

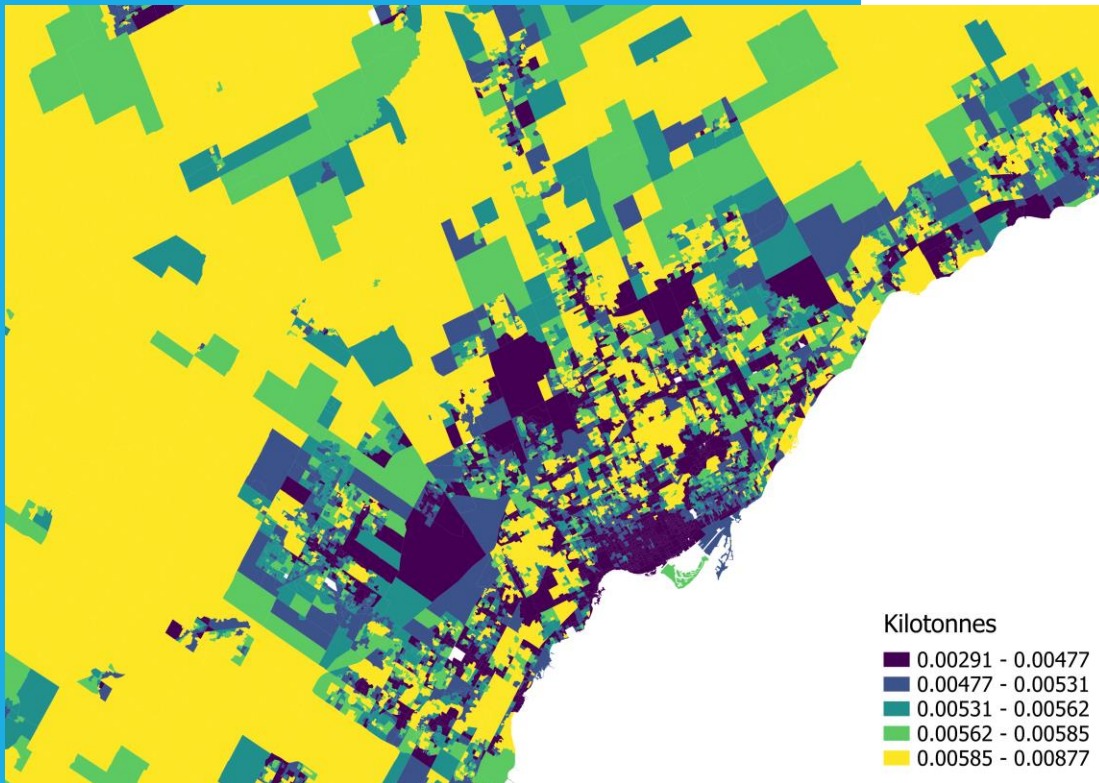
# **CANADIAN URBAN INSTITUTE INTERNSHIP**

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# MAIN PROJECT: HOUSEHOLD GREENHOUSE GAS (GHG) MAP



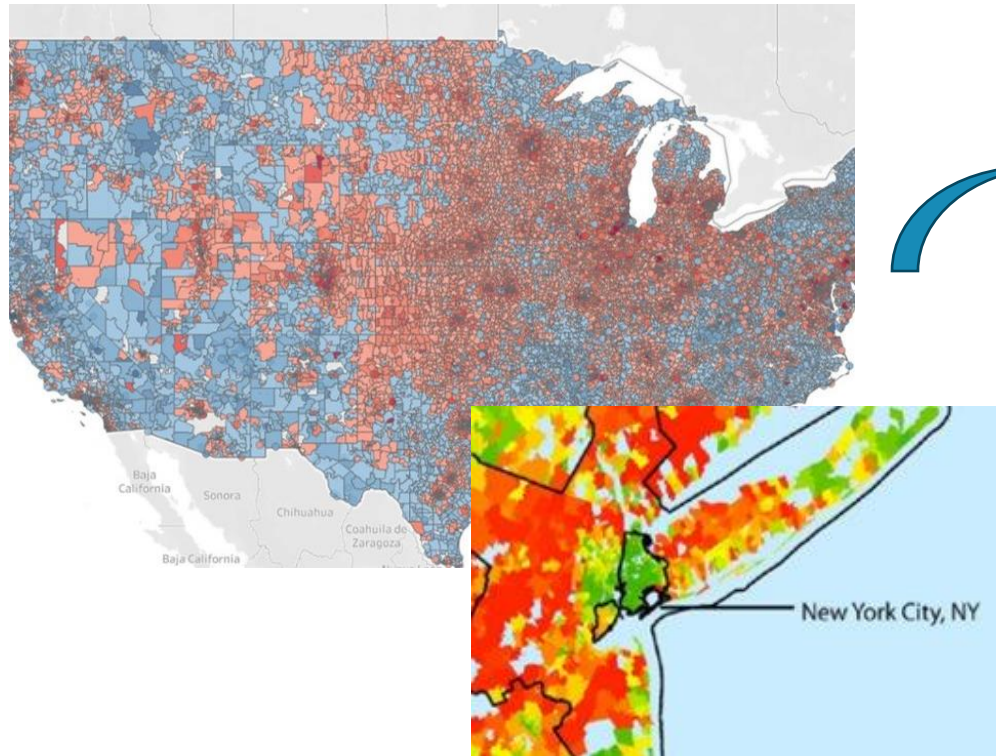
Modelling, mapping, and making sense of trends in the GHG emissions of household activity in Canada





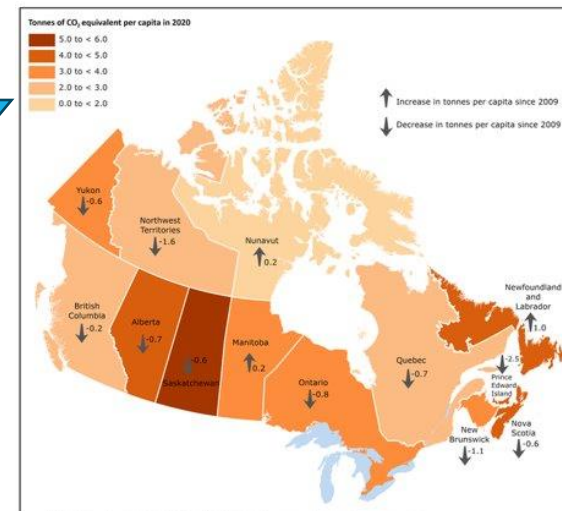
# BACKGROUND

## USA Household GHG model by UC Berkeley, 2013

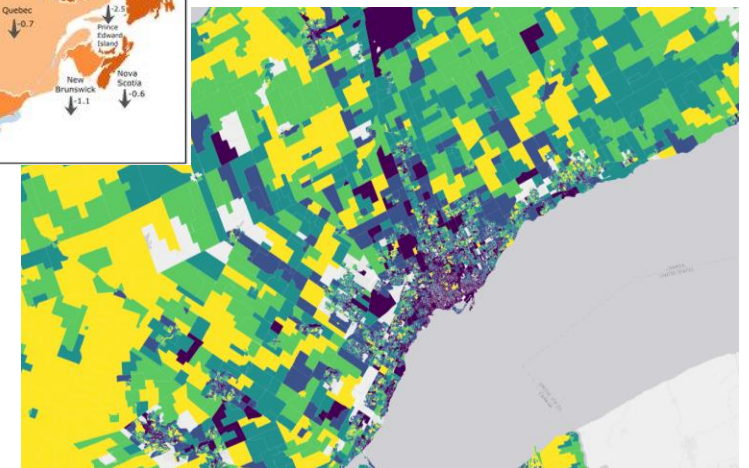


Finding: cities have **lower household GHG** than suburbs

## Canadian model of household GHG



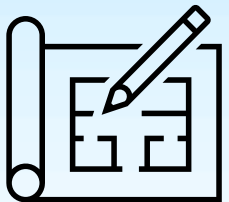
Move from provincial GHG to DA level emissions



# Project Goals

- ☐ Create a nationwide map of household GHG emissions, specified at a local level
- ☐ Model the emissions based on urban form (ex. housing type) and demographic (ex. income, commute time) variables
- ☐ Create a situational tool to compare GHG based on housing composition

Overall aim: demonstrate environmental benefits of cities/urban living (in comparison to suburbs)

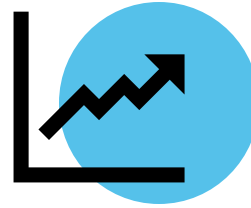


# MODEL PROCESS



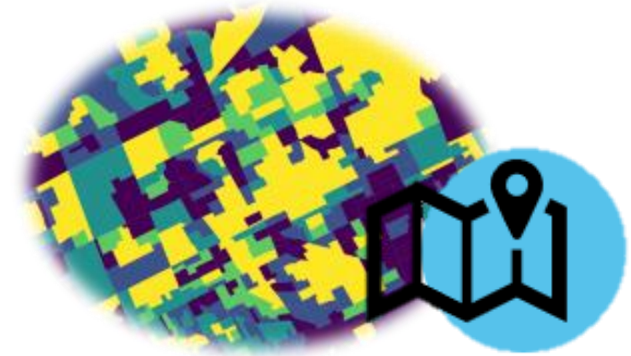
## RESEARCH & PLANNING

- Review original paper
- Narrow scope
- Identify key variables
- Find data sources



## MODEL BUILDING, ANALYSIS

- Clean and merge datasets
- Build linear regression model in R
  - Try different predictor variables
  - Adjust training data
  - Compare coefficients,  $R^2$  value
  - Identify inconsistencies, improve model



## MAPPING & REVISION

- Predict each DA's average household GHG based off model
- Map results in QGIS, check patterns
- Revise model



# Household GHG Emissions

**Energy consumption (EC)**  
Electricity, natural gas



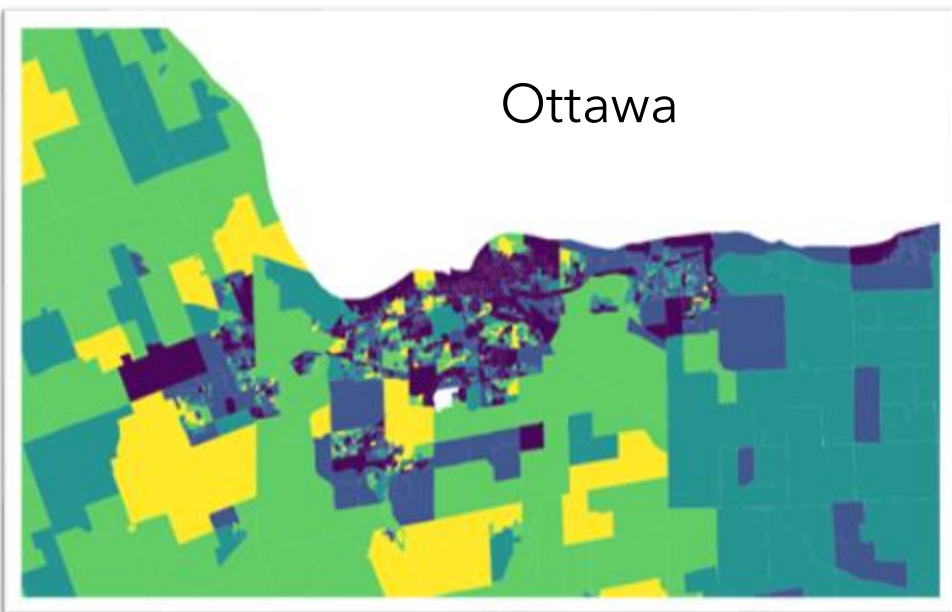
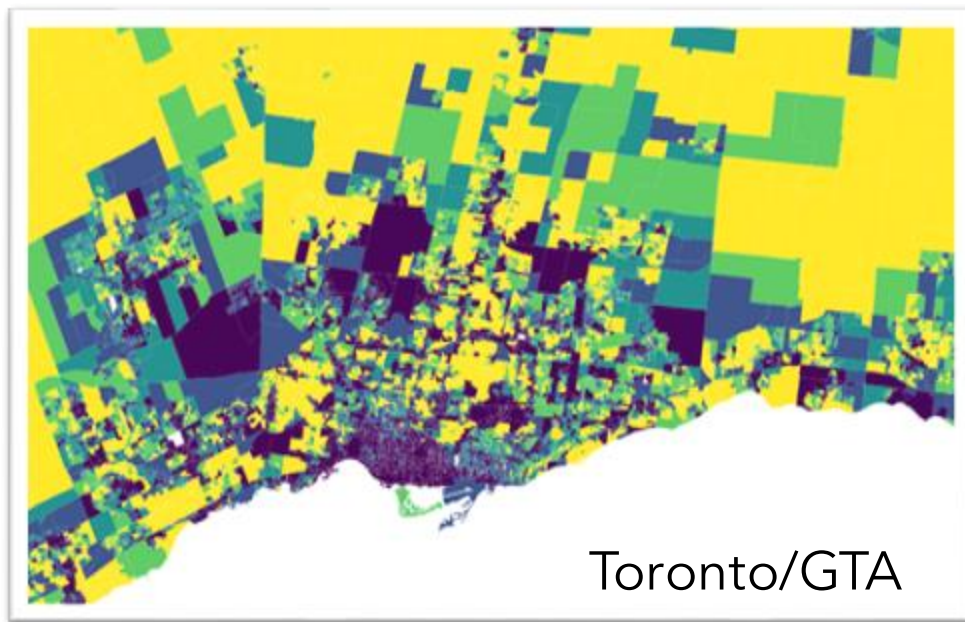
**Vehicle use**  
Motor fuels



Model emissions by linear regression

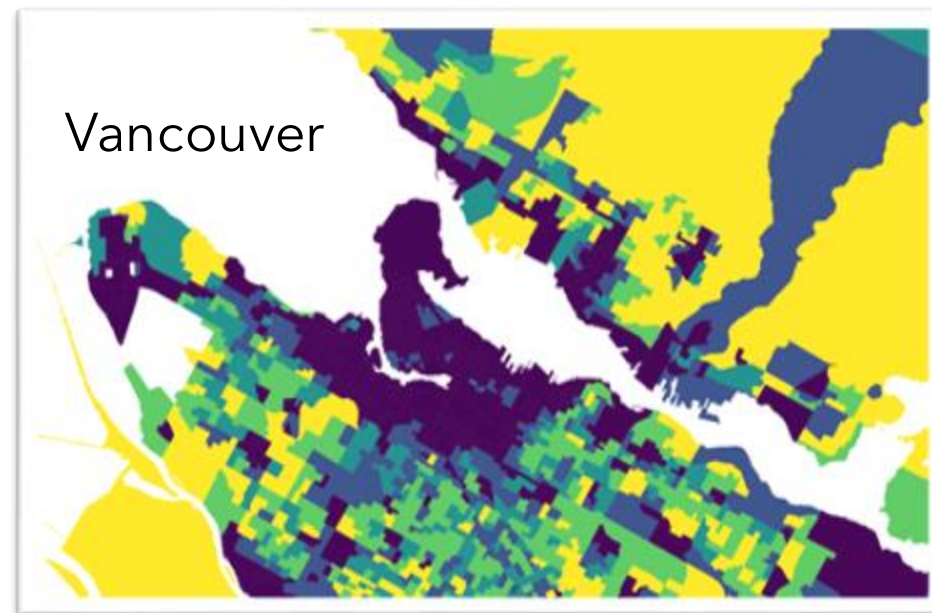
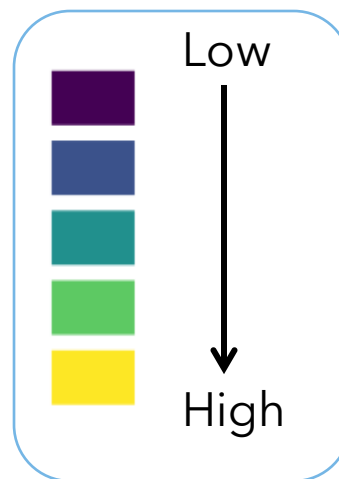
## Model Equation:

$$\begin{aligned} \text{Household GHG Emissions} = & \beta_0 + \beta_1(\text{Income}) + \beta_2(\text{Population density}) + \\ & \beta_3(\text{Commute}) + \beta_4(\% \text{ Houses}) + \beta_5(\% \text{ Apartments}) + \beta_6(\# \text{ vehicles}) + \\ & \beta_7(\text{Age of home}) + \beta_8(\text{Carbon intensity electricity}) + \beta_9(\text{Gas price}) + \\ & \beta_{10}(\text{Natural gas price}) \end{aligned}$$



## MAPS

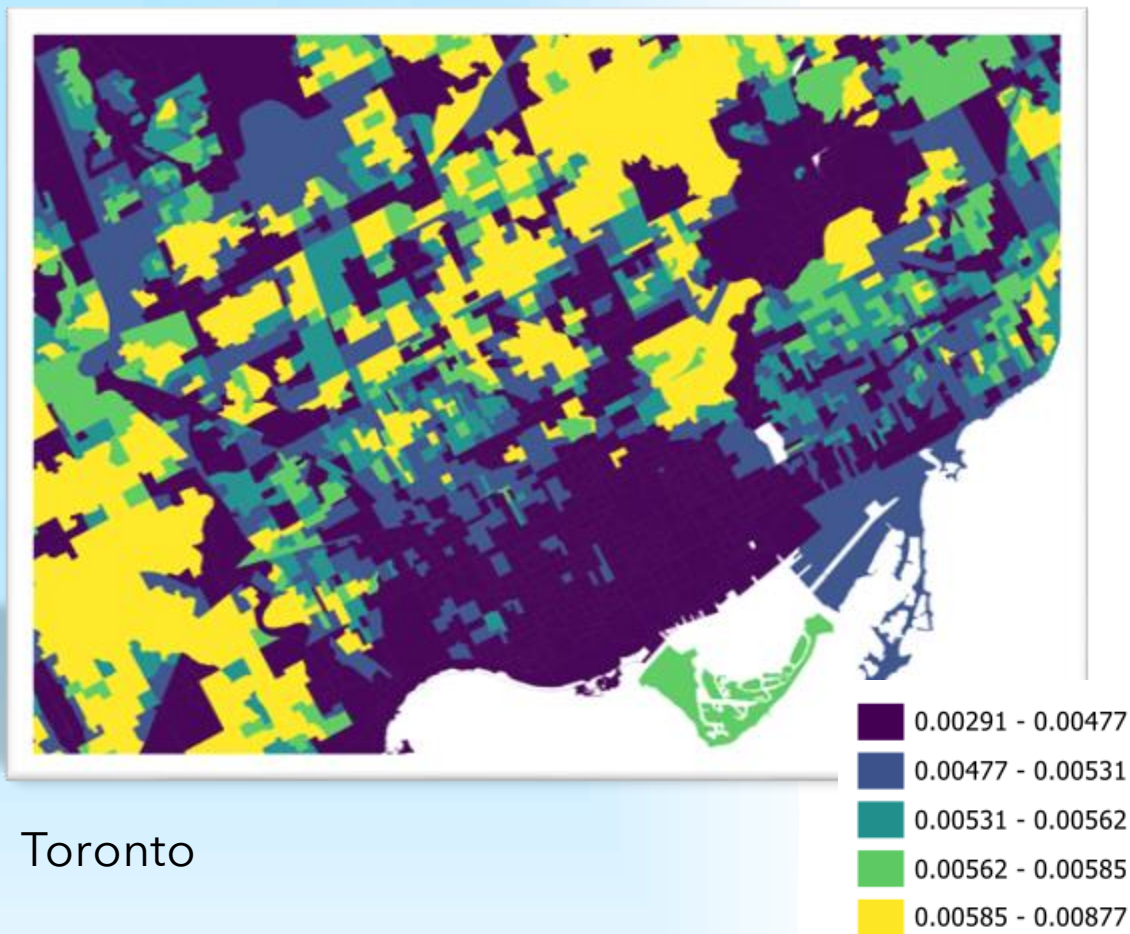
GHG Emissions  
(Kilotonnes)





## Key Findings

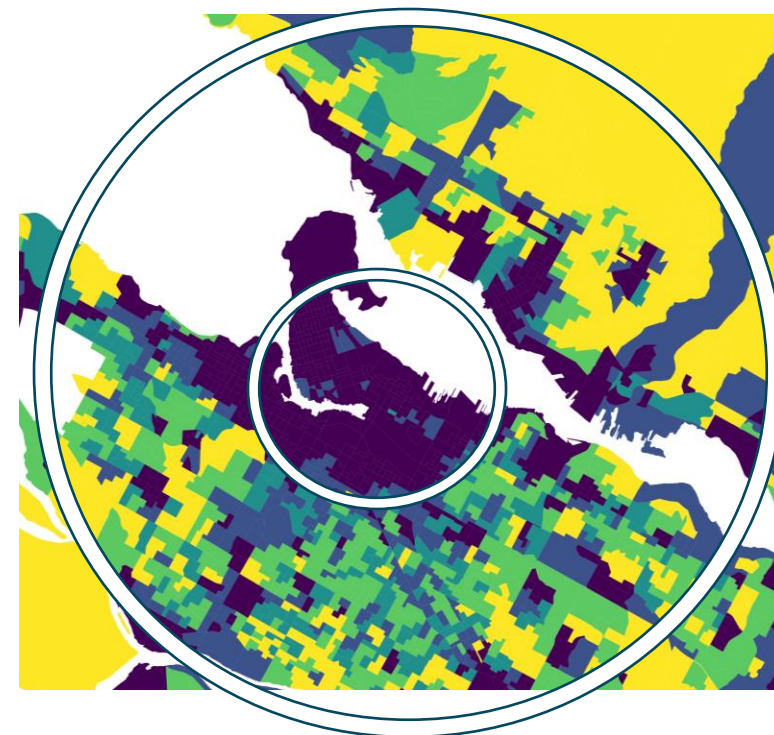
- Downtown areas : **lower household emissions**  
(smaller dwellings, public transit, etc.)
- Suburban areas : **higher household emissions**  
(larger homes, longer commute, etc.)



Toronto

- **Ring of higher emissions**  
around city core

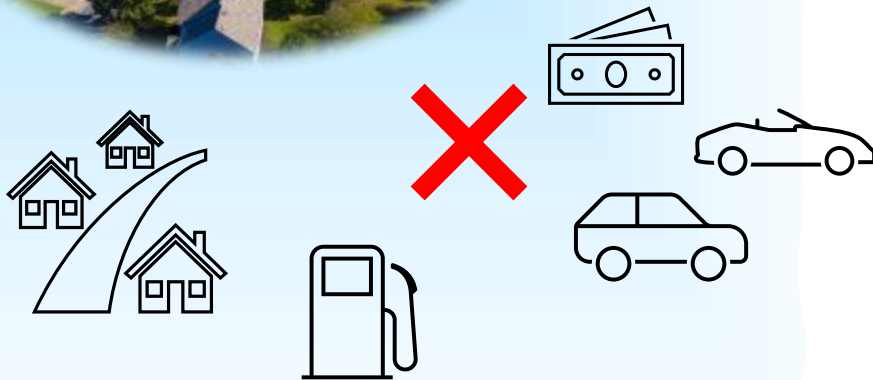
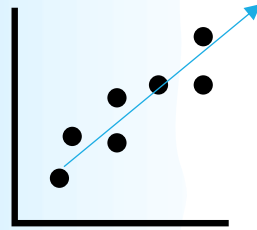
Vancouver



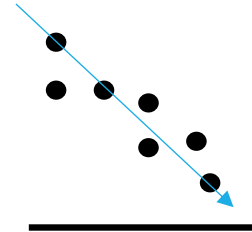


# Key Findings

- Income, commute time, % houses, # vehicles and home age are **positively correlated** with GHG



- % apartments and gas prices are **negatively correlated** with GHG



# Interactive tool

## Goal

Develop a tool to demonstrate how changes in housing type will affect GHG emissions per household

## How does it work?

1. Choose a location (i.e. DA number)
2. Extract original demographic values
3. Calculate original GHG emission
4. Modify housing data by:
  - Changing # houses
  - Changing # apartments
5. Calculate new GHG emission
6. Present difference in GHG emission due to direct changes in housing type





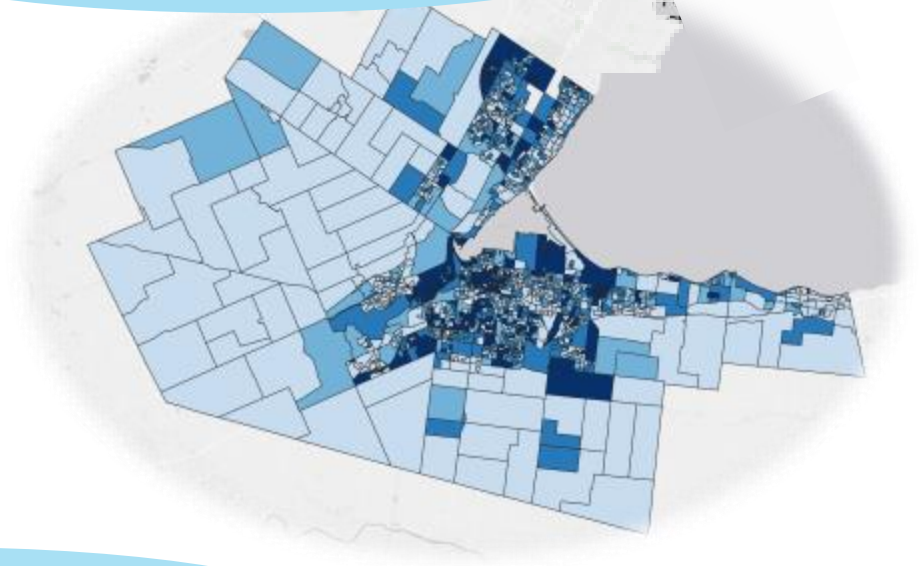
# ADDITIONAL PROJECTS



Business GHG Emissions

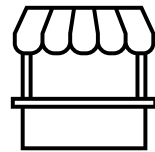
- 250+ maps each of CMA data

Demographic Mapping



Local Business Index

- National dataset of businesses



- Distinguish chain vs. local

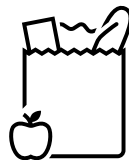
- Frequency of name



0.22  Toronto

0

Local

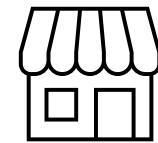


- Index of "localness"

- Food vs retail

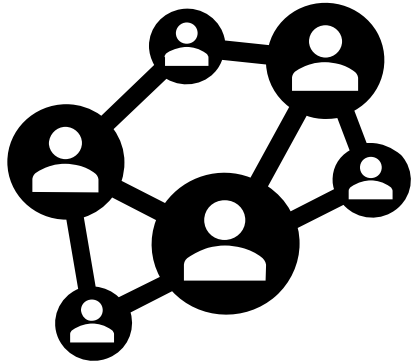
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Chain



# ADDITIONAL PROJECTS

## Social connection



- Multinomial logistic regression model
- Examine the relationships between urban form variables (independent variables) and social connection variables (dependent variables)

Local personal contacts  
Sense of belonging



Income \$  
Density  
Housing type  
Immigrants  
Visible minorities  
Education  
Main mobility type







# TAKEAWAYS

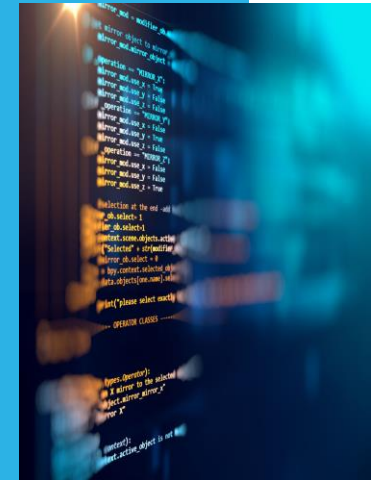
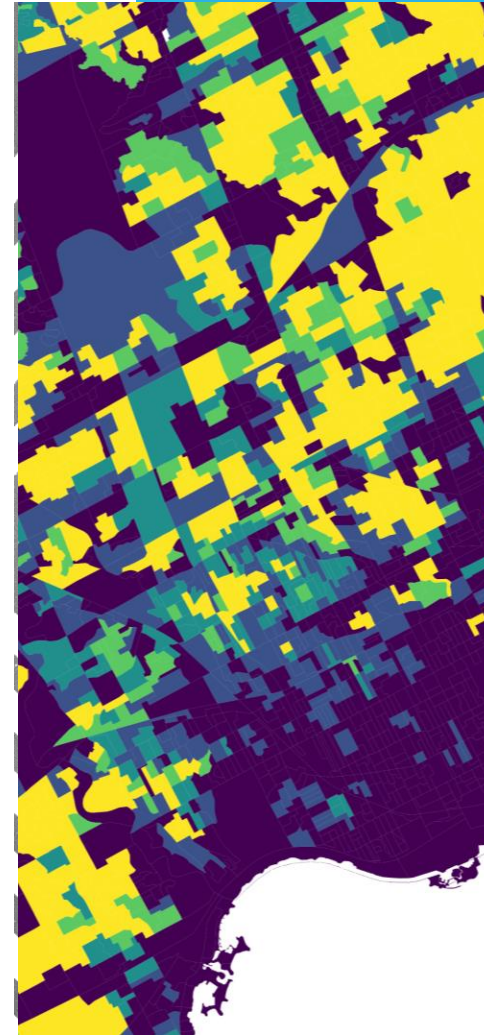
## EXPLORATION, EXPERIENCE, & ENRICHMENT

### Data science skills

- Data cleaning, management, and analysis
- Programming in R
- Data visualization
- Regression analysis
- Excel, MS Office
- QGIS

### Transferable skills

- Communication
- Organization
- Collaboration
- Critical thinking
- Immersion in workplace
- Participated in meetings
- Presentations
- Observation
- Work-life balance







**THANK YOU**