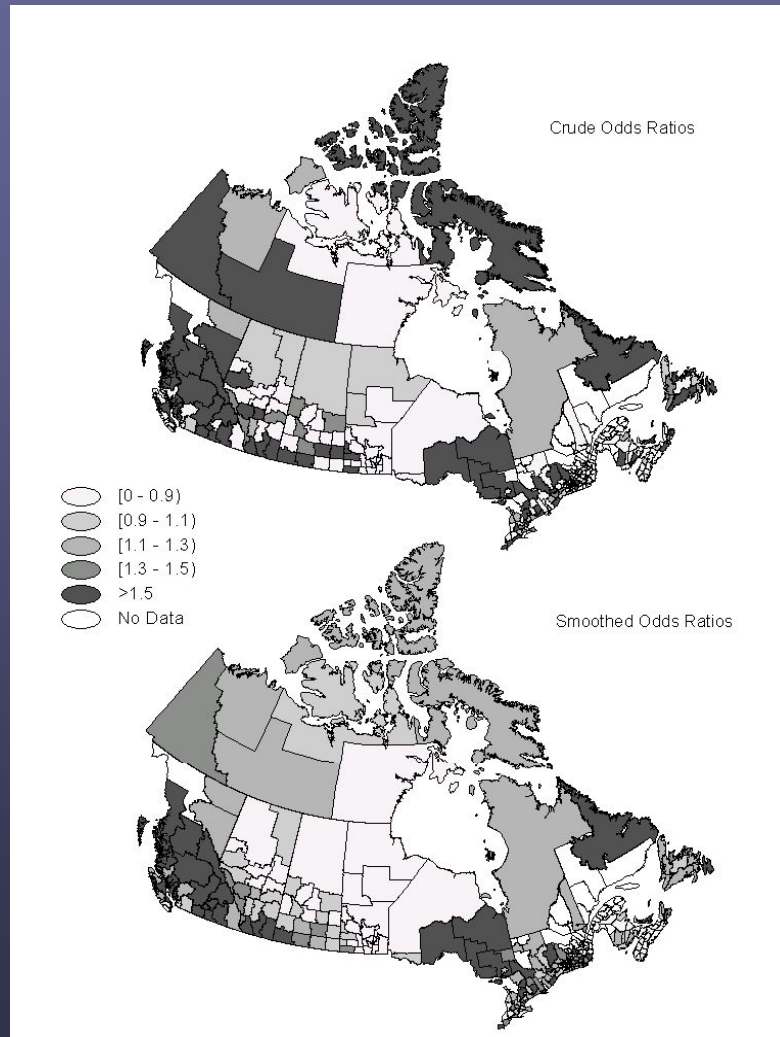
- Crude rates or ratios may have poor precision when rare diseases are investigated in small areas
- Areas of low population tend to present extreme rates/ratios, which may not reflect truly elevated disease risk
- Rates with large chance variation tend to highlight the map
- Such map does not yield meaningful/useful interpretation





Chronic Lung disease, Canadian NICU data, 1996-1997

Smoothed ratios obtained via hybrid Markov Chain Monte Carlo algorithm



Model-based disease mapping (Bayesian)

- Predict the local area risks *ensemble* in an optimal way
- Separate systematic variability from random noise, a component that usually overshadows the crude maps
- Produce `clean' maps, `clean' of random noise and any artifacts of population variation
- `Borrow' strength and smoothing

Spatial smoothing

Non-spatial “Global” smoothing

Bayesian ecological regression

- Examine response-covariate association
- Identify regional determinates (or sources of relative risk variation)
- Contrast unadjusted and (risk) adjusted relative risks (quantify sources of RR variations)

Spatio-temporal models (Spatio-temporal smoothing)

- Model non-linear rate trends (both “global” and “local”)

BC study for example:

global rates: annual BC rates

local rates: annual LHA rates

- Uncover evolution of spatial risk clusters

Neonatal Health Services in Canada

Use Bayesian hierarchical methodologies to

- examine hospital variations in neonatal intensive care unit (NICU) outcomes and utilization rates in Canada
- assess geographic variations of adverse birth outcomes, NICU outcomes and NICU resource use
- link practice patterns to outcome variations
- develop a national neonatal health geographical information system (GIS)



Canadian Neonatal Database

- Population-based:

 - 20 488 admissions (19 507 infants) Jan 96 – Oct 97

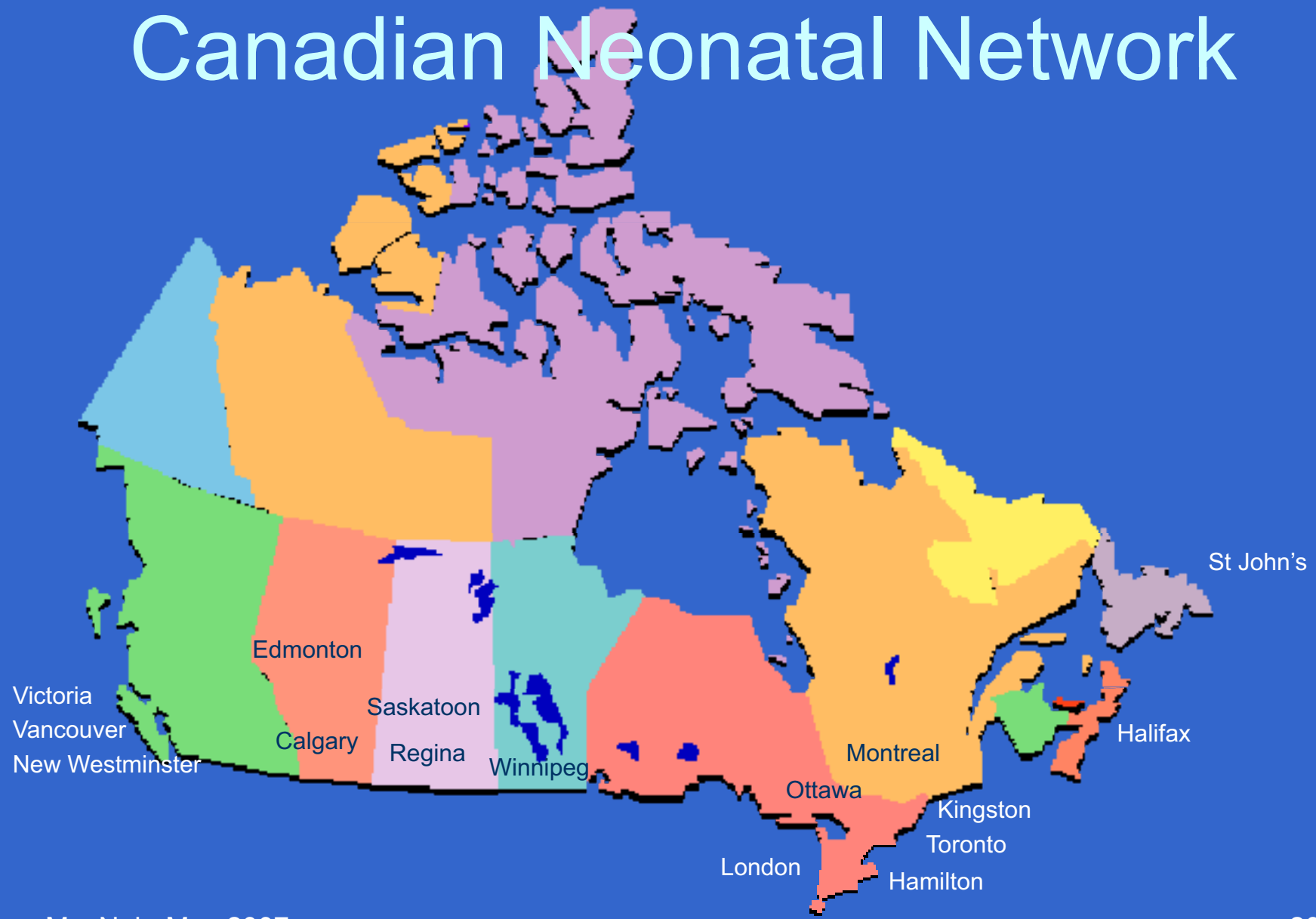
 - 17 NICUs across Canada (75% tertiary level NICU beds)

- Detailed patient information:

demographic information, antenatal history, mode of delivery, problems at delivery, illness severity, selected NICU practices and procedures, therapeutic intensity, use of technology and resources, selected patient outcomes

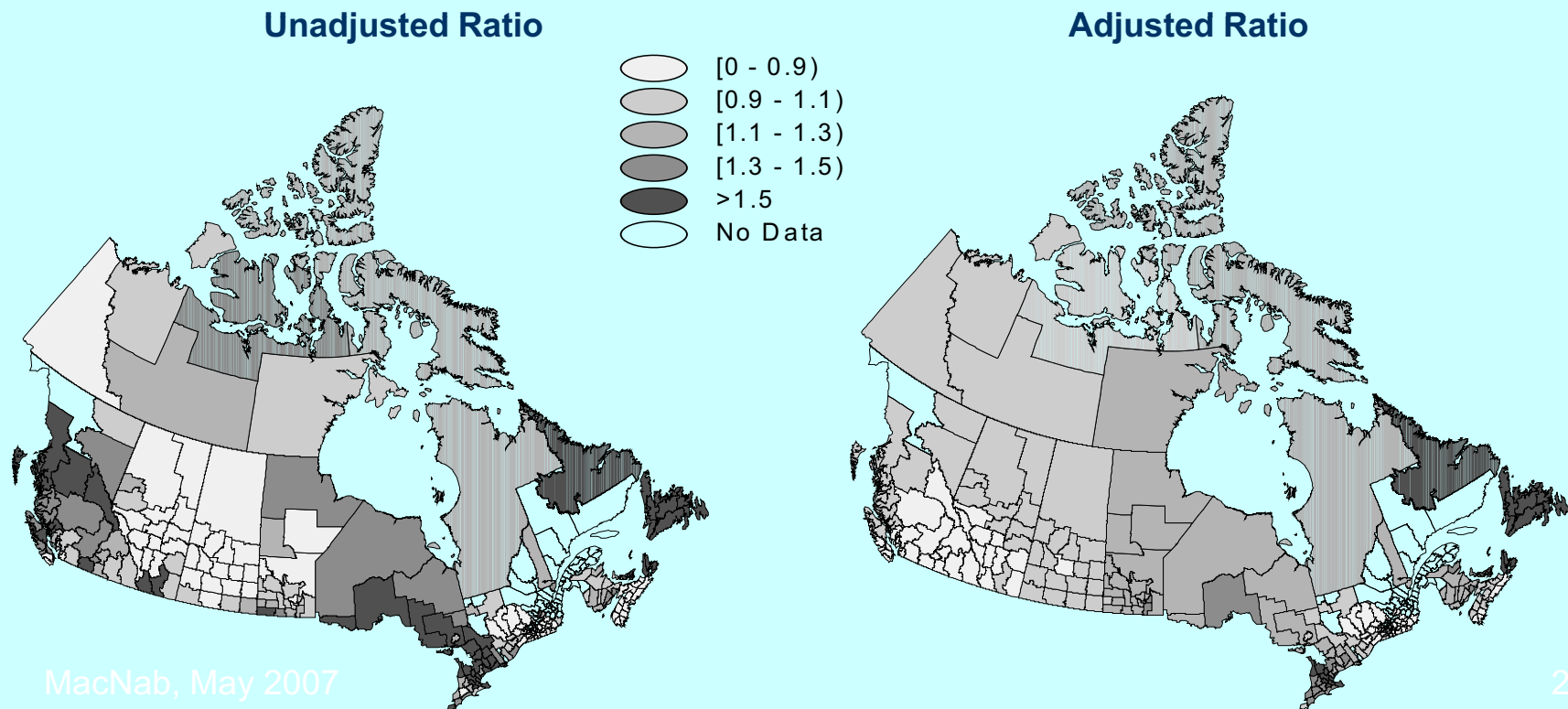
- Hospital characteristics:

Canadian Neonatal Network



Intraventricular hemorrhage incidence, Canadian NICU data, 1996-1997

- Ratios obtained via hybrid Markov Chain Monte Carlo algorithm.
- Adjusted ratios: adjusted for VLBW, LBW, outborn, SGA, SNAP-II >20.



Burden of Injury in BC and Its Local Communities: Information and Evidence for Community-based Prevention Strategy, Health Policy and Service Provision

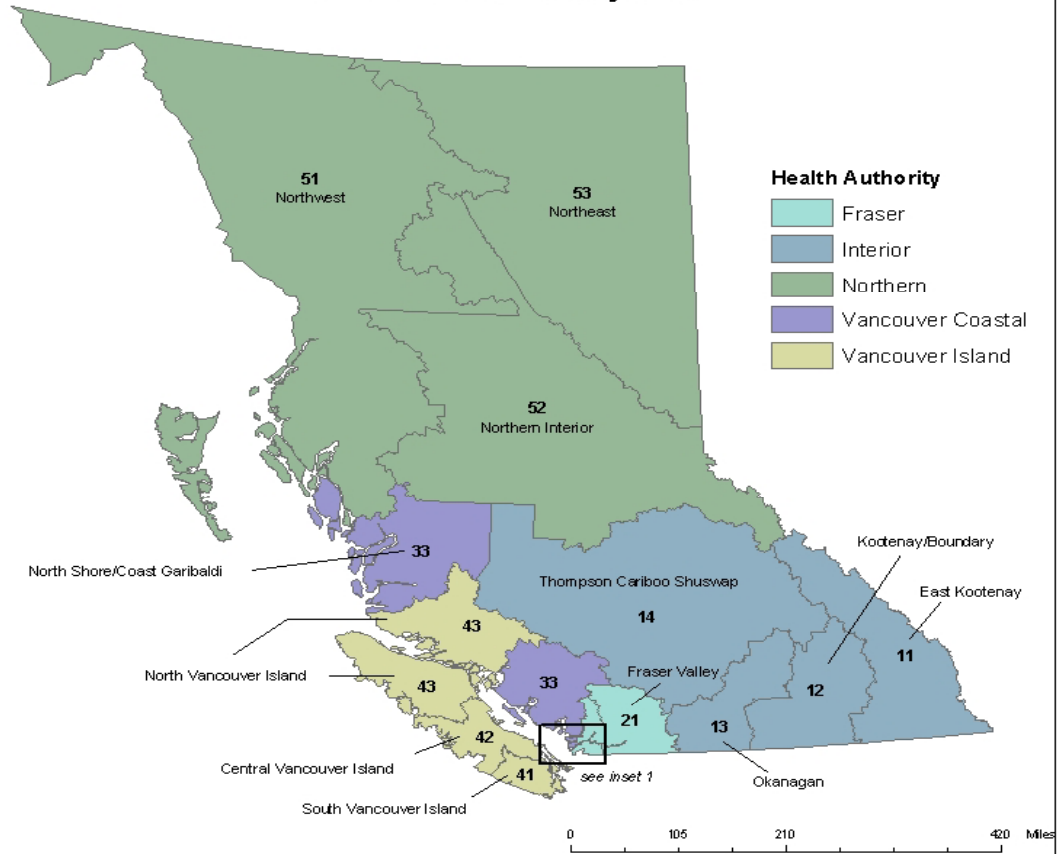
This study aims

- to develop a synthesis of analytic methods for a systematic burden of injury research framework that encompasses space-time surveillance monitoring, burden assessment, risk assessment, research dissemination and knowledge translation (priority setting in particular)
- to apply these methods to examine burden of injury mortality and disability in British Columbia (BC) and its local communities.

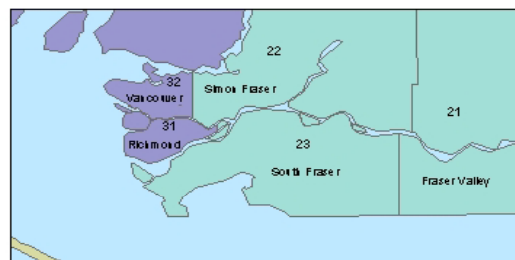
British Columbia Health Regions:

- *5 Health Authority Areas (HAs)*
- *16 Health Service Delivery Areas (HSDAs)*
- *84 Local Health Areas (LHAs)*

Health Service Delivery Areas



inset 1



Health Authority

- 01 Interior
- 02 Fraser
- 03 Vancouver Coastal
- 04 Vancouver Island
- 05 Northern

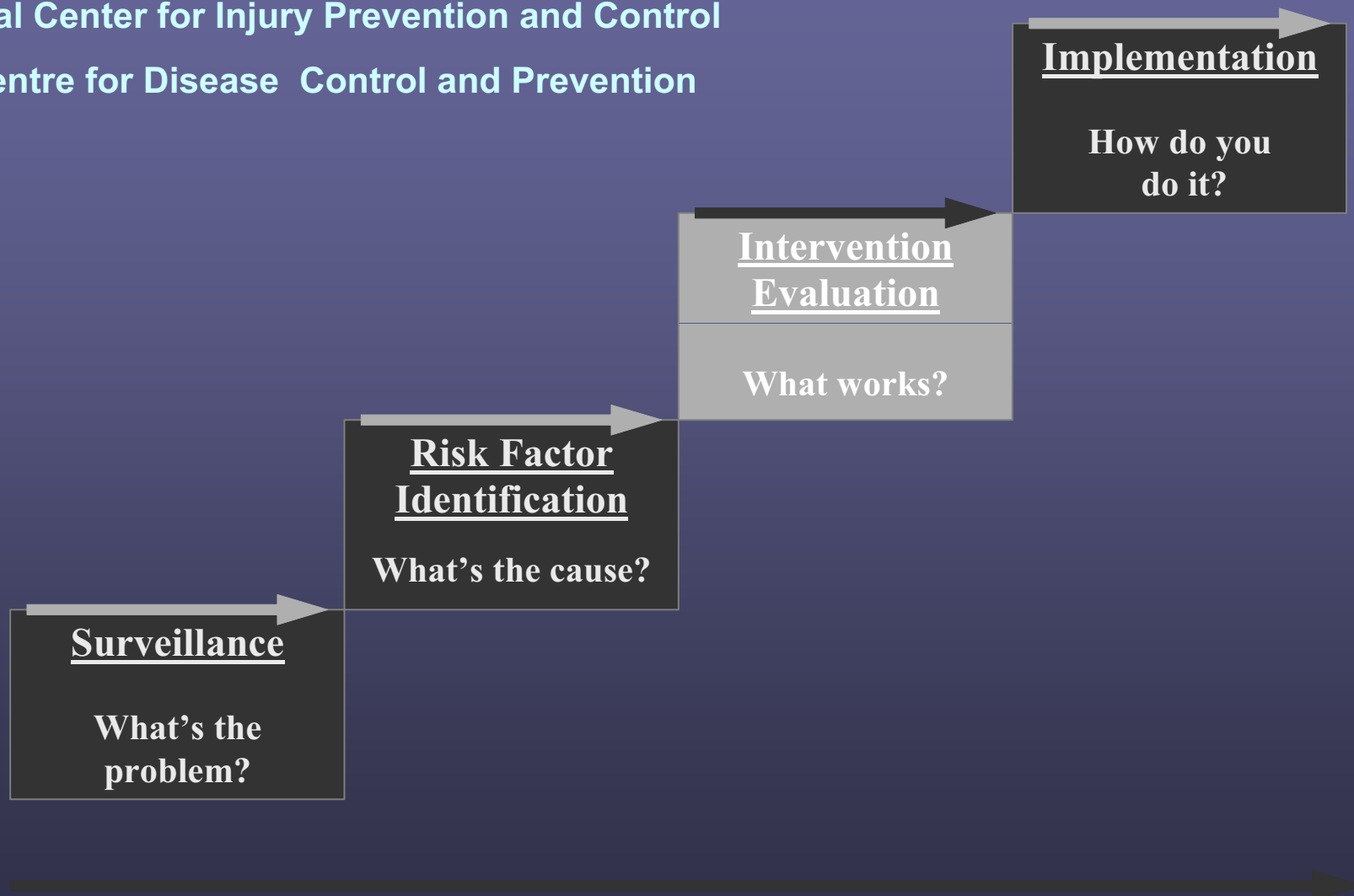
Health Service Delivery Area

- 11 East Kootenay
- 12 Kootenay Boundary
- 13 Okanagan
- 14 Thompson Cariboo
- 21 Fraser Valley
- 22 Simon Fraser
- 23 South Fraser
- 31 Richmond
- 32 Vancouver
- 33 North Shore/Coast Garibaldi
- 41 South Vancouver Island
- 42 Central Vancouver Island
- 43 North Vancouver Island
- 51 Northwest
- 52 Northern Interior
- 53 Northeast

Public health approach to injury prevention and control

National Center for Injury Prevention and Control

U.S. Centre for Disease Control and Prevention



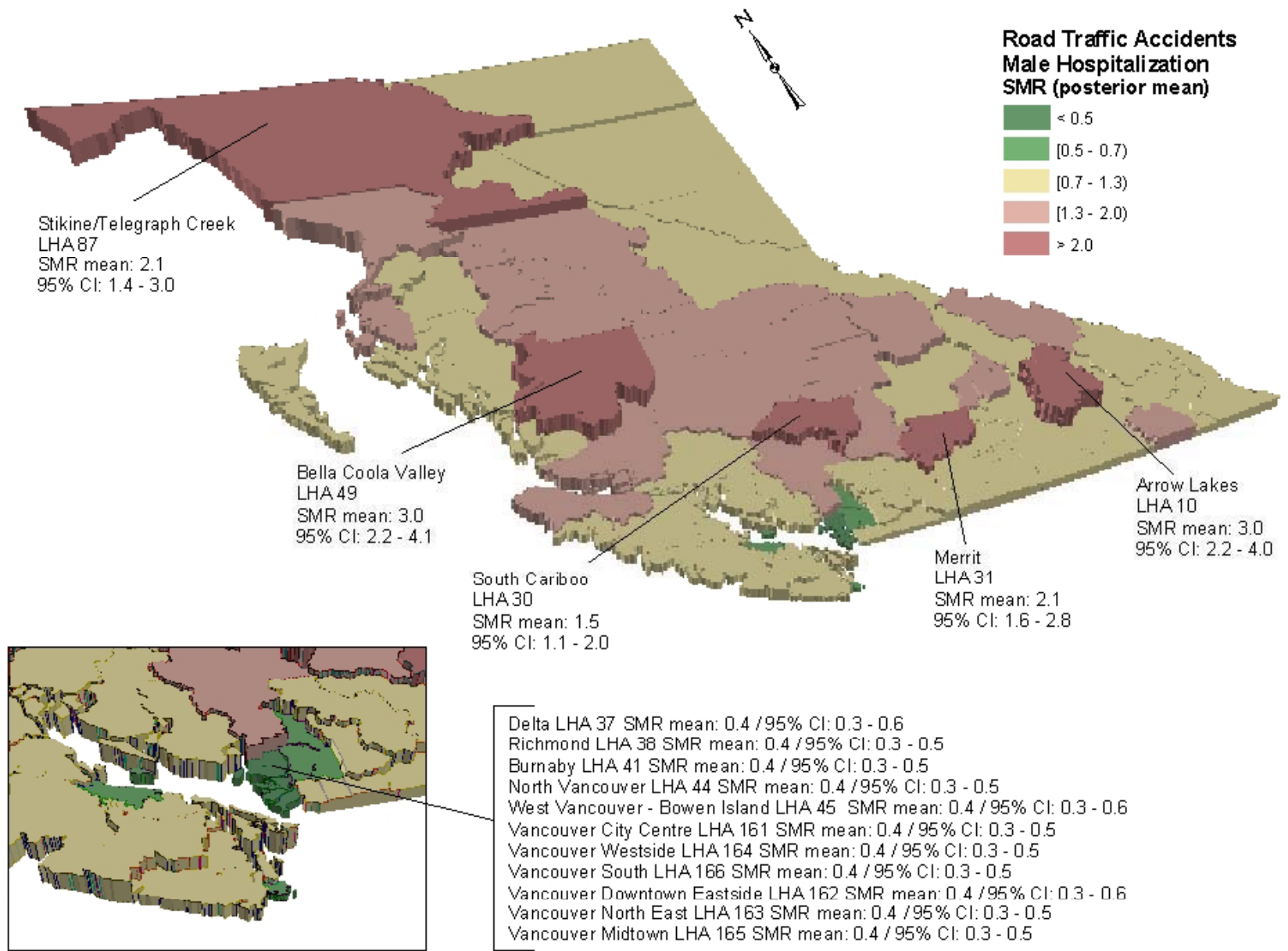
Spatial modeling

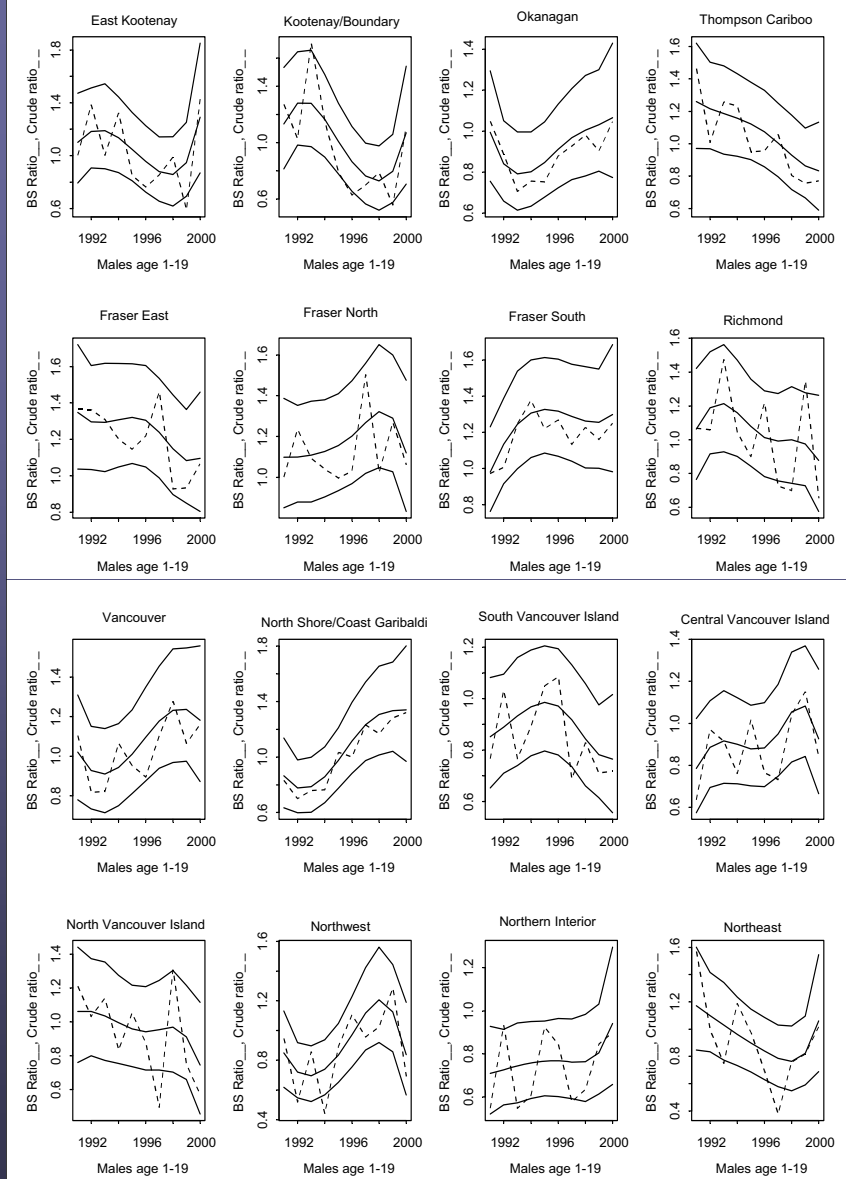
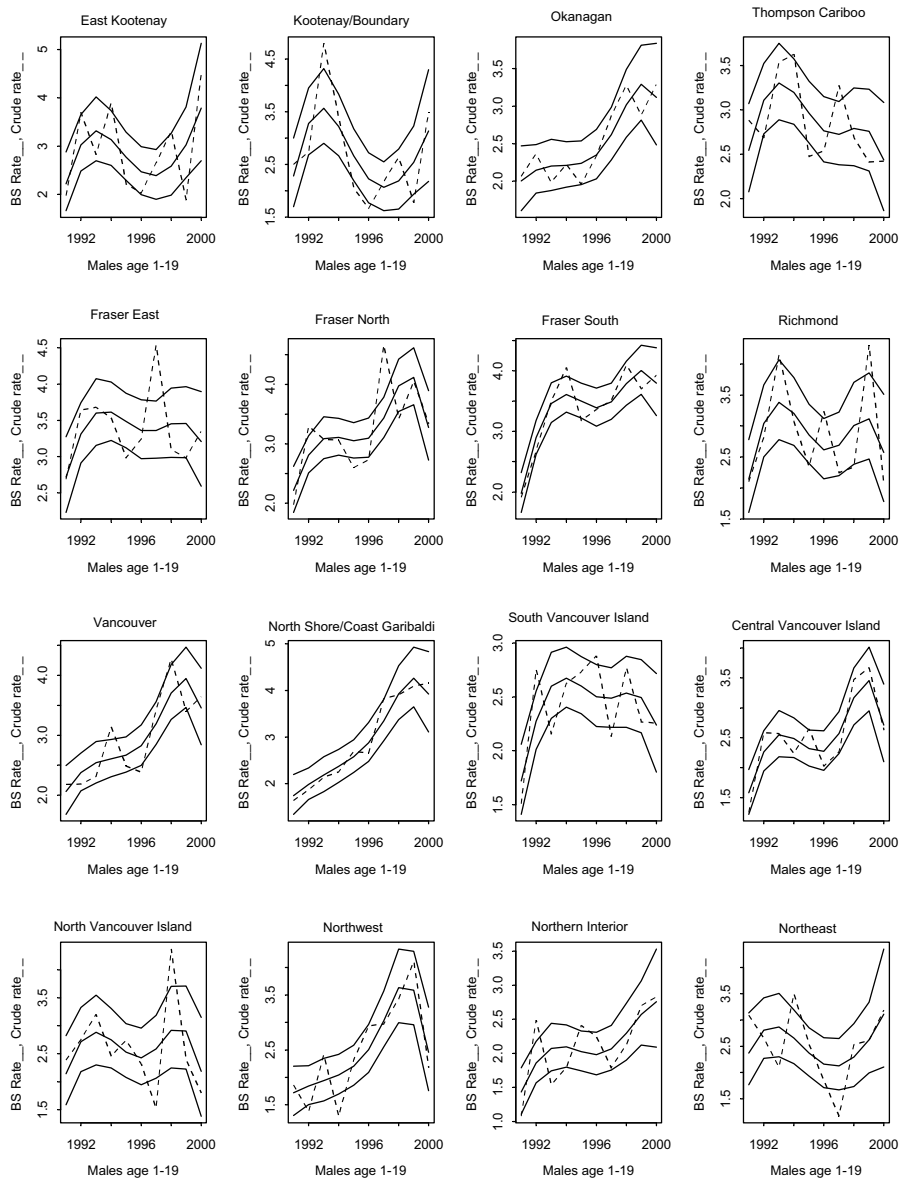
Ecological modeling

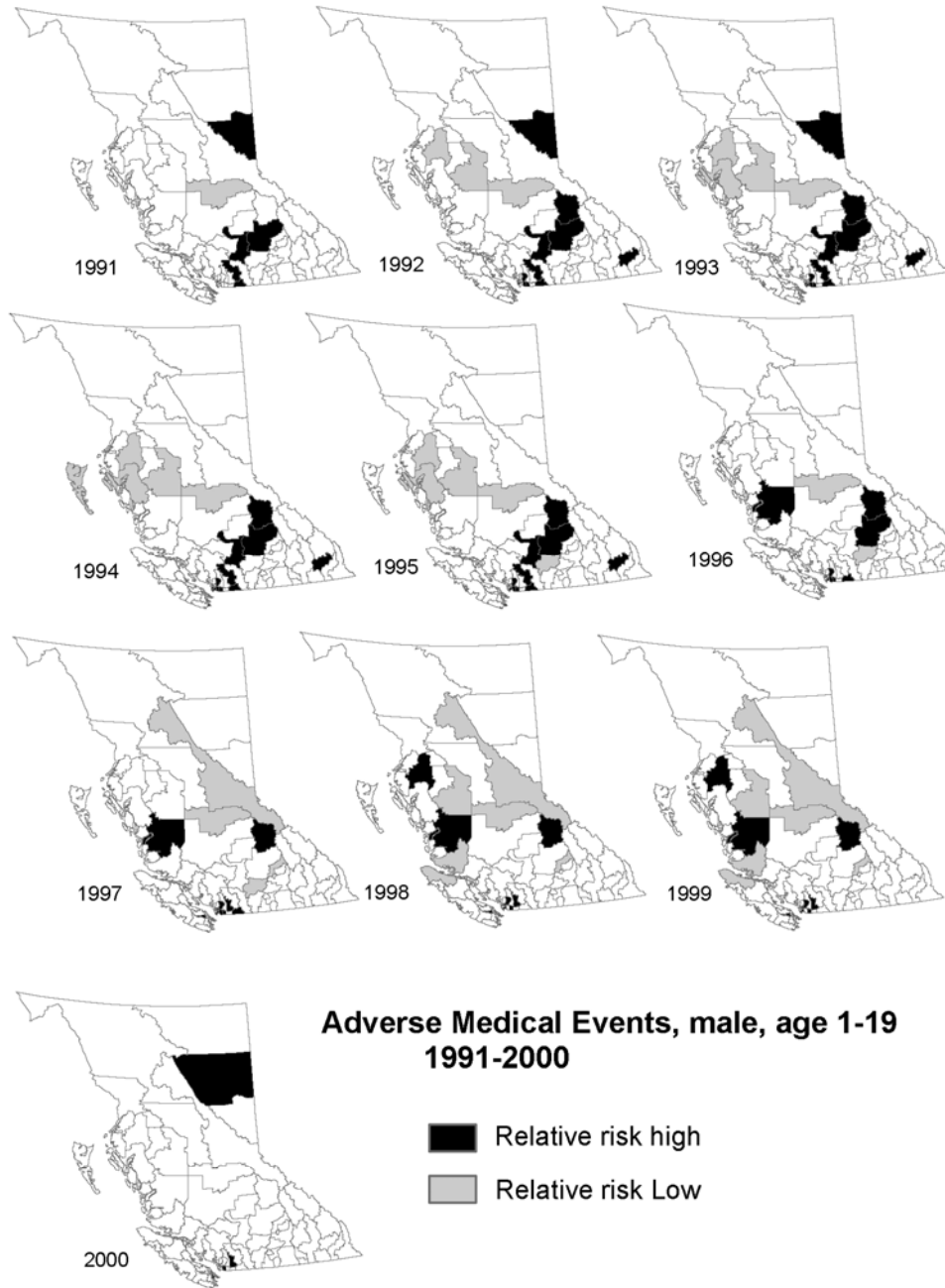
Spatio-temporal modeling

Small area disability adjusted life year

(Disease mapping methodology + Disability adjusted life year methodology)







The end