Using air pollution sensors carried by cyclists and pedestrians to capture the spatio-temporal variability of air pollution in Toronto

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Objectives



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HEALTH



Air pollution boosts heart attack risk

 Public health significantly affected by air quality

 CBC News
 Posted: Feb 15, 2012 12:18 PM ET
 Last Updated: Feb 15, 2012 12:30 F



Spatial analysis of exposure to traffic-related air pollution at birth and childhood atopic asthma in Toronto Optorio

K. Shankardass^{a, b,} 📥 M. Jerrett^{c,} 🏧, S.D. E



Environment Inte

Volume 74, January 2015, I

Exposure to traffic-related air pollut developing breast cancer among w_ provinces: A case–control study

Perry Hystad^{a,} ▲, ᢂ, Paul J. Villeneuve^b, Mark S. Goldberg^{c, d}, Dan L Kenneth Johnson^f, the Canadian Cancer Registries Epidemiology R€



Environmental Research

Volume 115, May 2012, Pages 18-25

Neurobehavioral effects of exposure to traffic-relate pollution and transportation noise in primary school

Elise van Kempen^{a,} , Paul Fischer^a, Nicole Janssen^a, Danny Hou Kamp^a, Stephen Stansfeld^b, Flemming Cassee^a

Journal of Toxicology and Environmental Health, Part A: Current Issues

Publication details, including instructions for authors and subscription information: http://www.tandfonline.com/loi/uteh20

The Association Between Chronic Exposure to Traffic-Related Air Pollution and Ischemic Heart Disease

Bernardo S. Beckerman ^a, Michael Jerrett ^a, Murray Finkelstein ^b, Pavlos Kanaroglou ^c, Jeffrey R. Brook ^d, M. Altaf Arain ^c, Malcolm R. Sears ^e, David Stieb ^f, John Balmes ^{a g} & Kenneth Chapman ^h

 $^{\rm a}$ School of Public Health , University of California Berkeley , Berkeley , California , USA $^{\rm b}$ Department of Family and Community Medicine , University of Toronto , Toronto , Canada



Perinatal Exposure to Traffic-Related Air Pollution and Atopy at 1 Year of Age in a Multi-Center Canadian Birth Cohort Study

Hind Sbihi, Ryan W. Allen, Allan Becker, Jeffrey R. Brook, Piush Mandhane, James A. Scott, Malcolm R. Sears, Padmaja Subbarao, Tim K. Takaro, Stuart E. Turvey, and Michael Brauer

Motivation

The development of air pollution surfaces is crucial for a better understanding of population exposure

- Live air pollution data can be used to investigate the effect of traffic management on air pollution hot spots (e.g. low emission zones)
- Mobile monitoring campaigns provide unparalleled coverage of an urban area

The rise of portable air pollution monitors coupled with GPS devices



Aeroqual sensor (NO2 and O3)



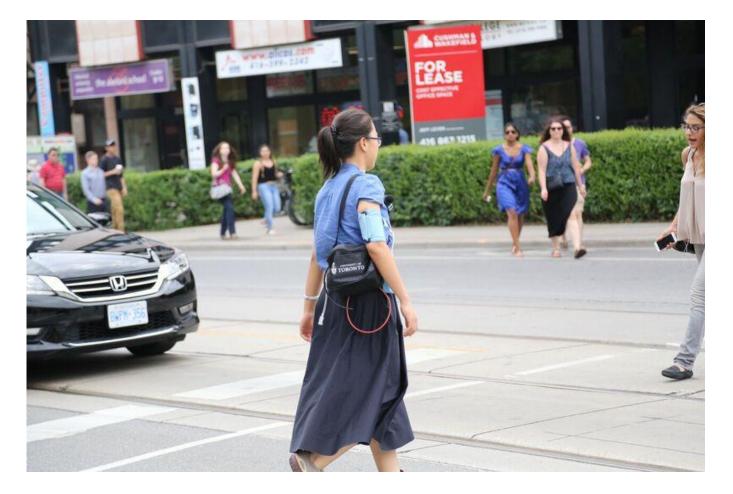
DiscMini (UFP)



MicroAeth (BC)

enables....

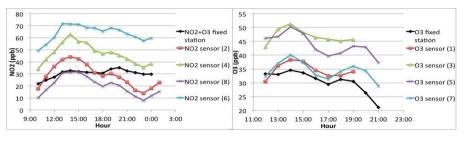
...mobile monitoring and panel studies





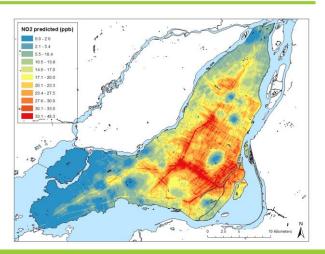
Some hypotheses have not been tested and many questions regarding the role of air pollution sensors remain unanswered

Sensors need calibration





Is air pollution data collected in an ad-hoc way useful to capture the spatio-temporal variability of air pollution in an urban area?



Comparing the performance of various data collection protocols

□ Fixed points



Pedestrians



Cyclists



Panel





Fixed points



Cyclists



Four data collection protocols conducted in

the same campaign







Panel study

Gold standard for measuring exposure

Recruiting participants from the general population

Personal exposure measured throughout the day, monitors are close to the body

Physiological measures conducted to relate with acute health effects

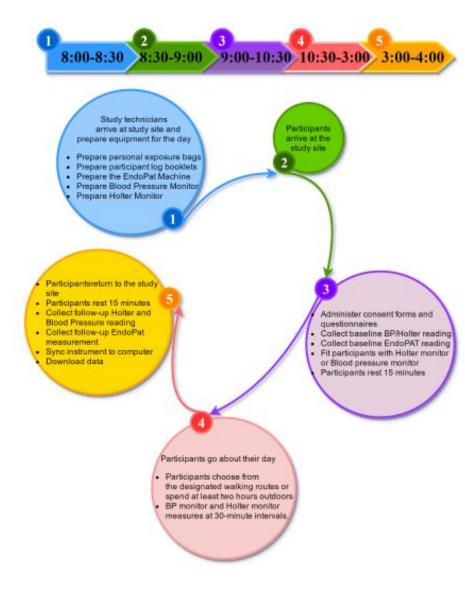
Panel study

Are you a healthy, non-smoking adult between 18-60? Are you willing to participate in a study of traffic related air pollution and health effects? Would you

> consider wearing air pollution monitors and health sensors as you walk around the city on two separate days?

Help us better understand the potential health effects of traffic pollution in Toronto!

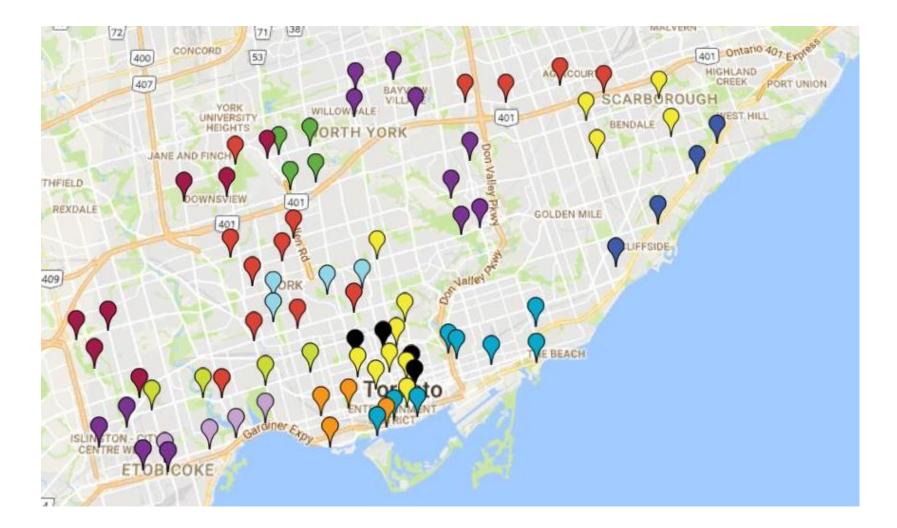
This study was approved by the research ethics board of the University of Toronto For volunteering, please contact airpollution.health.study@gmail.com, alternatively, call 416-458-1737. Compensation would total \$60.



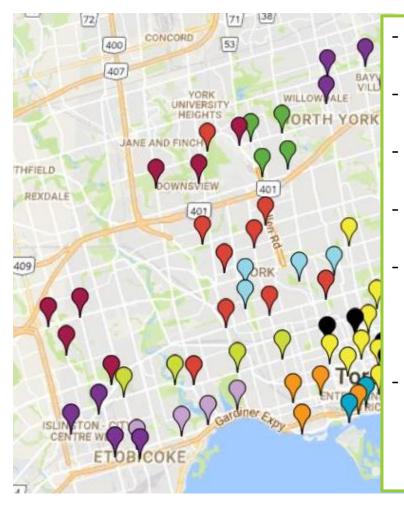
Participants arrive at the study site, undergo baseline health measures, and go about their day, then come back at the end of the day

- Can we predict their exposures without having to conduct personal measurements?
- What is the error associated with potential mis-classification of exposure?

Fixed points



Fixed points

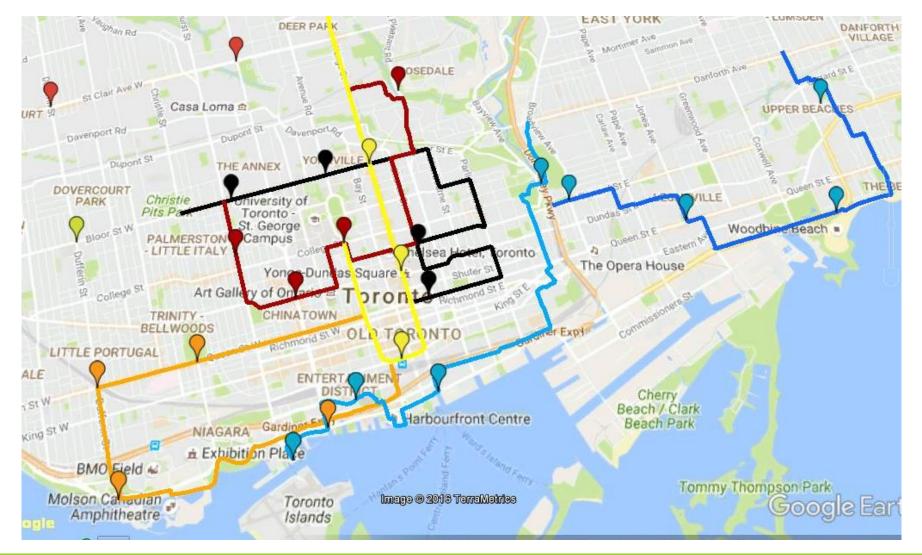


- 96 points
- Intersections and mid-block locations
- 6 visits per point, different times of day
- 20 minutes per visit
 - Traffic counts (passenger car, passenger truck, light duty truck, transit bus, SHT, LHT, school bus, coach bus) and cyclists

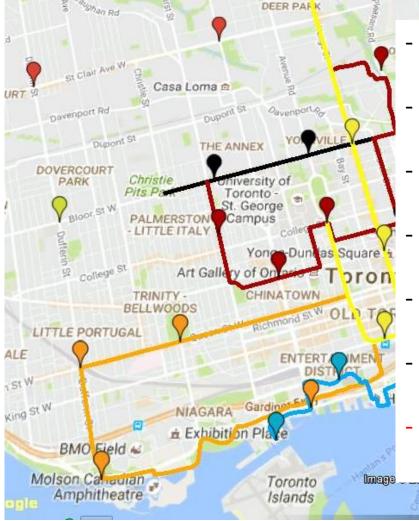
UFP, BC, noise

Time block	Time			
1	7 am to 11 am			
2	11 am to 3 pm			
3	3 pm to 7 pm			

Pedestrian routes designed to overlap with fixed locations



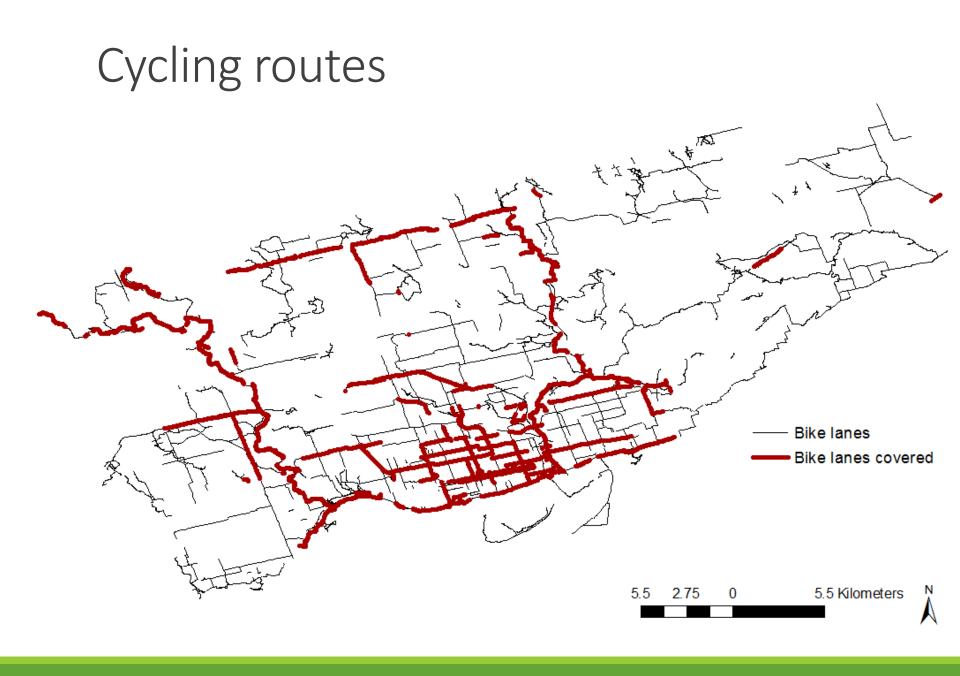
Pedestrian routes designed to overlap with fixed locations

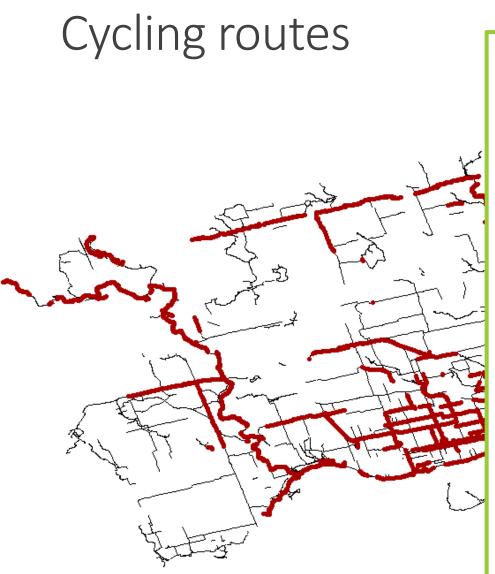


23 routes

- 7 to 10 km each
- Covered of 200Km of unique roads
- Average of 6 repetitions
- Different times of day
- Total of 1080 km
- Data processed per road segment (approx. 4,000 unique road segments)

SEa





- 10 routes, 24 to 31 km each
- 270 km covered in total
- Partial overlap with pedestrian routes
- Each route was repeated 6 to 8 times, at least once per time block
- Total of 1860 km
- Approx. 3,900 road segments

Time block	Time			
1	7 am to 9 am			
2	9 am to 11 am			
3	11 am to 1 pm			
4	1 pm to 3 pm			
5	3 pm to 5 pm			
6	5 pm to 7 pm			

Preliminary analysis of cycling data

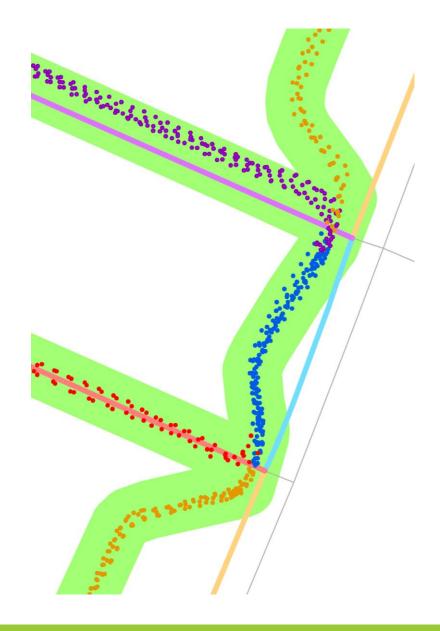
Database

Every GPS point is given a unique ID and associated with:

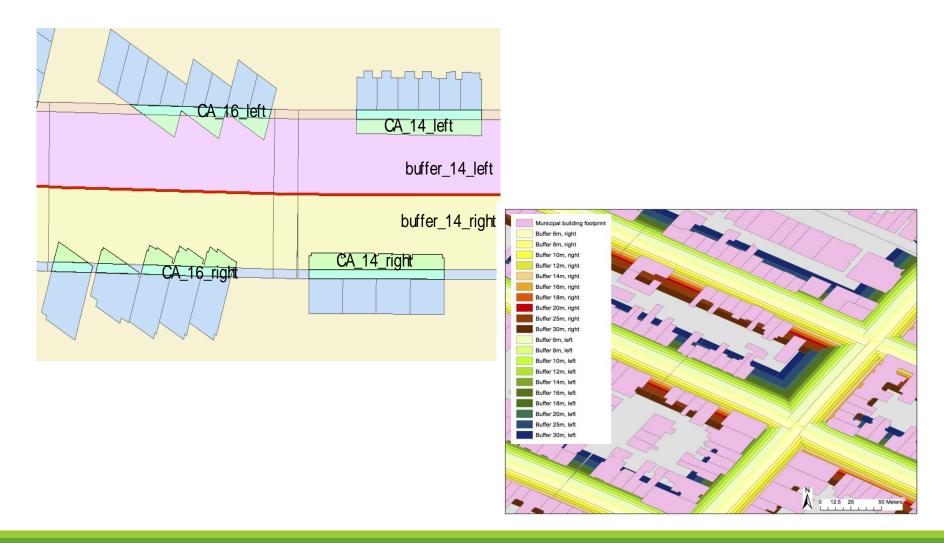
- Air pollution level
- Road segment (approx. 3900)
- Day
- Time
- Meteorology (wind speed, direction, RH, temperature)
- Average air pollutant concentration per segment per visit is the outcome variable (UFP, BC, noise)

Coefficient of variation for each segment/visit

Allocating GPS points to road segments



Land-use and built environment around each road segment



Land-use and built environment around each road segment



List of LU + BE characteristics

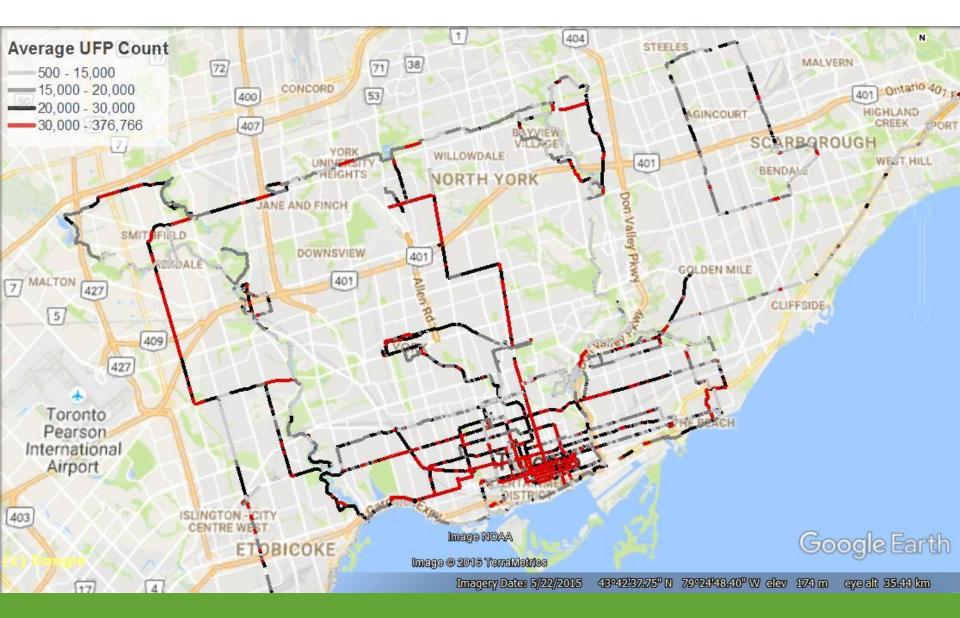
Buffers of 25, 50, 100, 200, 300, 500, 1000m

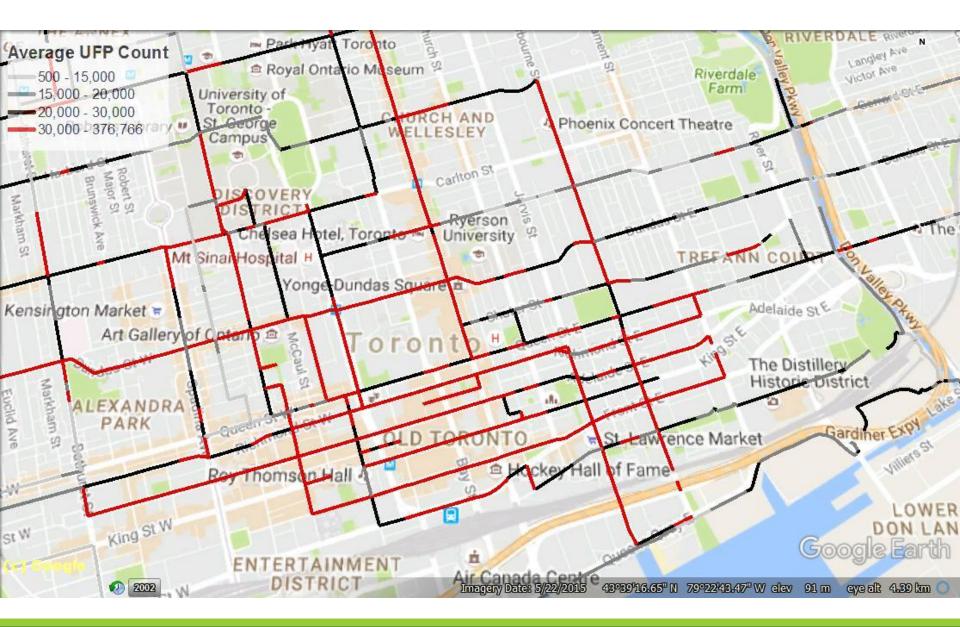
- Distance from the shore (m) (d_shore)
- Distance from the closest railline (m) (d_railline)
- Distance from the closest major road (m) (d_majrd)
- Distance from the closest highway (m) (d_highway)
- Distance from the closest airport (m) (d_airport)
- Distance to the closest NOx emitting chimney (m) (d_NPRI_NOx)
- Distance to the closest PM emitting chimney (m) (d_NPRI_PM)
- Area of the buildings (m2) (build_25m to build_1000m)
- Area of the commercial land use (m2) (com_25m to com_1000m)
- Area of the governmental and institutional land use (m2) (gov_25m to gov_1000m)
- Area of the resource and industrial land use (m2) (ind_25m to ind_1000m)
- Area of the open area land use (m2) (open_25m to open_1000m)
- Area of the parks land use (m2) (park_25m to park_1000m)
- Area of the residential land use (m2) (resid_25m to resid_1000m)
- Area of the waterbody land use (m2) (water_25m to water_1000m)
- Length of the bus routes (m) (busline_25m to busline_route_1000m)
- Length of the major roads (type 4) (m) (majrd_25m to majrd_1000m)
- Length of the highways (types 1, 2 and 3) (m) (highway_25m to highway_1000m)
- Length of the roads (types 1, 2, 3, 4, 5 and 6) (m) (roads_25m to roads_1000m)
- Number of bus stops (count) (bus_25m to bus_1000m)
- Number of intersections (count) (inter_25m to inter_1000m)
- Number of trees (count) (trees_25m to trees_1000m)
- Population (count) (pop_500m to pop_1000m)
- Average height of buildings (m) (build_height_25m to build_height_100m)
- Maximum height of buildings (m) (max_build_height_25m to max_build_height_100m)
- Number of NOx emitting chimneys (count) (NPRI_NOx_25m to NPRI_NOx_1000m)
- Number of PM emitting chimneys (count) (NPRI_PM_25m to NPRI_PM_1000m)
- Length of rail lines (m) (rai_25m to rail_1000m)

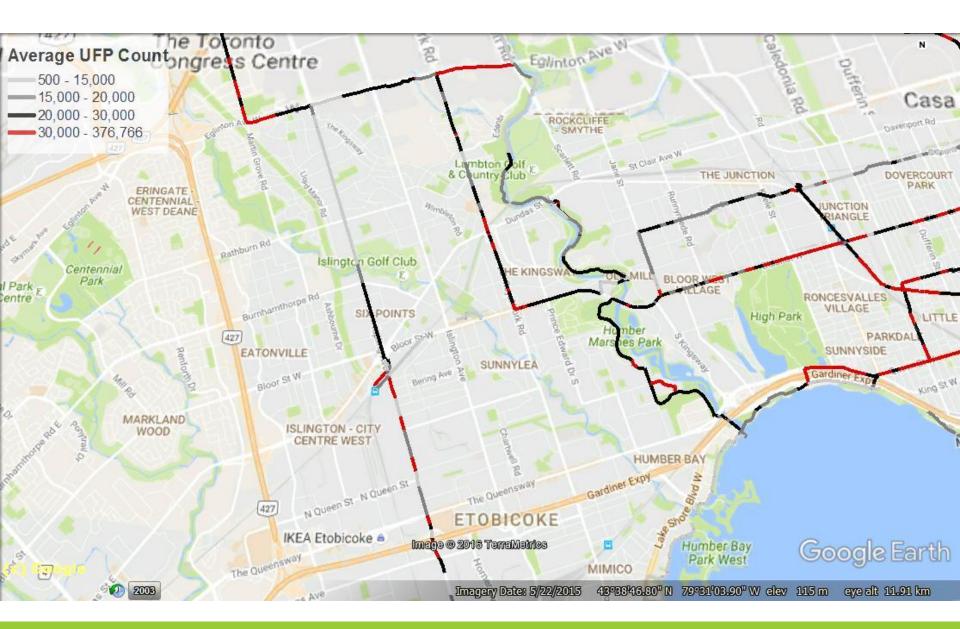
Descriptive statistics (by segment)

	Mean	Std Dev	Min	Percentiles			
				25 th	50 th	75 th	100 th
Average ultrafine particles (particles / cm ³)	23436	15837	500	14447	19969	28603	376766
Average Black Carbon (ng/m ³)	1761	2839	15	757	1235	1822	103046
LAeq (dB)	72.83	4.06	55	71	73	75	89

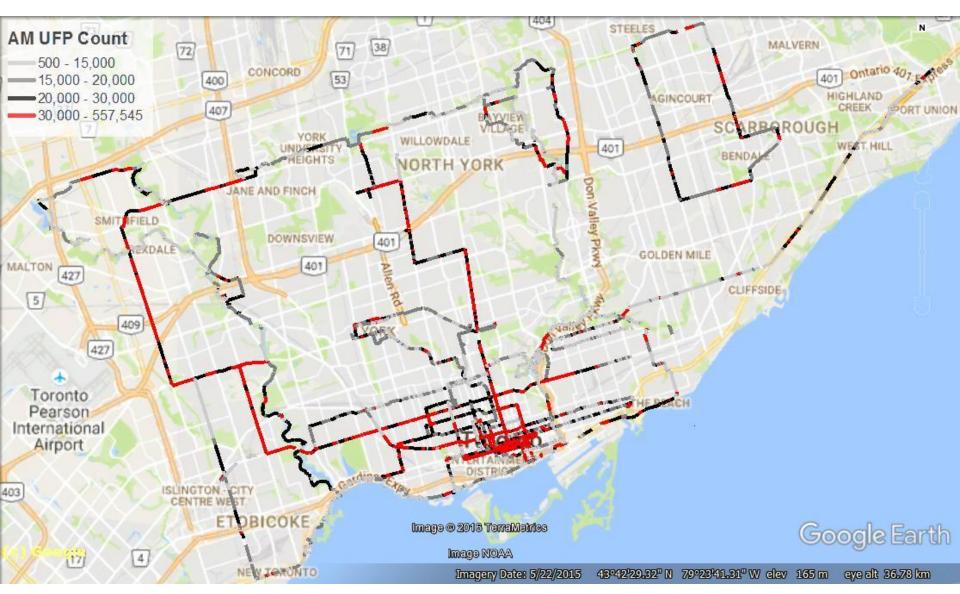
UFP along bicycle facilities



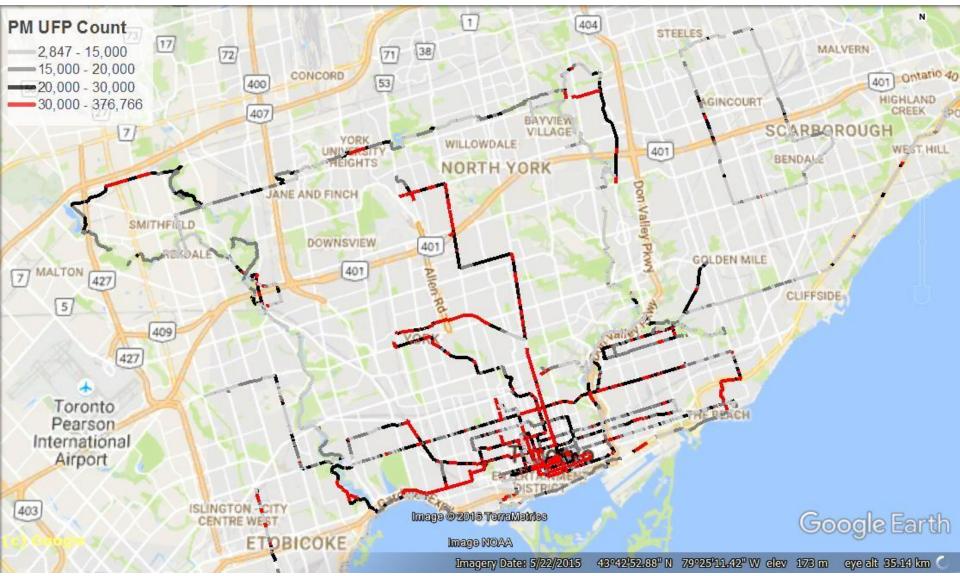




AM



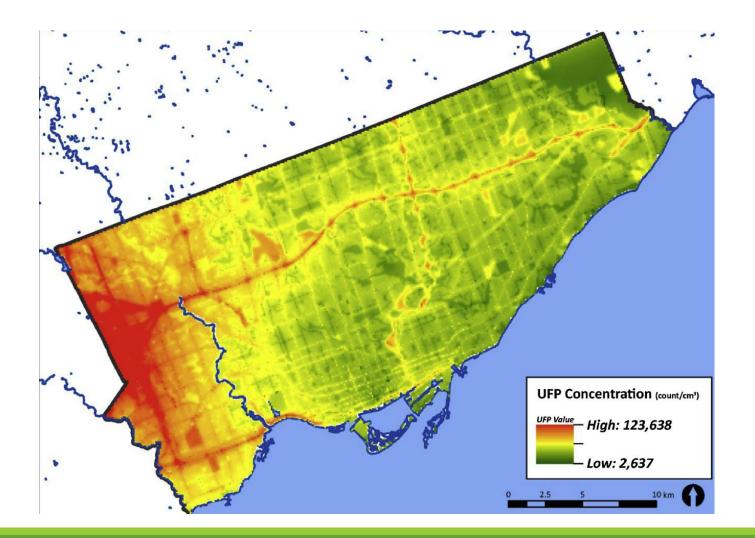
PM



Early models of air pollution along cycling facilities

	For LN of UFP average							
	for increase of IQ if not otherwise indicated							
	Mean Change	95% CI for Mean	Mean Change					
Model		Change		Mean Change				
<mark>.</mark> A state of the		ffect model	GLM					
		del 35	Model 36					
AIC		310.50	4487.440					
Adjusted R ²	0.	2591	0.3566					
Windspeed	-0.26	-0.27, -0.24	-12.15	-13.17, -11.11				
timeblock - A (6, 7 and 8) -	1		NA					
Reference								
timeblock - B (9 and 10)	-0.32	-0.35, -0.28						
timeblock - C (11 and 12)	-0.07	-0.12, -0.03						
timeblock - D (13 and 14)	0.04	0.00, 0.08						
timeblock - E (15 and 16)	-0.30	-0.34, -0.26						
timeblock - F (17, 18 and 19)	-0.31	-0.35, -0.26						
Relative Humidity	-0.11	-0.13, -0.09	-8.65	-9.90, -7.37				
d_airport	-0.15	-0.17, -0.14	-12.21	-14.10, -10.28				
build_1000m	0.12	0.10, 0.14	11.95	9.53, 14.43				
park_1000m	-0.019	-0.036, -0.003						
max_build_height_25m	0.06	0.03, 0.08						
d_NPRI_NOx	-0.05	-0.07, -0.02						
trees_750m	0.11	0.08, 0.13	14.93	11.39, 18.59				
d_shore	-0.05	-0.06, -0.03	-8.00	-9.82, -6.16				
d_majrd	-0.017	-0.021, -0.014	-3.07	-4.00, -2.14				
gov_1000m	-0.014	-0.024, -0.003						
open_1000m	0.04	0.03, 0.05	4.63	2.93, 6.35				
highway_25m	0.0008	0.0005, 0.0011	0.12	0.07, 0.16				

Developing predictive models



What is the potential of human-centric air pollution sensing in producing reliable air pollution maps?

What is the potential of human-centric air pollution sensing in producing reliable air pollution maps?

To evaluate the potential of embedded network of monitors in the generation of urban air quality maps useful for quantifying population exposure patterns

