# D<sub>o</sub>MiNO - Spatial data mining exploring co-location of adverse birth outcomes and environmental variables.

Dr. Alvaro R. Osornio-Vargas

Canadian Perinatal Programs Coalition
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# D<sub>o</sub>MiNO is:

- Interdisciplinary and Exploratory
- Uses publicly funded data bases
- DATA MINING

# Background

Current research identifies <u>associations between</u>
ABO and various determinants of health:

- social factors (e.g. poverty, stress),
- biological factors (e.g. diabetes, infection, maternal age)
- environmental pollutants (e.g. metals, PM<sub>10</sub>, SO<sub>2</sub>)

# Growing evidence linking Urban Air Pollutants and Adverse Birth outcomes

# Background

#### Complex problem:

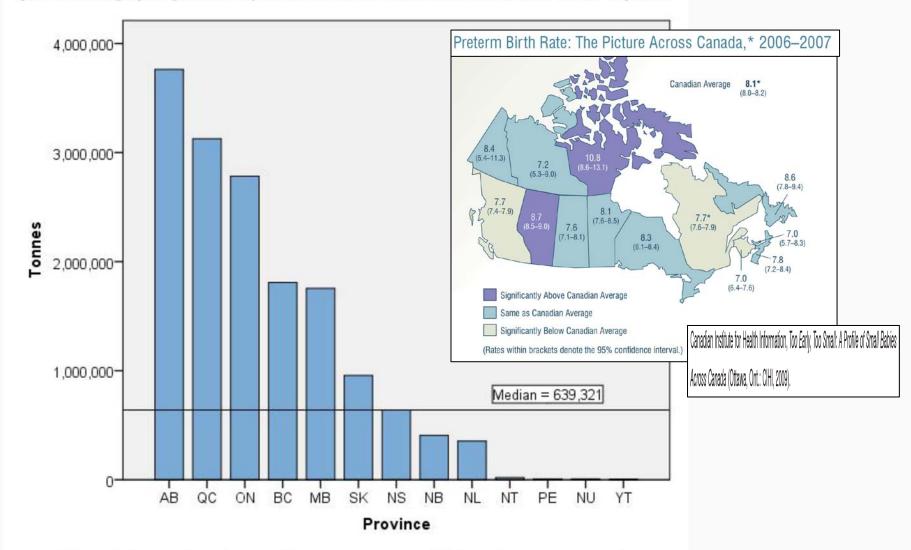
- Multiple sources of pollutants (e.g. traffic, industry).
- Interactions, dispersion, transport and fate of pollutants
- Intrinsic toxicity of pollutants
- Interactions between, social, biological, chemical and physical factors

# Background

In order to advance research on links between <u>environmental pollutants and ABO</u>, methods for comprehensive assessment of the multiple variables interacting in complex ways are required.

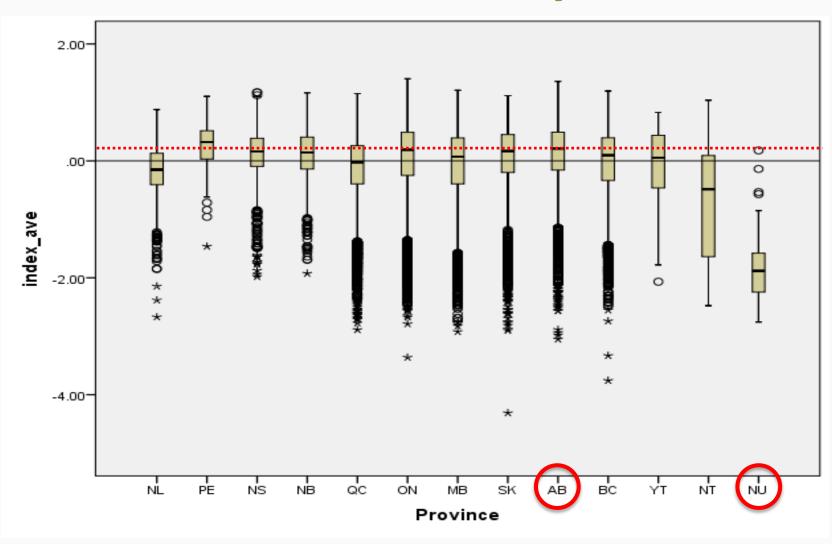
### DATA MINING

Spatial data mining exploring co-location of adverse birth outcomes and environmental variables. Osornio-Vargas, A.R.



**Figure 2.** Tonnes of developmental toxicants reported to NPRI as released to air in each Province of Canada, from 2006 – 2010. Numbers of chemicals released vary by Province. The line marks median emissions.

# Distribution of SES index by province and territory



# Adverse birth outcomes and the environment

Statistics Canada
National Pollutants Release Inventory
Wind Patterns
The Canadian Neonatal Network
Alberta Perinatal Health Program

Data Mining

Identify patterns

Hypothesis
(e.g. collocation)

# Objective

#### Aims to:

Identify patterns
Generate <u>hypothesis</u>
Spatial collocation

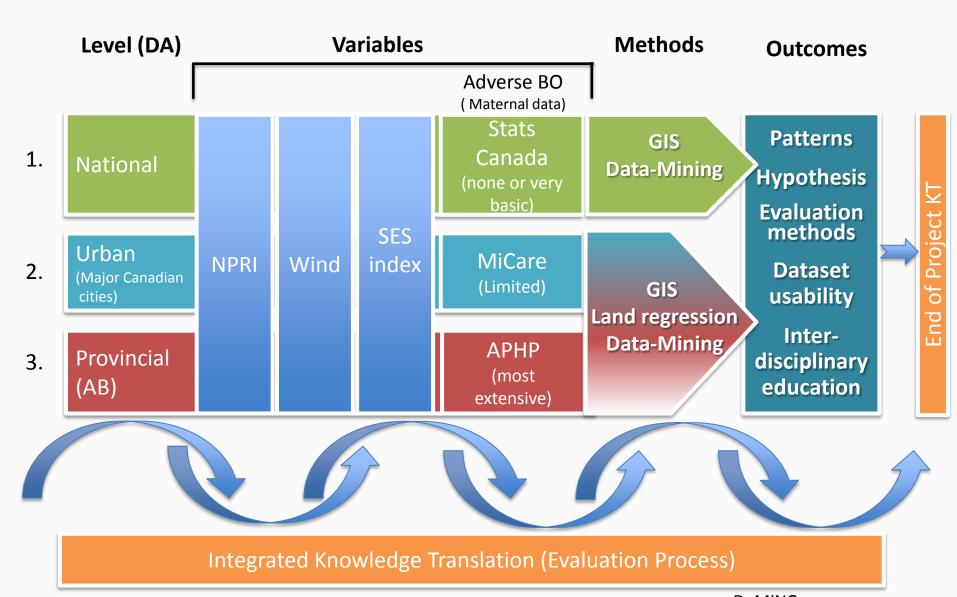
Adverse birth outcomes

(pre-term, low birth weight and perinatal mortality)

**Environmental Factors** 

(pollution, SES)

#### **Research Framework**



DoMiNO
Data Mining newborn outcomes

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	Chemical name	Tonnes		Chemical name	Tonnes
1	Sulphur dioxide	7,614,400	30	HCFC-22	487
2	Carbon monoxide	4,744,224	31	Chloroform	475
3	Volatile Organic Compounds (VOCs)	1,279,186	32	Naphthalene	449
4	PM Total Particulate Matter	962,176	33	Arsenic (and its compounds)	353
5	PM <sub>10</sub> Particulate Matter ≤ 10 μm	520,352	34	Methyl methacrylate	302
6	PM <sub>2.5</sub> Particulate Matter ≤ 2.5 μm	277,572	35	Acetonitrile	197
7	Methanol	69,679	36	Tetrachloroethylene	196
8	n-Hexane	26,108	37	1,3-Butadiene	193
9	Xylene (all isomers)	25,897	38	tert-Butyl alcohol	126
10	Toluene	21,220	39	Cadmium (and its compounds)	121
11	Hydrogen fluoride	16,984	40	Acrylonitrile	69
12	Carbon disulphide	16,377	41	Butyl benzyl phthalate	47
13	Styrene	9,522	42	N,N-Dimethylformamide	46
14	Methyl ethyl ketone	8,653	43	Sodium nitrite	45
15	Isopropyl alcohol	6,947	44	Benzo(a)pyrene - PAH	42
16	Acetaldehyde	5,117	45	Bis(2-ethylhexyl) phthalate	37
17	Ethylbenzene	4,055	46	1,2,4-Trichlorobenzene	27
18	Benzene	3,257	47	p-Dichlorobenzene	27
19	Phenol (and its salts)	3,031	48	Ethylene oxide	22
20	2-Butoxyethanol	2,747	49	Mercury (and its compounds)	22
21	Chloromethane	2,242	50	Biphenyl	19
22	Chlorine dioxide	2,118	51	Vinyl chloride	10
23	Methyl isobutyl ketone	1,412	52	Dibutyl phthalate	7
24	Trichloroethylene	1,270	53	1,2-Dichloroethane	6
25	Lead (and its compounds)	1,144	54	2-Ethoxyethyl acetate	6
26	Nickel (and its compounds)	1,131	55	Ethylene thiourea	4
27	Ethylene glycol	830	56	Bromomethane	1
28	Acrolein	715	57	Chlorobenzene	1
29	N-Methyl-2-pyrrolidone	673	58	Ethyl acrylate	1
	Total	15,629,039		Total	3,338
				GRAND TOTAL	15,632,377

Table I :Total amounts of developmental toxicants reported to NPRI as released to air in Canada in 2006-2010

#### 4 Jundong Li, Osmar R. Zaïane and Alvaro Osornio-Vargas

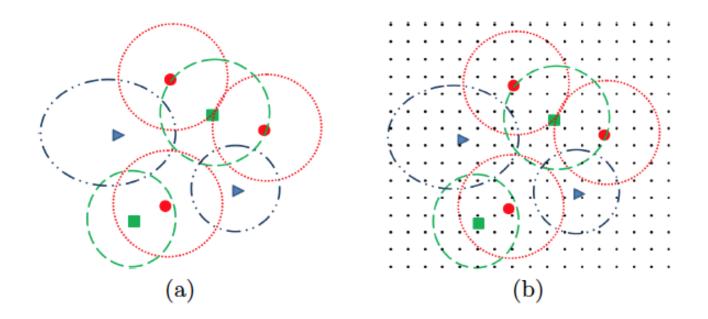


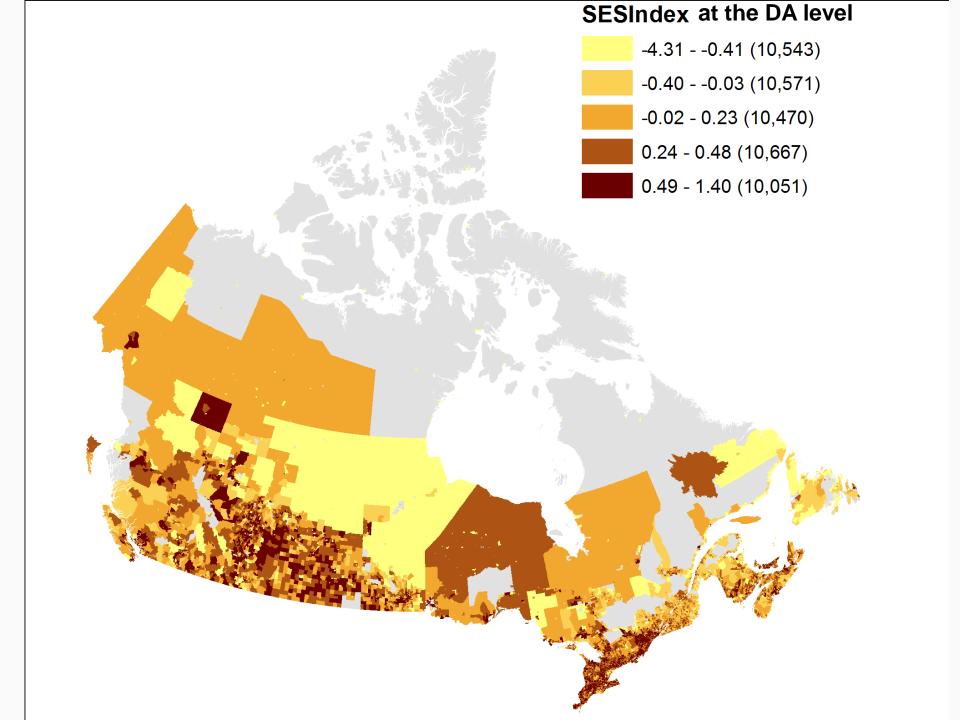
Fig. 1. Transactionization step: (a) An example spatial dataset with point feature instances and their buffers; (b) Grids imposed over the space.

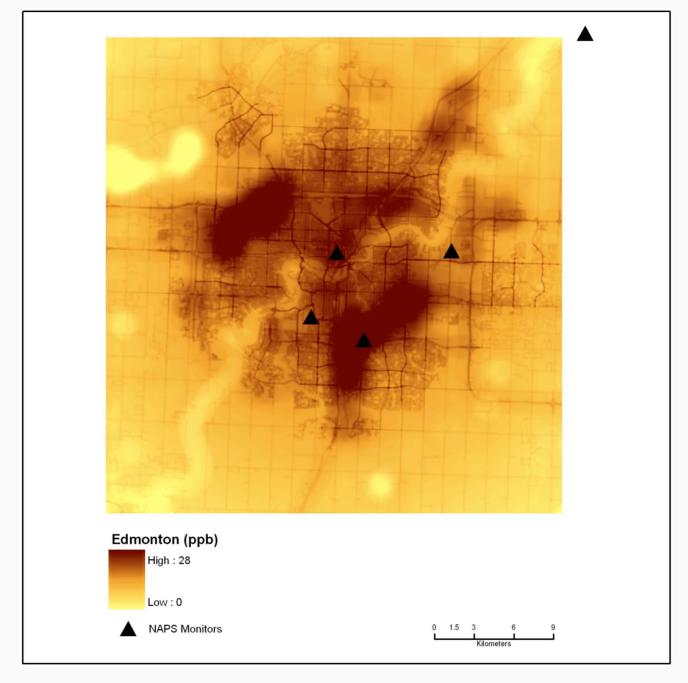
$$p = \sum_{i=\sigma(XA)}^{\sigma(A)} \binom{n}{i} (P(X)P(A))^{i} (1 - P(X)P(A))^{n-i}$$
 (1)

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Algorithm 1 CMCStatApriori Algorithm.

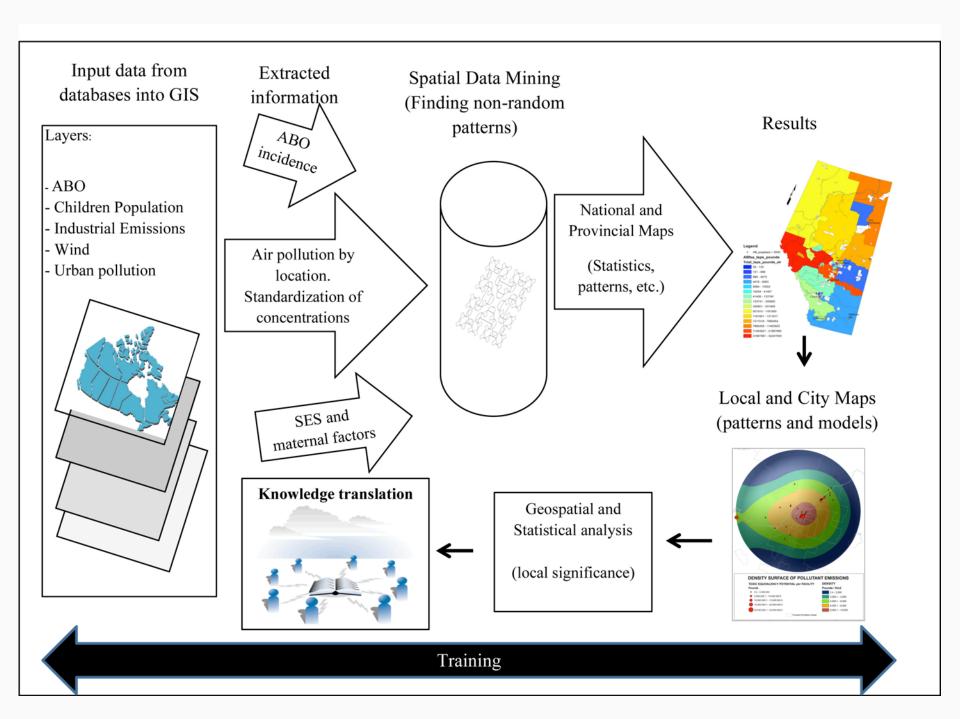
Require: Set of antecedent features F \setminus A, the consequent feature A, derived transaction dataset T, the threshold z_{min} for the z-score

Ensure: Set of potential statistical significant co-location rules P
1: P_1 = \{f_i \in F \setminus A | PS(f_i) = 1\}
2: l = 1
3: while (P_l \neq \varnothing) do
4: C_{l+1} = GenCands(P_l, A)
5: P_{l+1} = PrunCands(C_{l+1}, z_{min}, A)
6: l = l + 1
7: end while
8: P = \bigcup_i P_i
9: return P
```





R.W. Allen et al. / Atmospheric Environment 45 (2011) 369-378



#### Table II. Evaluation of the integrated KT plan.

<b>Evaluation question</b>	Proposed methods		
What perspectives do researchers from	1. Participant observation		
different disciplines and knowledge users have	2. Individual semi-structured, time series interviews		
on interdisciplinary partnership research? How			
do these perspectives change over the course of			
the research project?			
What challenges and barriers are	3. Participant observation		
experienced to interdisciplinary team	4. Individual semi-structured, time series interviews		
development?	5. Content analysis		
	6. Analysis of project log		
	7. End of project survey		
What strategies are most useful in building	8. Participant observation		
collaboration and addressing identified	9. Individual semi-structured, time series interviews		
barriers?	10. End of project survey		
What are the added benefits of	11. End of project survey		
interdisciplinary research-knowledge user	12. End of project focus group		
collaboration?	13. Analysis of project deliverables		

#### **Team**

#### **University of Alberta**

Faculty of Medicine & Dentistry

Dr. Osornio-Vargas\* Principal Investigator

Dr. Irena Buka Children's environmental health

Dr. Khalid Aziz Neonatology Dr. Manoj Kumar Neonatology

Or. Sue Chandra
Osnat Wine
Emily Chan
Obstetrics & Gynecology
Knowledge Translation
Socioeconomic variables

**Computing Sciences** 

Dr. Osmar Zaiane\* Principal Investigator

Jundong Li Data mining Dr. Dr. Sajib Barua Data mining

School of Public Health

Dr. Sarah Bowen Knowledge Translation

Dr. Yan Yuan Biostatistics
Dr. Yutaka Yasui Biostatistics
Jesus Serrano Geostatistics

Faculty of Sciences

Charlene Nielsen Geostatistics

**Carlton University** 

Department of Health Sciences,

Dr. Paul Villeneuve Epidemiology

**University of Victoria** 

**Interdisciplinary Studies** 

Dr. Laura Arbour Paediatrics and Genetics

Anders Erickson GIS

**Oregon State University** 

School of Biological & Population Health Sciences
Dr. Perry Hystad Spatial exposure assessment

**CAREX** 

Dr. Eleanor Setton Exposure Assessment

Dr. Paul Demers Epidemiology

CIHR Maternal-Infant Care (MiCare) Program

Dr. Prakeshkumar Shah Neonatology

**Knowledge Users** 

Health Canada

Dr. David Stieb Epidemiology
Dr. Phil Blagden Science Advisor

Alberta Perinatal Health Program

Nancy Aelicks

Canadian Partnership for Children's Health & Environment

Erica Phipps

## Thank you

osornio@ualberta.ca

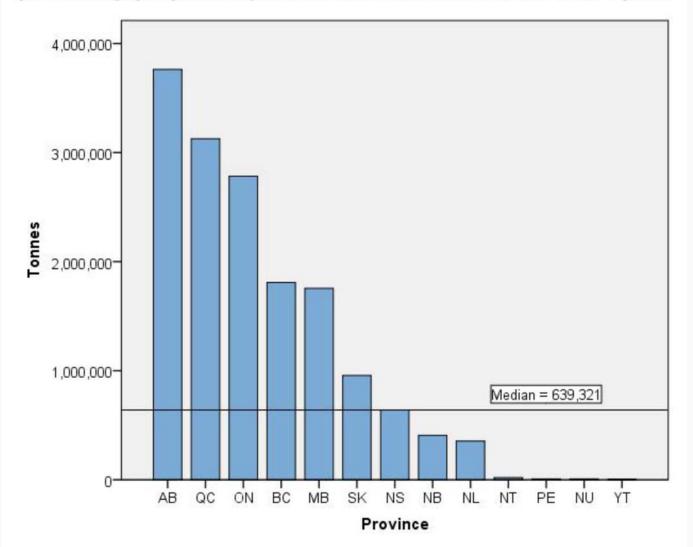


Figure 2. Tonnes of developmental toxicants reported to NPRI as released to air in each Province of Canada, from 2006 – 2010. Numbers of chemicals released vary by Province. The line marks median emissions.

# Data Mining

#### Conclusion

- New framework which uses (buffer & grid)-based transactionization and preserves spatial information better
- Statistical testing eliminates the usage of one global prevalence threshold
- Consideration of wind data and pollutants amounts to improve accuracy of results