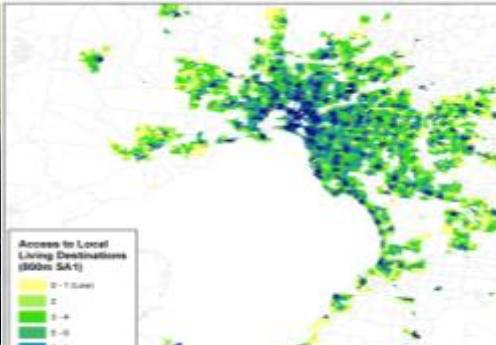


# Applications of GIS in public health

**Dr Suzanne Mavoa**

Senior Research Fellow, NHMRC Early Career Research Fellow  
Melbourne School of Population and Global Health

May 2020



# Outline

1. Introduction to public health
2. Examples of GIS applications in public health
3. Data challenges

# **Public health**

**“the art and science of preventing disease, prolonging life  
and promoting health through the organized efforts of  
society”**

Acheson, D (1988) Public Health in England: The Report of the Committee of Inquiry into the Future Development of the Public Health Function. London: HMSO

# Public health

Disease surveillance/monitoring

Risk assessment

Health service access

Community health

Health promotion

Health programs and interventions

planning, monitoring, implementing,  
advocating/promoting, policy making, researching...

# Epidemiology

“Epidemiology is the study of the distribution and determinants of health-related states or events in specified populations, and the application of this study to the control of health problem”

Last JM, editor. Dictionary of epidemiology. 4th ed. New York: Oxford University Press; 2001. p. 61.

# **Focus on health and place underpins the roots of public health**

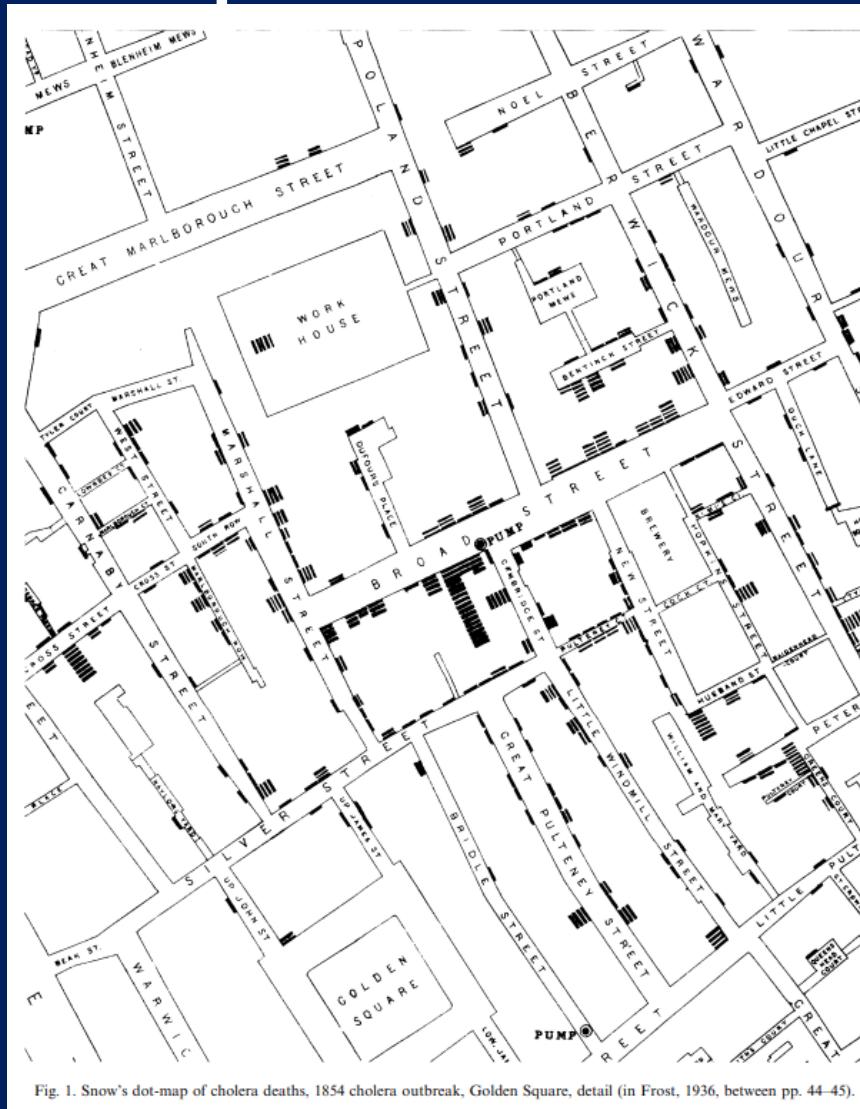
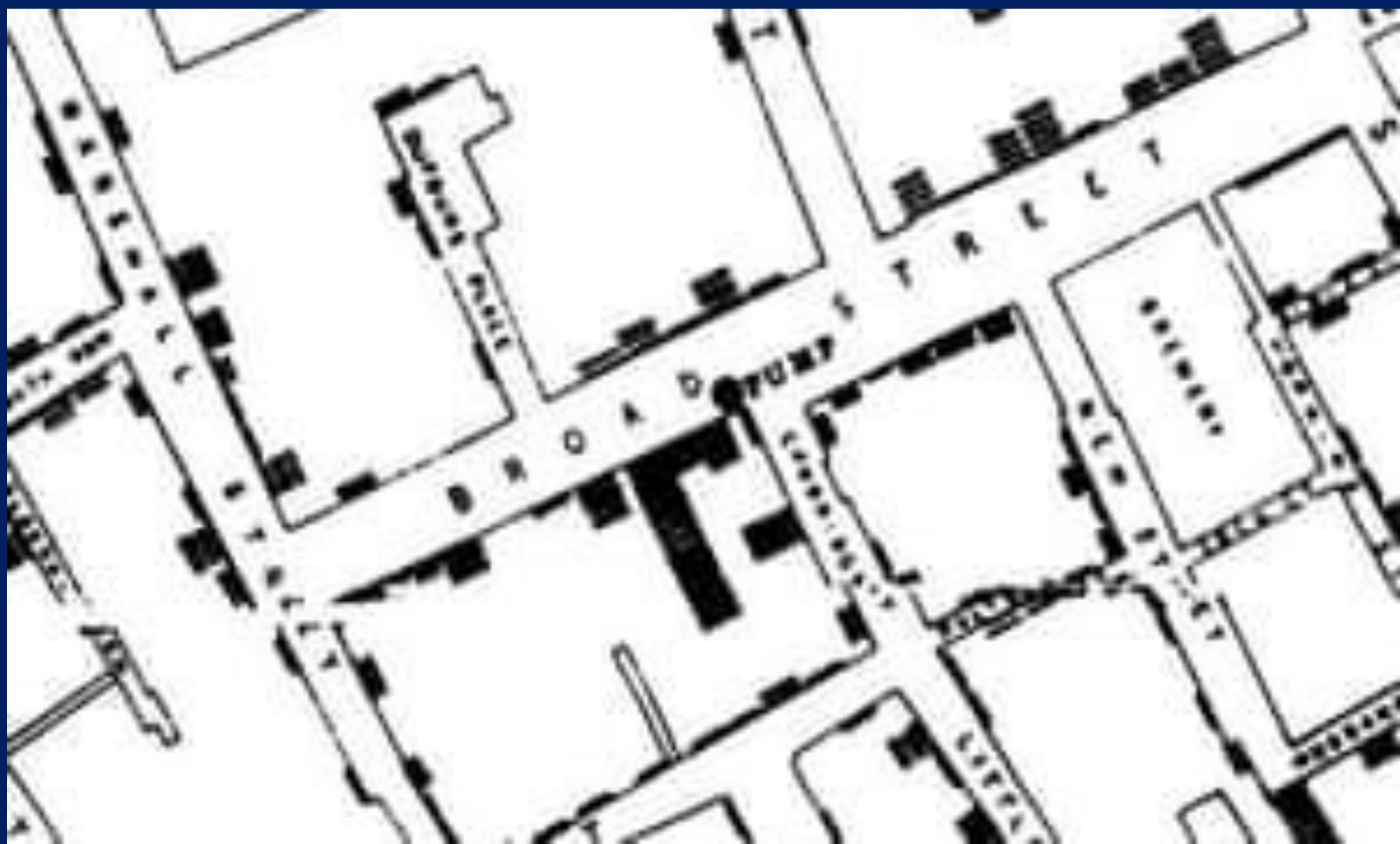
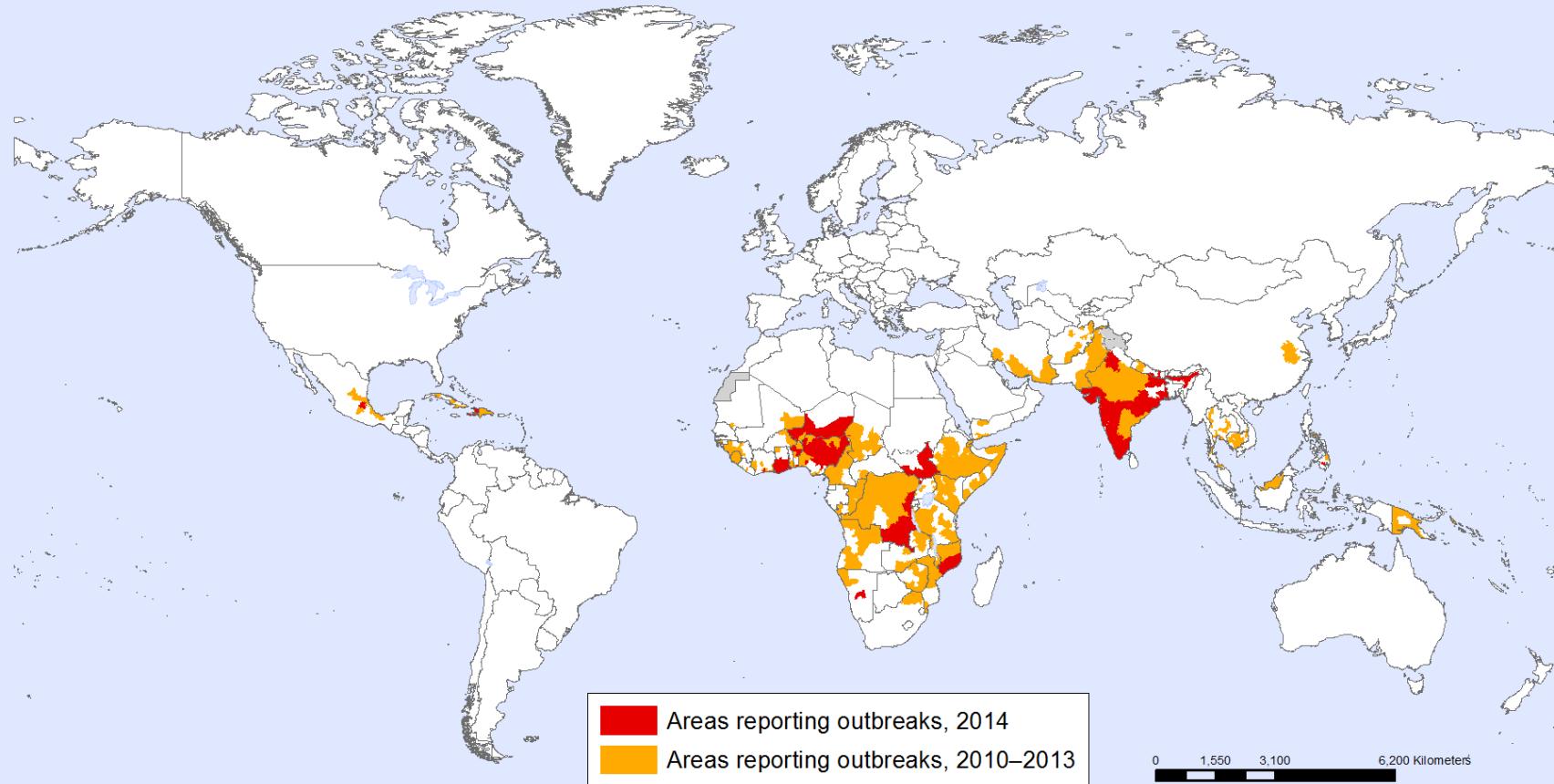


Fig. 1. Snow's dot-map of cholera deaths, 1854 cholera outbreak, Golden Square, detail (in Frost, 1936, between pp. 44–45).

Snow, John (1855). *On the Mode of Communication of Cholera* (2nd ed.). London: John Churchill.



## Cholera, areas reporting outbreaks, 2010–2014



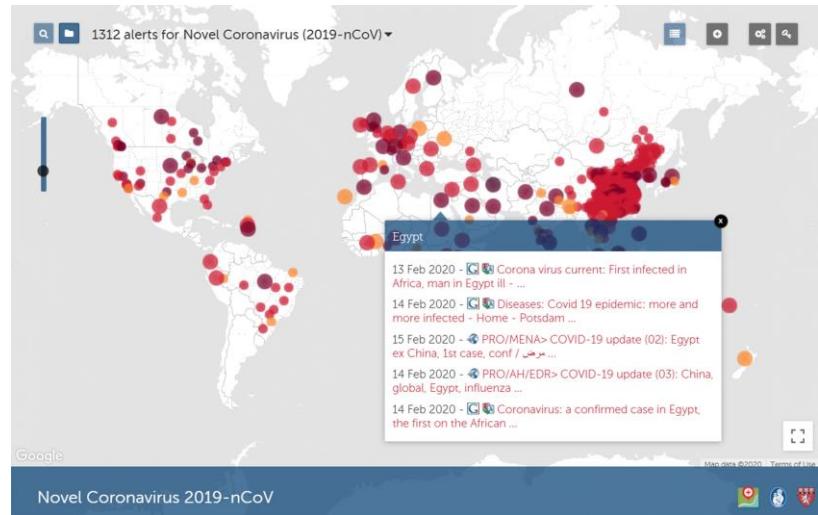
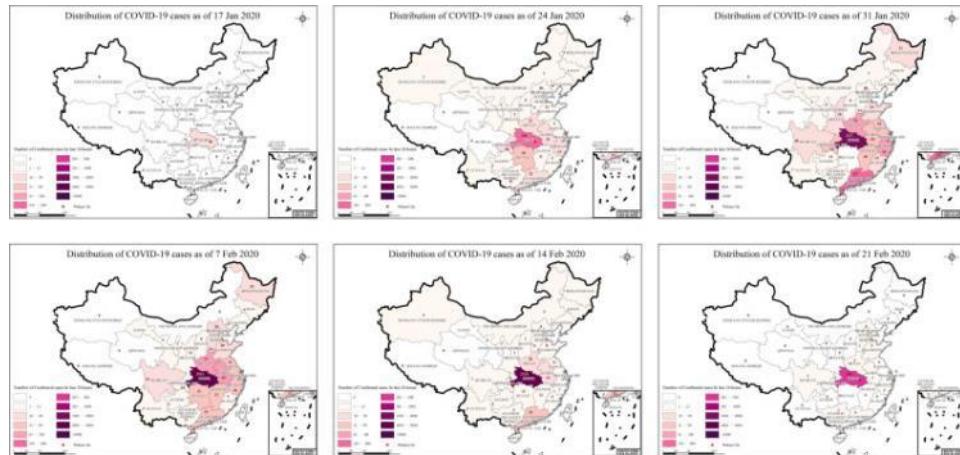
The boundaries and names shown and the designations used on this map do not imply the expression of any opinion whatsoever on the part of the World Health Organization concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. Dotted and dashed lines on maps represent approximate border lines for which there may not yet be full agreement.

Data Source: World Health Organization  
Map Production: Health Statistics and Information Systems (HSI)  
World Health Organization



© WHO 2015. All rights reserved.

# COVID-19



Images sourced from:

Boulos, M.N.K. and Geraghty, E.M., 2020. Geographical tracking and mapping of coronavirus disease COVID-19/severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) epidemic and associated events around the world: how 21st century GIS technologies are supporting the global fight against outbreaks and epidemics.

Zhou, C., Su, F., Pei, T., Zhang, A., Du, Y., Luo, B., Cao, Z., Wang, J., Yuan, W., Zhu, Y. and Song, C., 2020. COVID-19: challenges to GIS with big data. *Geography and Sustainability*.

Google.

# Determinants of health

Individual

Social

Environmental

# Determinants of health

Individual

Social

Environmental

biology

behaviour

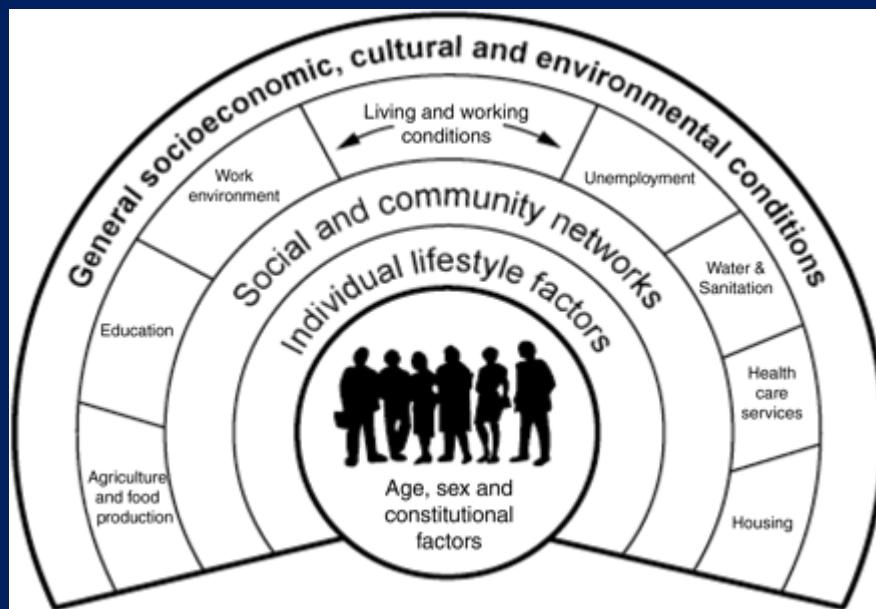


# Determinants of health

Individual

Social

Environmental



Source: Dahlgren G, Whitehead M., *Policies and strategies to promote social equity in health.*, 1991 Stockholm Institute for Future Studies

# Determinants of health

Individual  
chemicals/pollutants



natural environment



Social

Environmental

built environment



# (Geo)spatial questions related to health

Where are the disease outbreaks? Where are the clusters/hotspots of disease?

Where are the at risk populations? Where should we intervene?

What health services are available in different areas of a city/country?

Where are the gaps in health services?

Are the habitats/climate that X species needs present in this environment?

How well do different neighbourhoods enable walking/physical activity?

What healthy food is available in different neighbourhoods?

What characteristics of environments determine health?

Where are people spending time? What environments are they exposed to?

# A GIS/geospatial toolbox for public health

Collect/create data

Map/visualise

Spatial analysis

# GIS applications in health – part 1

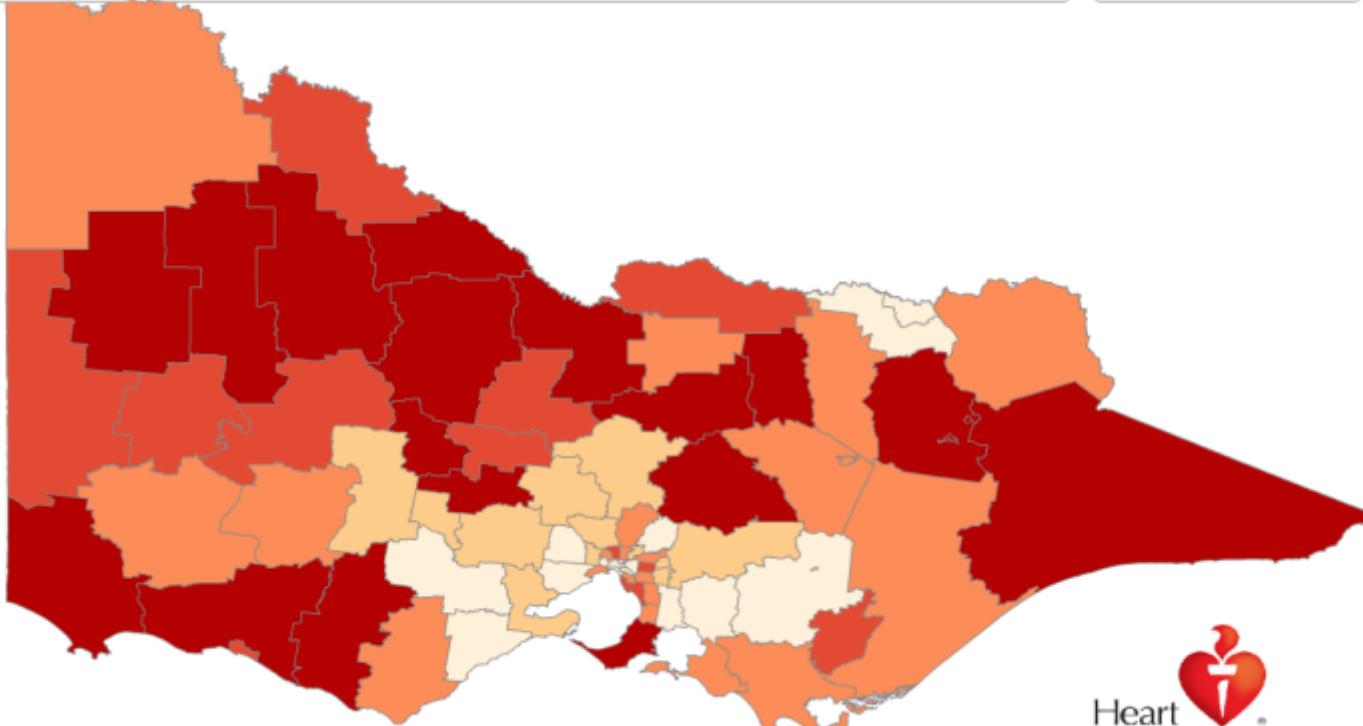
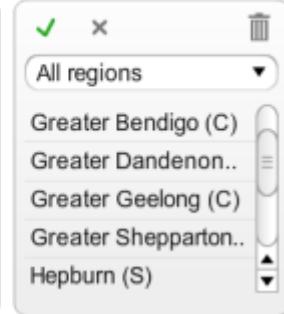
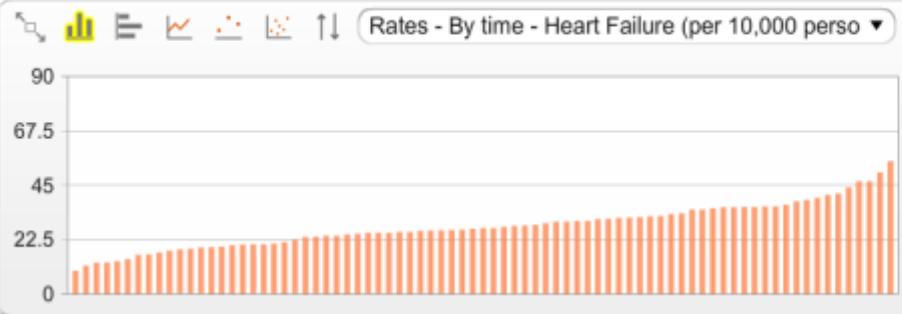
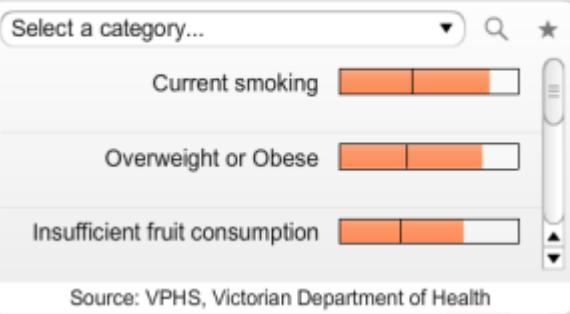
- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- Identifying spatial patterns, e.g. clusters of disease
- Spatial statistics, spatial regression, spatial interpolation
- Measure the environment (generate/test hypotheses )
- Assess resource allocation/accessibility (e.g., health services, schools, water)
- Estimate exposure to the environment (e.g, using GPS)

# GIS applications in health – part 2

- Mapping perceptions/qualitative data
- Simulating spatial behaviour (e.g, disease transmission, walking)
- Measuring health-related policy implementation
- Targeting interventions
- Assessing the impact of interventions
- Assisting in research design and data collection (e.g., identifying study area, design of sample frames )
- Communication/knowledge translation/community engagement

# **GIS applications in health– part 1**

- **Geographic distribution and variation of disease (incidence/prevalence)**
- **Surveillance and monitoring of disease (infectious and chronic)**

[Rates](#)[SMRs](#)[Risk Factors](#)[Socio-demographics](#)

## Australia's Health Tracker by Area, 2016

Illness (estimates), 2014–15 - Age-standardised rate per 100 >> Diabetes, people aged 25 to 64 years

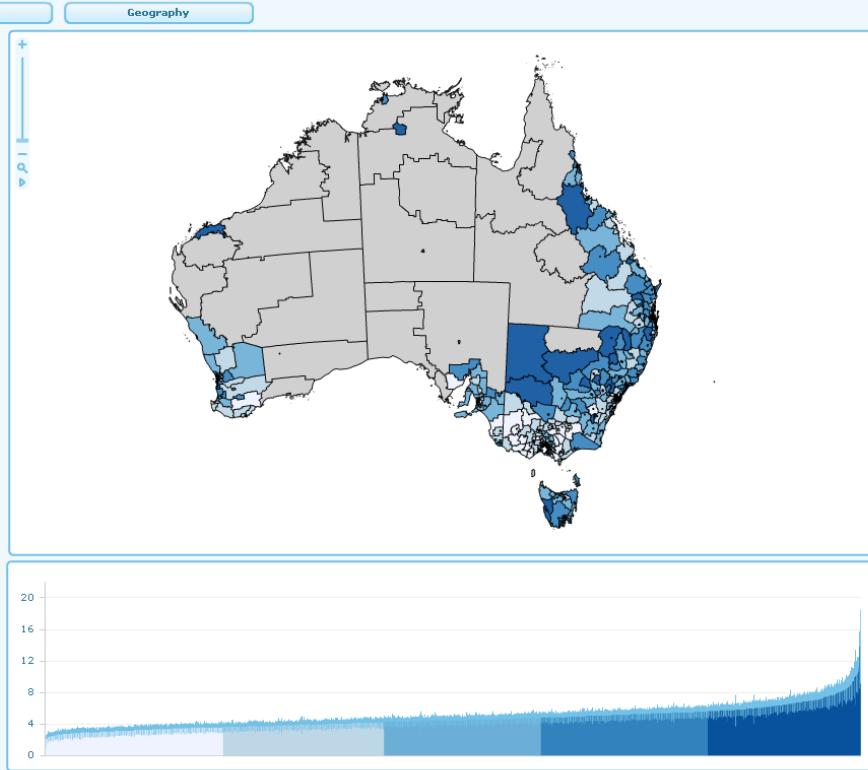


Select indicator	Name	Age-standardised rate	Number
ACT: ACT - East/ Hume/...		5.9	47
ACT: Belconnen North		3.6	515
ACT: Belconnen South		3.3	310
ACT: Belconnen West		4.4	554
ACT: Chifley/ Lyons (AC...		4.3	185
ACT: Cotter - Namadgi		10.6	109
ACT: Curtin/ Garran/ Hu...		2.9	182
ACT: Dunlop/ Gooromon...		4.5	278
ACT: Fadden/ Gowrie (A...		2.7	240
ACT: Farrer/ Isaacs/ Ma...		2.8	212
ACT: Flynn (ACT)/ Frase...		3.4	221
ACT: Greenway/ Oxley (...		4.0	271
ACT: Gungahlin - North		5.0	401
ACT: Gungahlin - South		6.5	587
ACT: Inner North Canber...		3.5	563
ACT: Inner North Canber...		5.7	451
ACT: Inner South Canbe...		1.8	61
ACT: Inner South Canbe...		3.0	355
ACT: Kambah		3.2	304
ACT: Ngunnawal/ Nicholls...		4.1	493
ACT: Tuggeranong - Fria...		4.2	409
ACT: Tuggeranong - South		3.9	653
ACT: Weston Creek		2.8	335
NSW: Adamstown - Kotara		3.9	306
NSW: Albion Park - Macq...		5.0	381
NSW: Albion Park Rail/ S...		5.5	547
NSW: Albury - East		4.1	290
NSW: Albury - North/ Lav...		5.3	686
NSW: Albury - South/ Al...		3.8	495

Note: The extent to which these estimates may vary is shown by the upper and lower limits. These can be seen by hovering the mouse pointer over the bar chart.

^ modelled estimates not produced for Very Remote areas, Aboriginal communities where population is less than 1,000; for other data, replaces numbers where total population is less than 100

RRMSE: ~use with caution; ~~unreliable for general use; \* unreliable; not shown



Name	Age-standardised rate
TARGET	4.1
Australia	4.7
All Greater Capital Cts..	4.5
All Rest of States/ NT	4.6
New South Wales	5.0
Greater Sydney	5.1
Rest of NSW	5.0
Victoria	4.0
Greater Melbourne	4.1
Rest of Vic.	3.9
Queensland	4.6
Greater Brisbane	4.6
Rest of Qld	4.6
South Australia	4.3
Greater Adelaide	4.3
Rest of SA	4.5
Western Australia	4.6
Greater Perth	4.6
Rest of WA	4.7

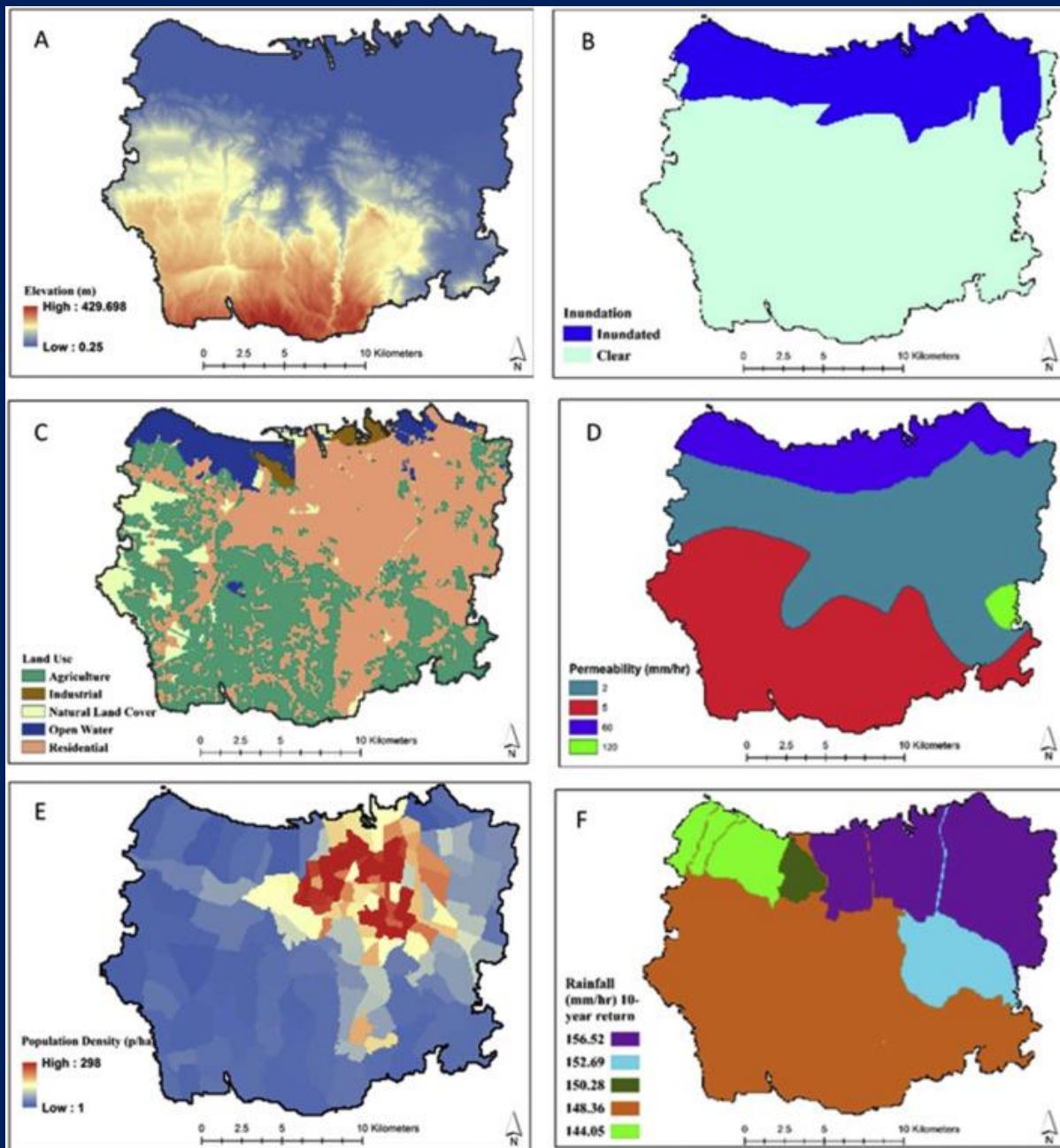


**Australia Health Tracker (Australian Health Policy Collaboration):**  
<http://www.atlasesaustralia.com.au/ahpc/atlas/atlas.html?indicator=i18>

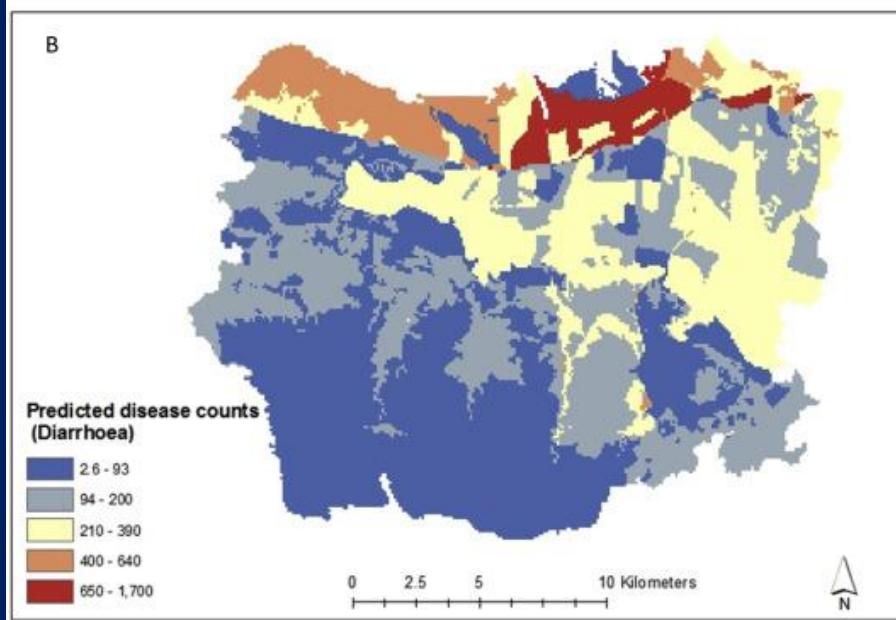
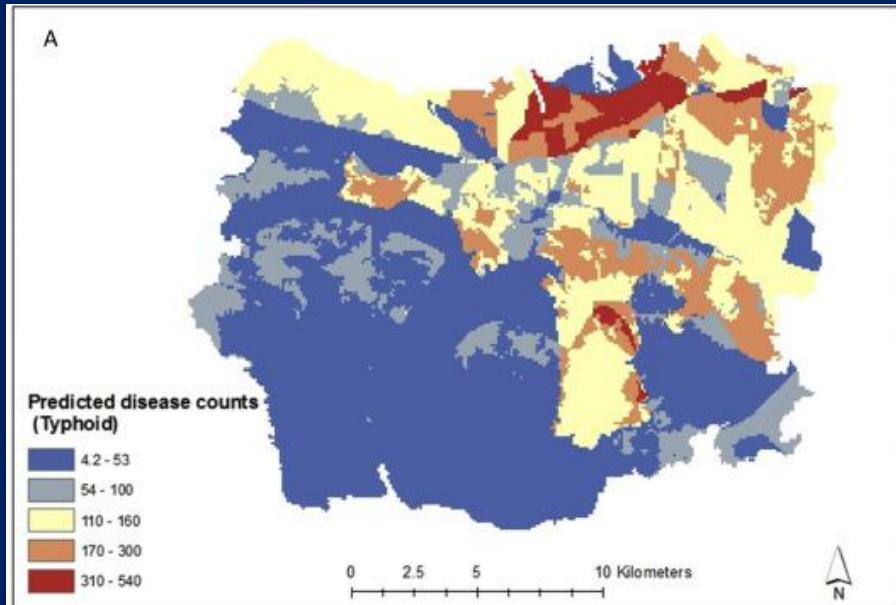
# GIS applications in health – part 1

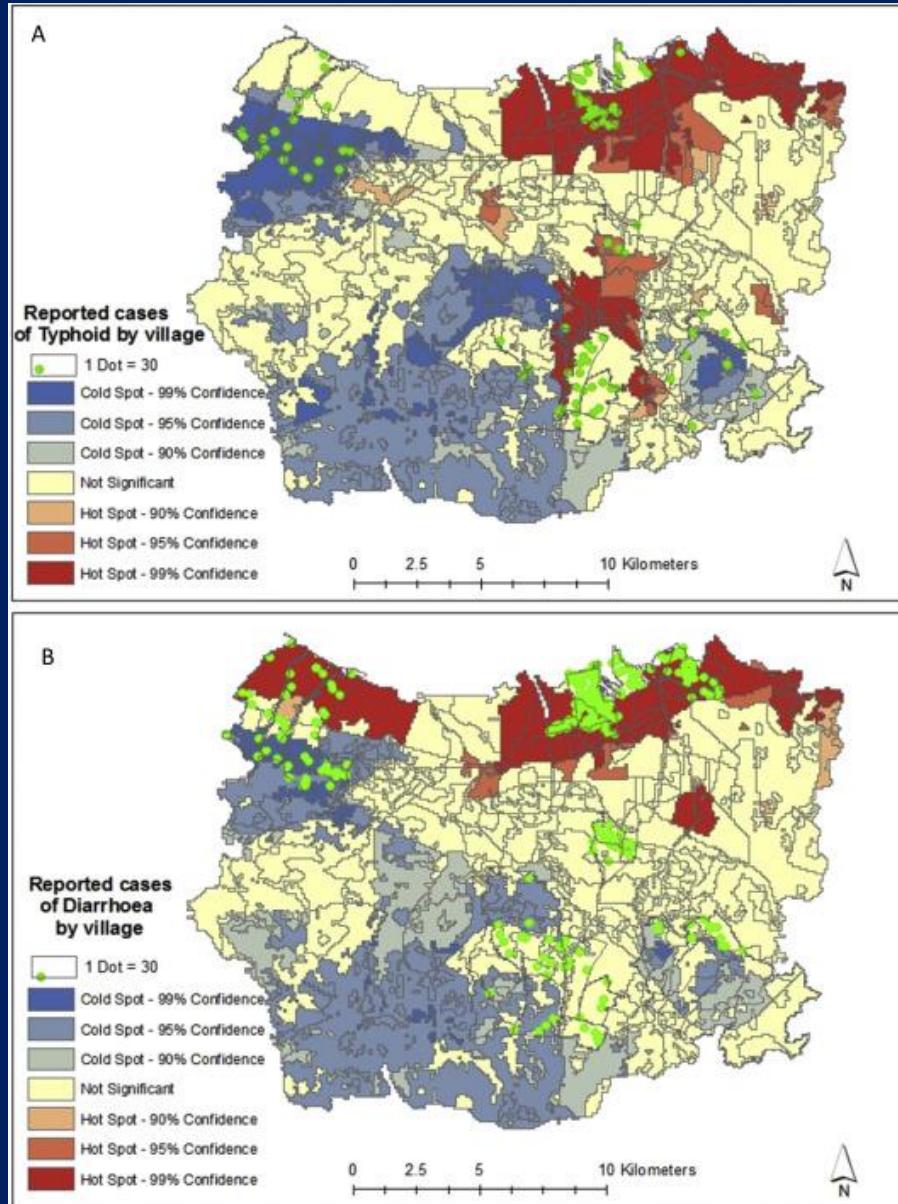
- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- **Identifying at risk populations**

Deilami, K., Hayes, J.F., McGree, J. and Goonetilleke, A., 2017.  
**Application of landscape epidemiology to assess potential  
public health risk due to poor sanitation.** *Journal of  
Environmental Management*, 192, pp.124-133.



Deilami, K., Hayes, J.F., McGree, J. and Goonetilleke, A., 2017. Application of landscape epidemiology to assess potential public health risk due to poor sanitation. *Journal of Environmental Management*, 192, pp.124-133.





# GIS applications in health – part 1

- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- **Identifying spatial patterns, e.g. clusters of disease**

# Lyme disease cluster analysis

Lantos, P.M., Tsao, J., Nigrovic, L.E., Auwaerter, P.G., Fowler, V., Ruffin, F., Foster, E. and Hickling, G., 2017, January. **Geographic Expansion of Lyme disease in Michigan, 2000-2014**. In *Open Forum Infectious Diseases* (p. ofw269). Oxford University Press.

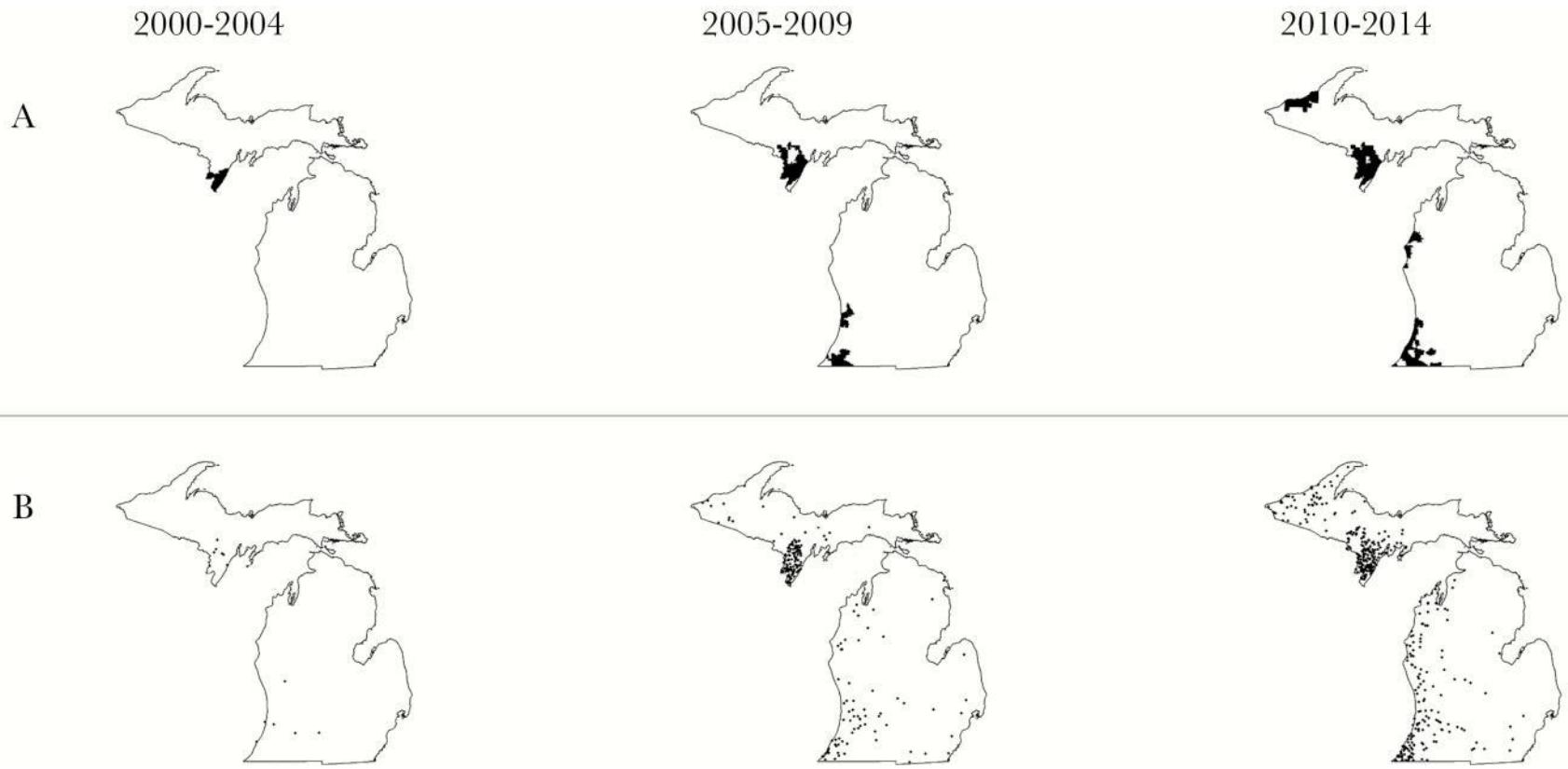
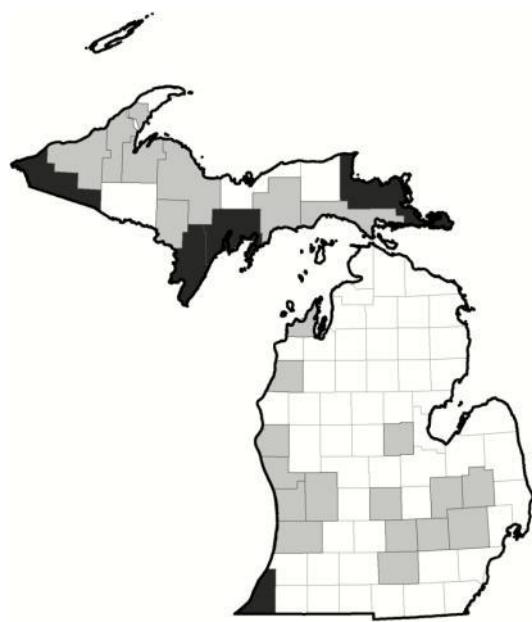


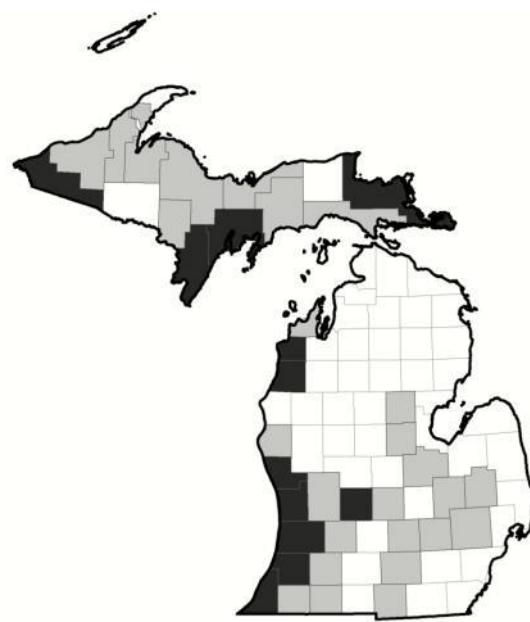
Figure 2. (A) Sequential cluster analysis of Lyme disease cases, grouped by 5-year interval.

Lyme disease clusters

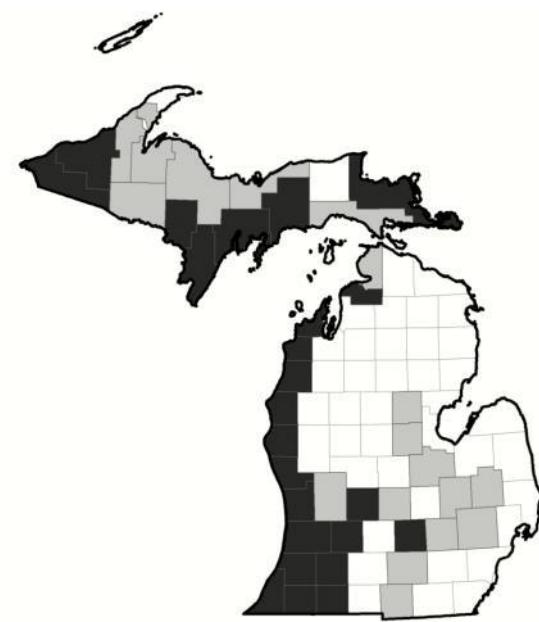
# Deer tick population



Dennis et al, 1998  
5 established  
22 reported



2007 (this study)  
12 established  
26 reported



Eisen et al, 2016  
24 established  
18 reported\*

\*Includes 2 not reported by Eisen

Established      Reported

Figure 3. Expanding distribution of established and reported *Ixodes scapularis* populations in Michigan, 1998–2016

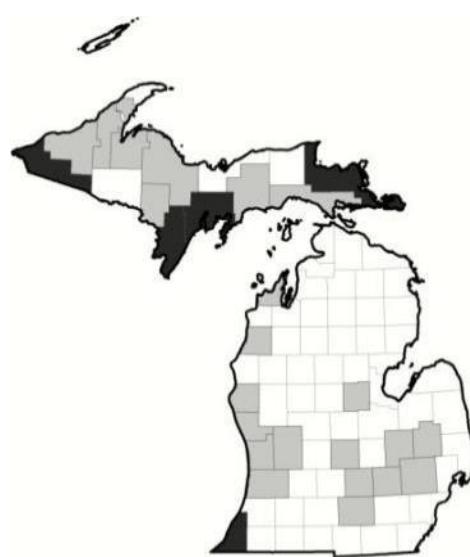
2000-2004



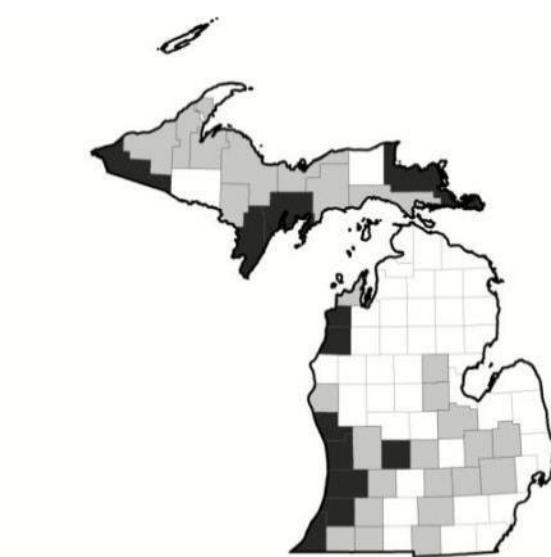
2005-2009



2010-2014



Dennis et al, 1998  
5 established  
22 reported



2007 (this study)  
12 established  
26 reported



Eisen et al, 2016  
24 established  
18 reported\*  
\*Includes 2 not reported by Eisen



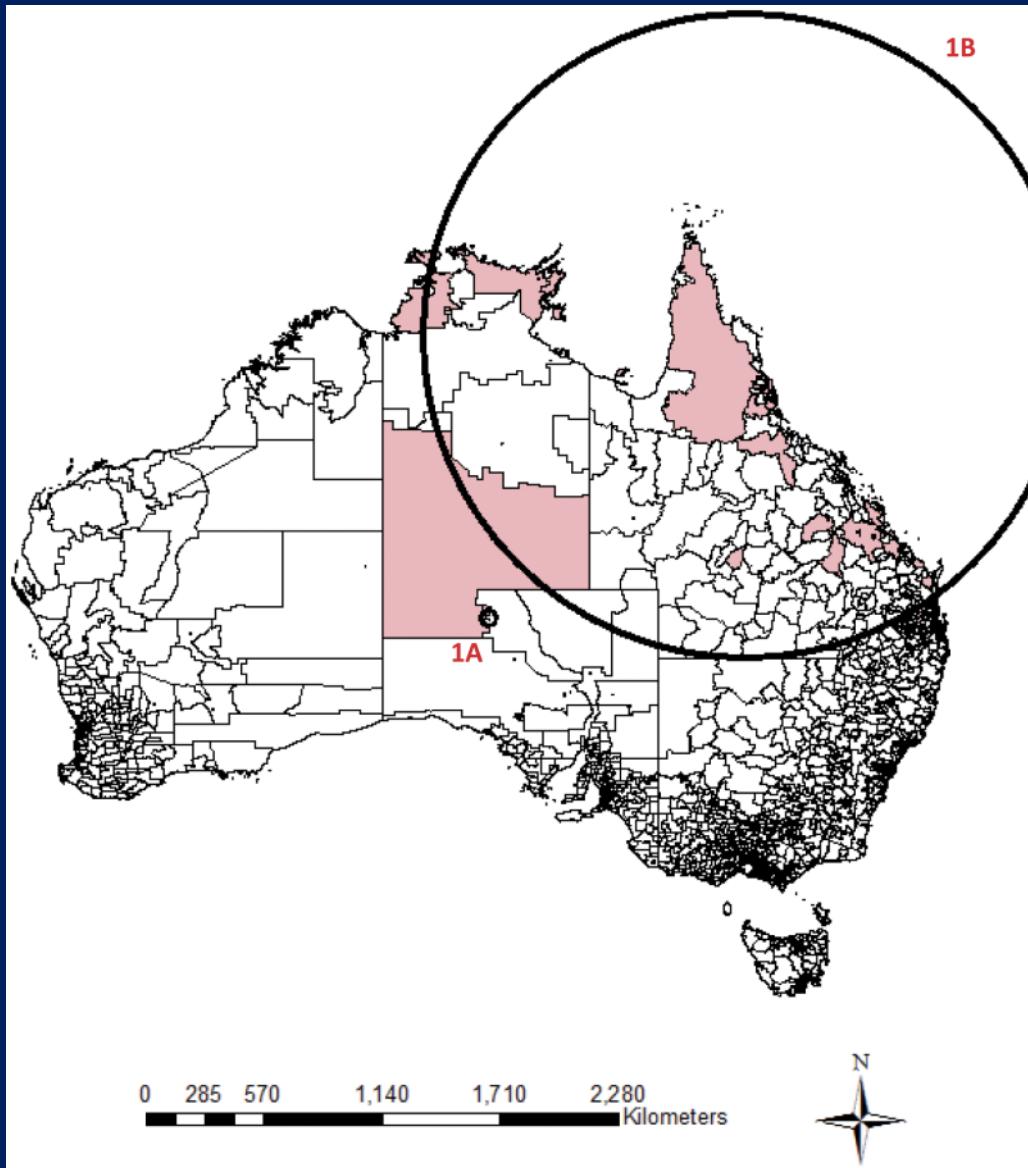
Established



Reported

# Suicide cluster analysis

Cheung, Y.T.D., Spittal, M.J., Williamson, M.K., Tung, S.J. and Pirkis, J., 2013. **Application of scan statistics to detect suicide clusters in Australia.** *PLoS One*, 8(1), p.e54168.



"the majority of spatial-temporal clusters of suicide were located in the inland northern areas, with socio-economic deprivation and higher proportions of indigenous people"

Cheung, Y.T.D., Spittal, M.J., Williamson, M.K., Tung, S.J. and Pirkis, J., 2013. **Application of scan statistics to detect suicide clusters in Australia.** *PLoS One*, 8(1), p.e54168.

# GIS applications in health – part 1

- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- Identifying spatial patterns, e.g. clusters of disease
- **Spatial statistics, spatial regression, spatial interpolation**

# Geographically weighted regression

Chi, S.H., Grigsby-Toussaint, D.S., Bradford, N. and Choi, J., 2013.  
**Can geographically weighted regression improve our  
contextual understanding of obesity in the US? Findings from  
the USDA Food Atlas.** *Applied Geography*, 44, pp.134-142.

(a) Percent obese



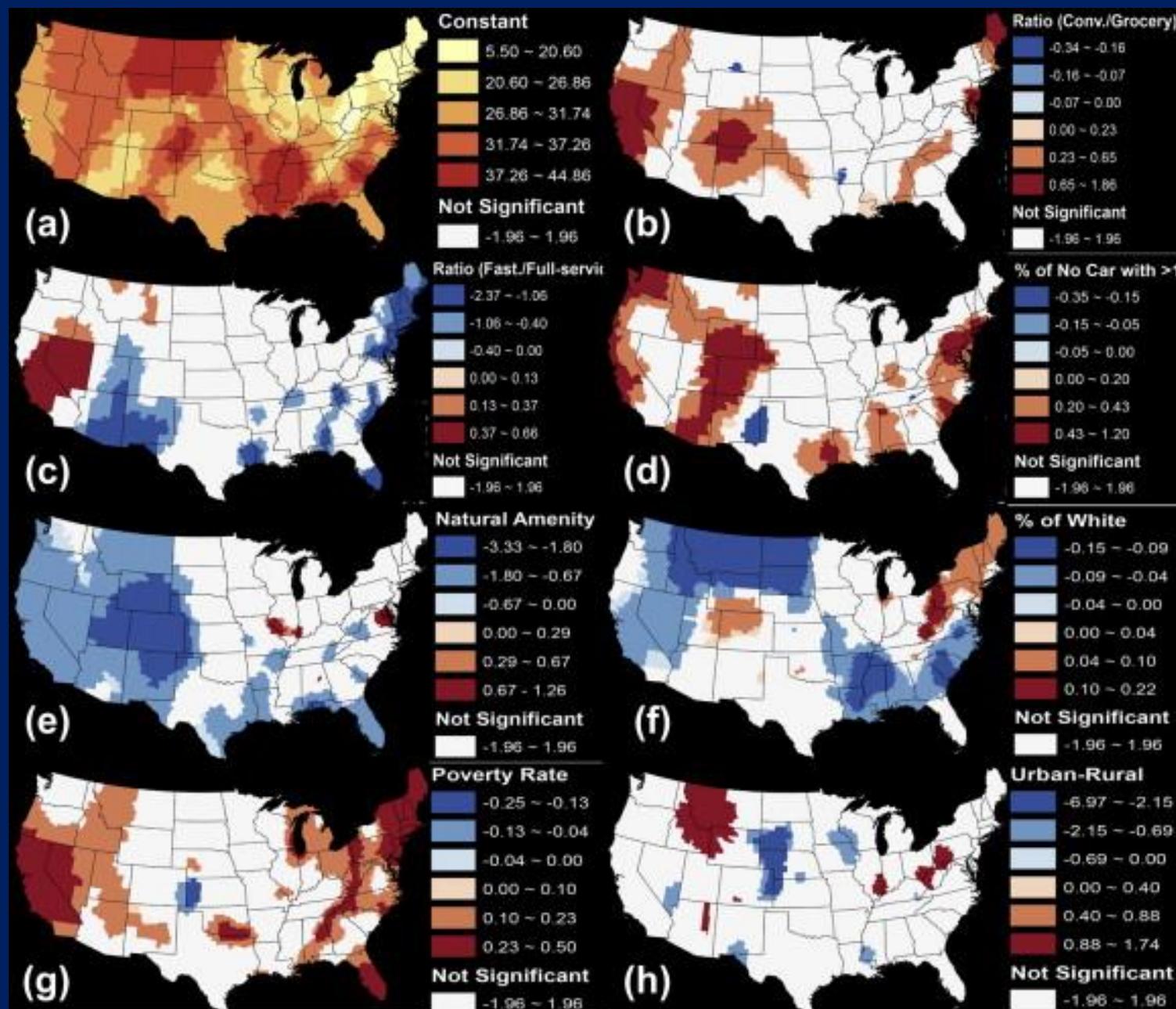
(b) Ratio: Grocery vs. Convenience Stores



(c) Ratio: Fast-food vs. Full-service Restaurants



Chi, S.H., Grigsby-Toussaint, D.S., Bradford, N. and Choi, J., 2013. Can geographically weighted regression improve our contextual understanding of obesity in the US? Findings from the USDA Food Atlas. *Applied Geography*, 44, pp.134-142.

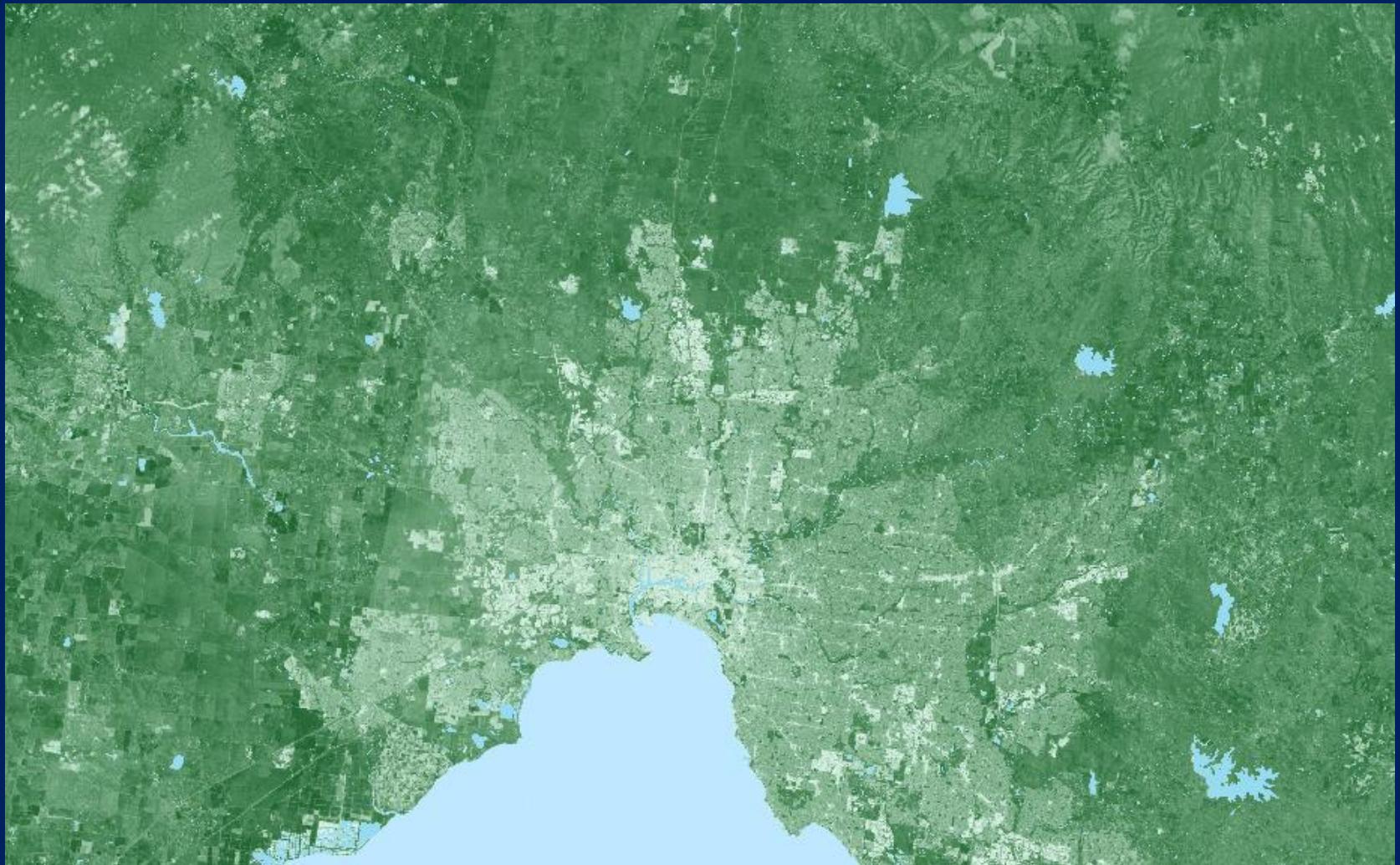


# GIS applications in health – part 1

- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- Identifying spatial patterns, e.g. clusters of disease
- Spatial statistics, spatial regression, spatial interpolation
- **Measure the environment (generate/test hypotheses )**

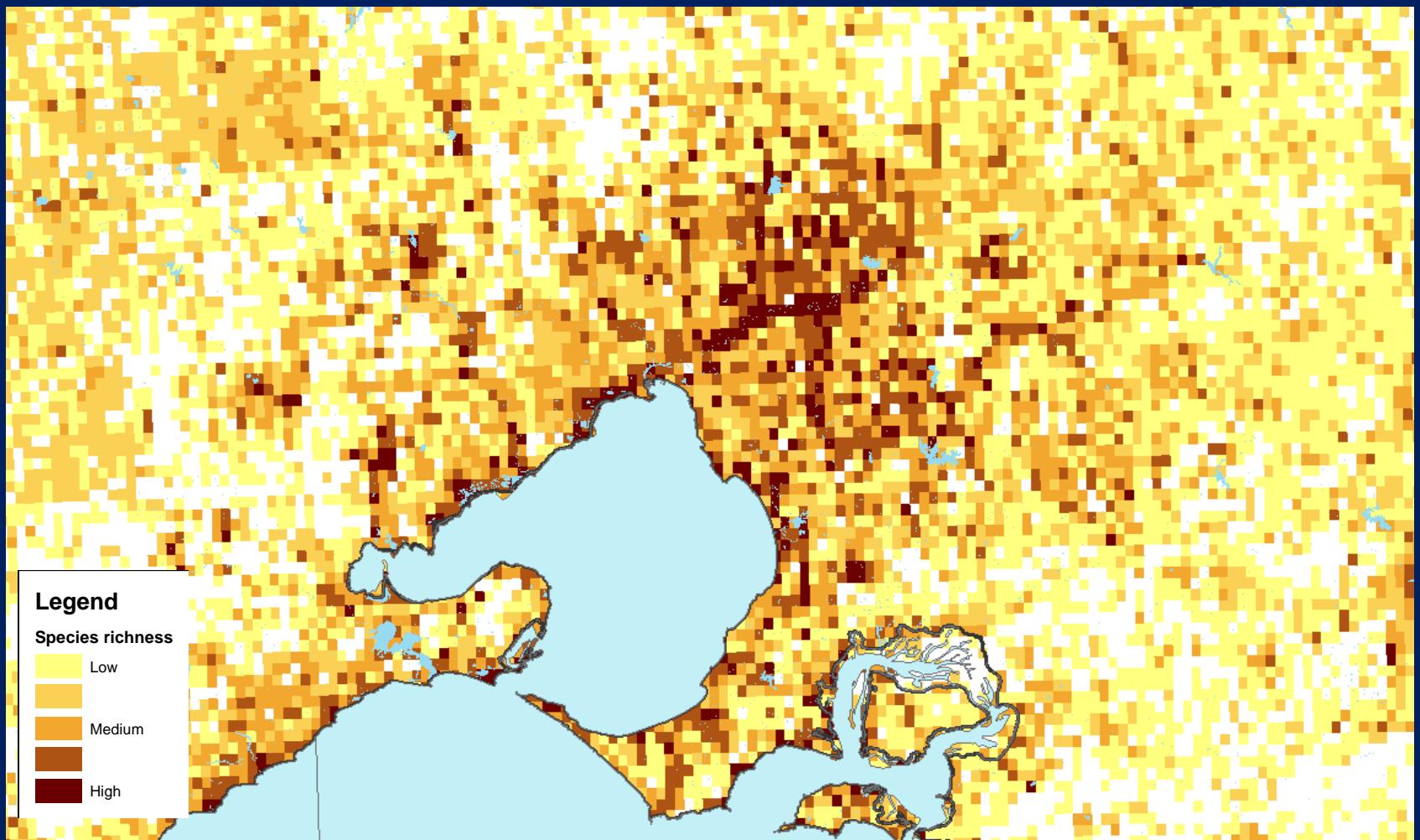
Do people who live in environments with more green, water (blue space), and biodiversity have more subjective wellbeing?

Greenness (NDVI) across Metropolitan Melbourne



Source: Mavoa, S. work in progress.

## Fauna species richness across Metropolitan Melbourne



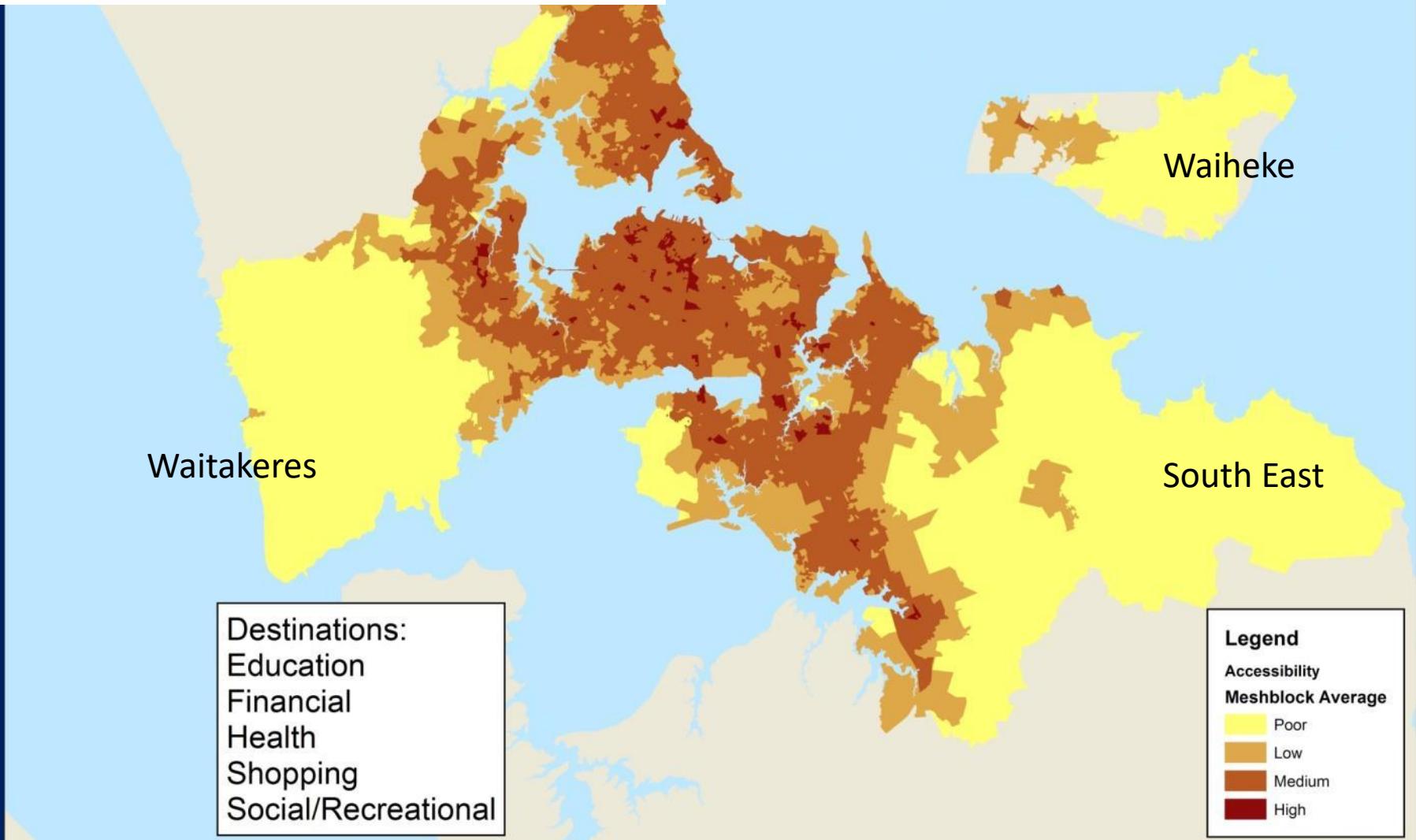
Source: Mavoa, S. work in progress.

# GIS applications in health – part 1

- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- Identifying spatial patterns, e.g. clusters of disease
- Spatial statistics, spatial regression, spatial interpolation
- **Measure the environment (generate/test hypotheses )**
- **Assess resource allocation/accessibility (e.g., health services, schools, water)**

- New public transit & walk access method
- Access **via** transit
- Used GTFS timetable data
- Calculated for every land parcel

## Transit and Walk Accessibility to Destinations in Auckland





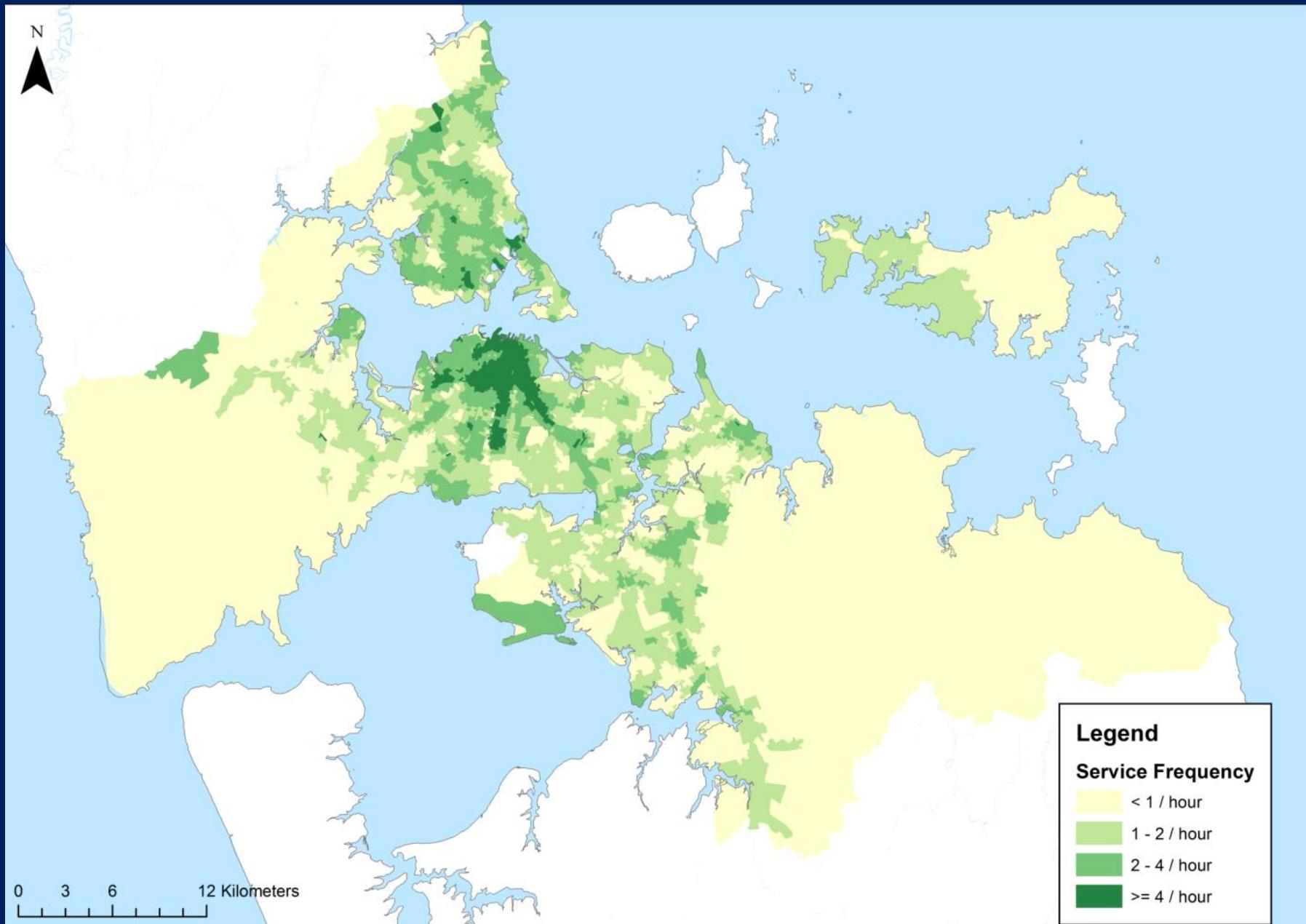
# Transit and Walk Accessibility to Social/Recreational Destinations in Auckland



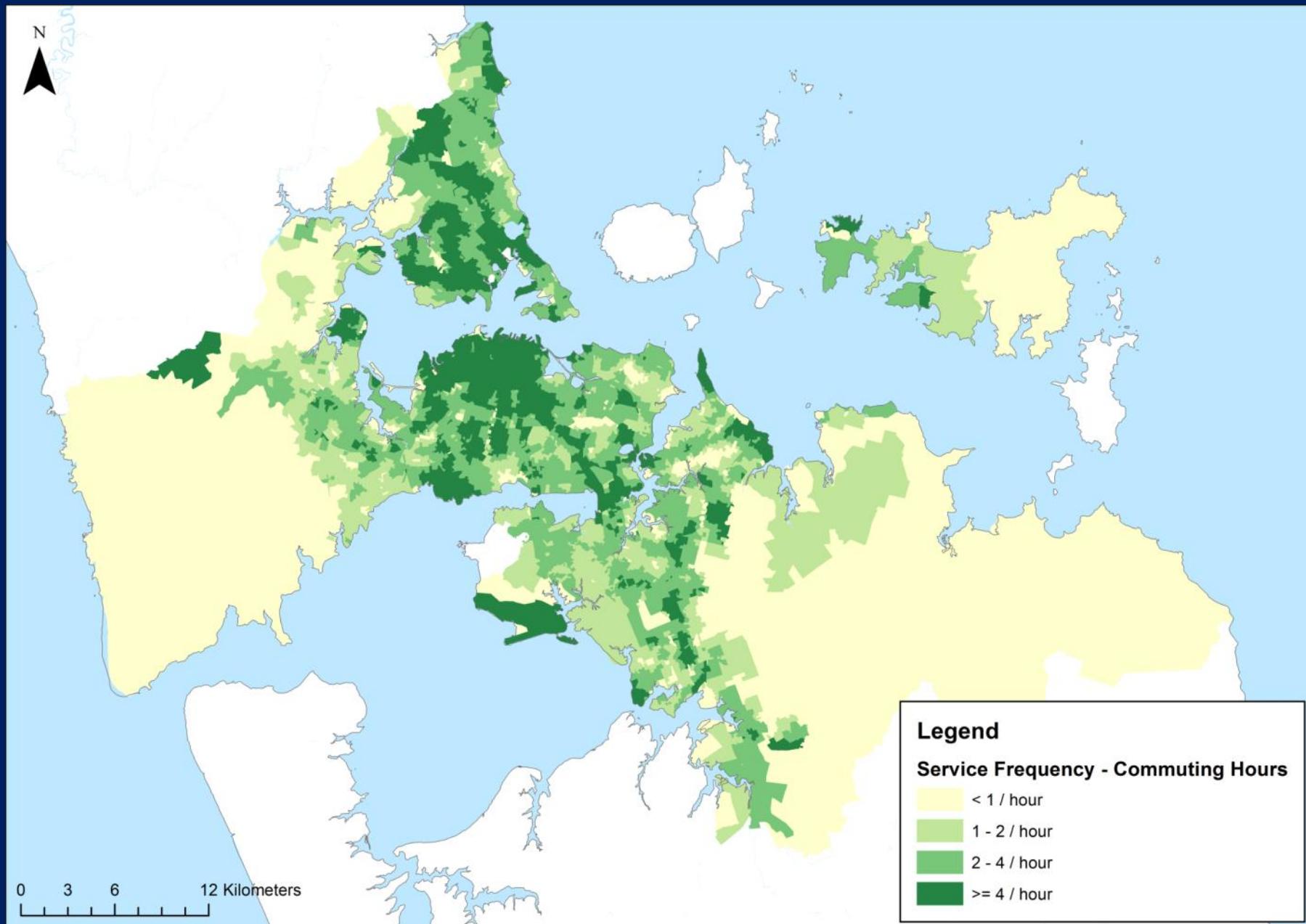
Social/Recreational Destinations:  
Cinemas  
Cafe/Restaurants  
Parks  
Beaches



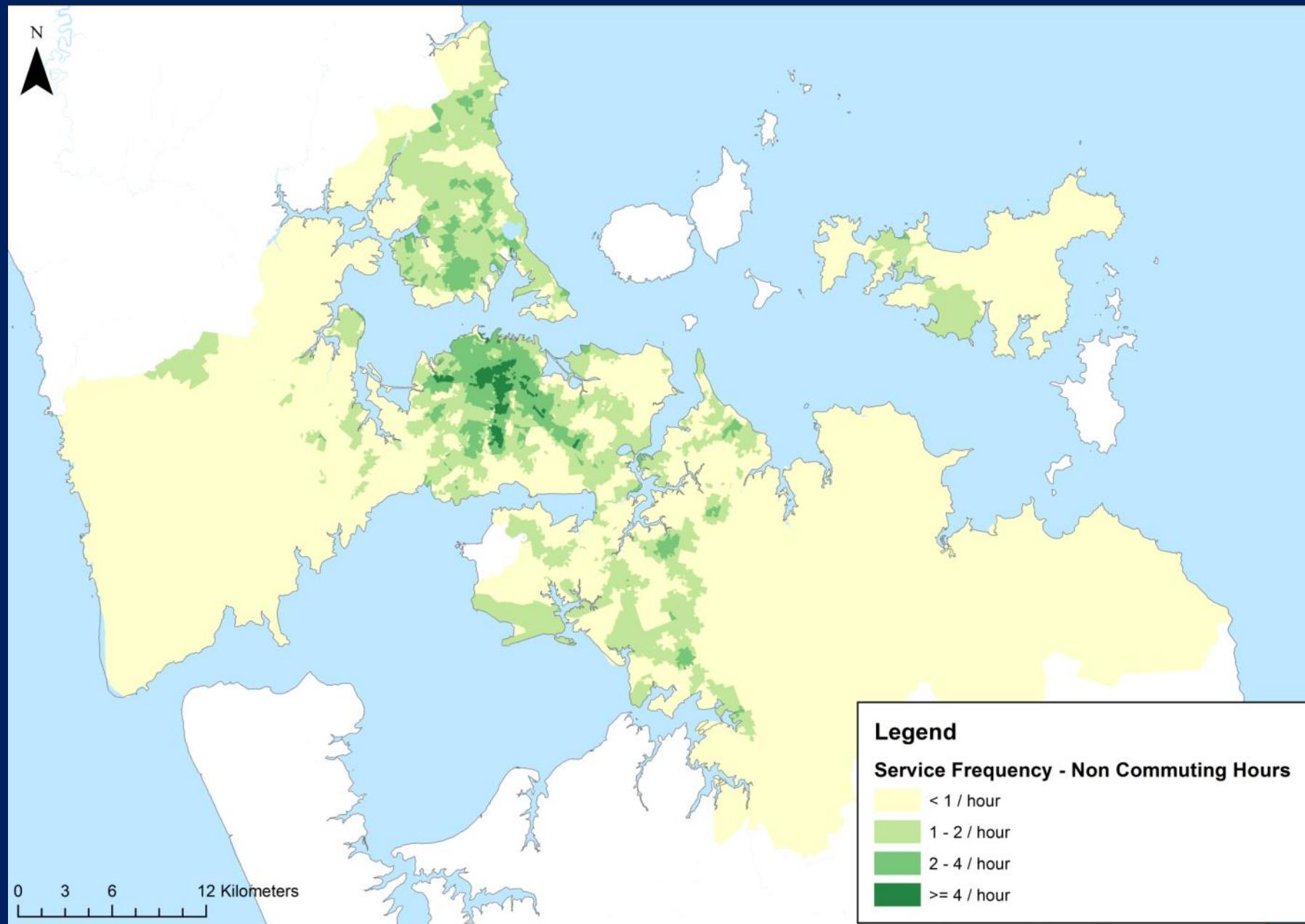
# Average public transit service frequency



# Average public transit service frequency during commuting hours

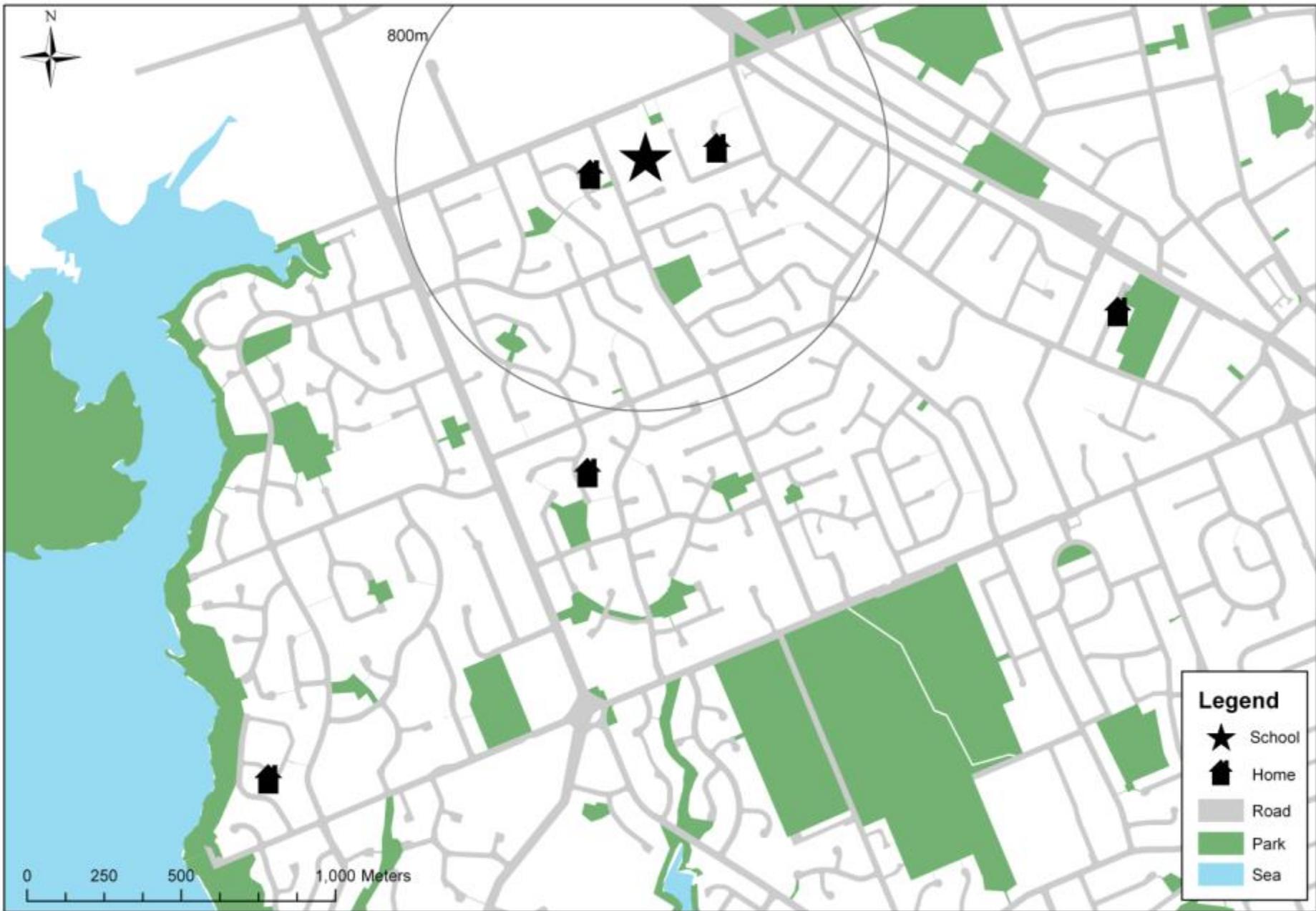


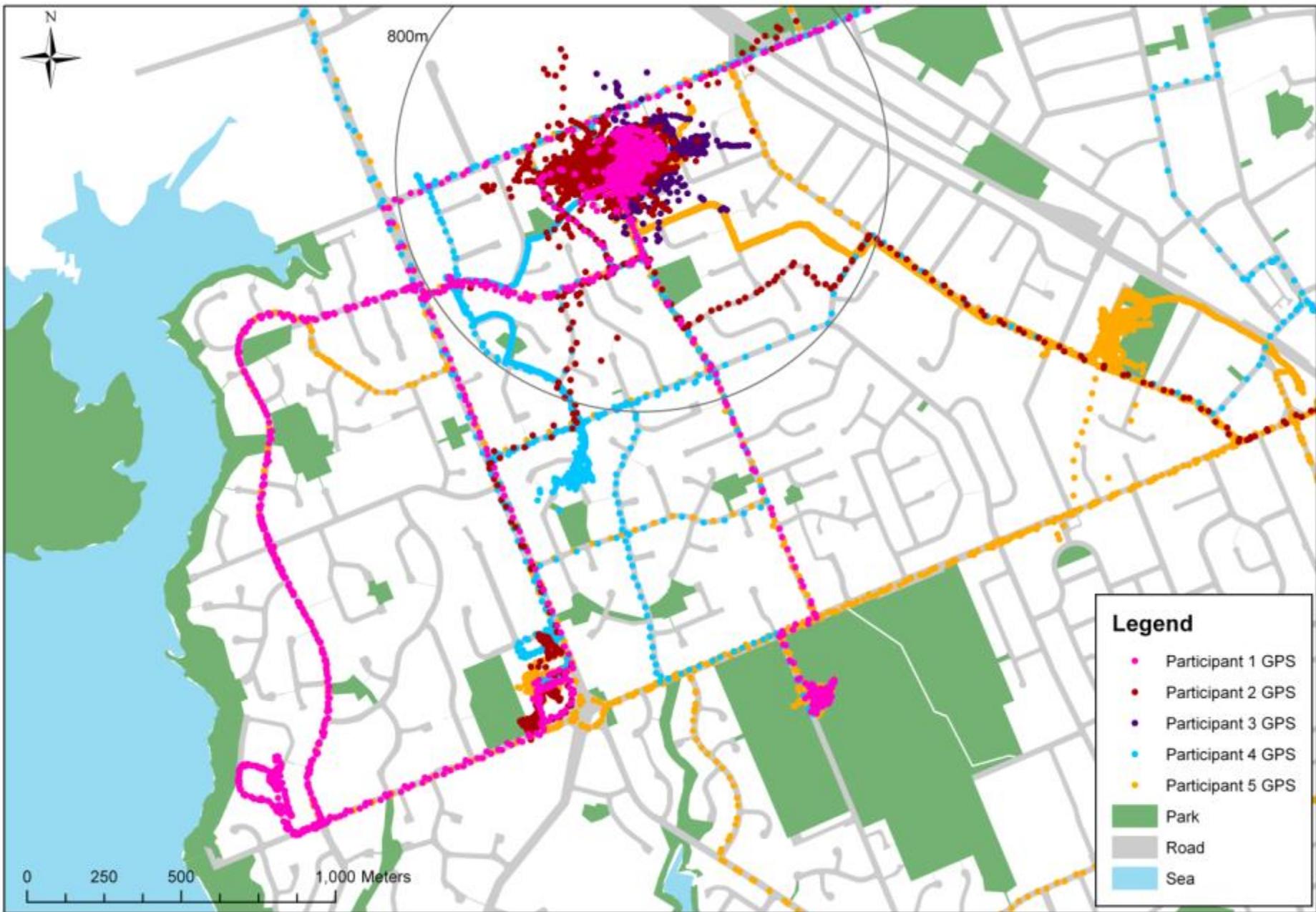
# Average public transit service frequency during non-commuting hours



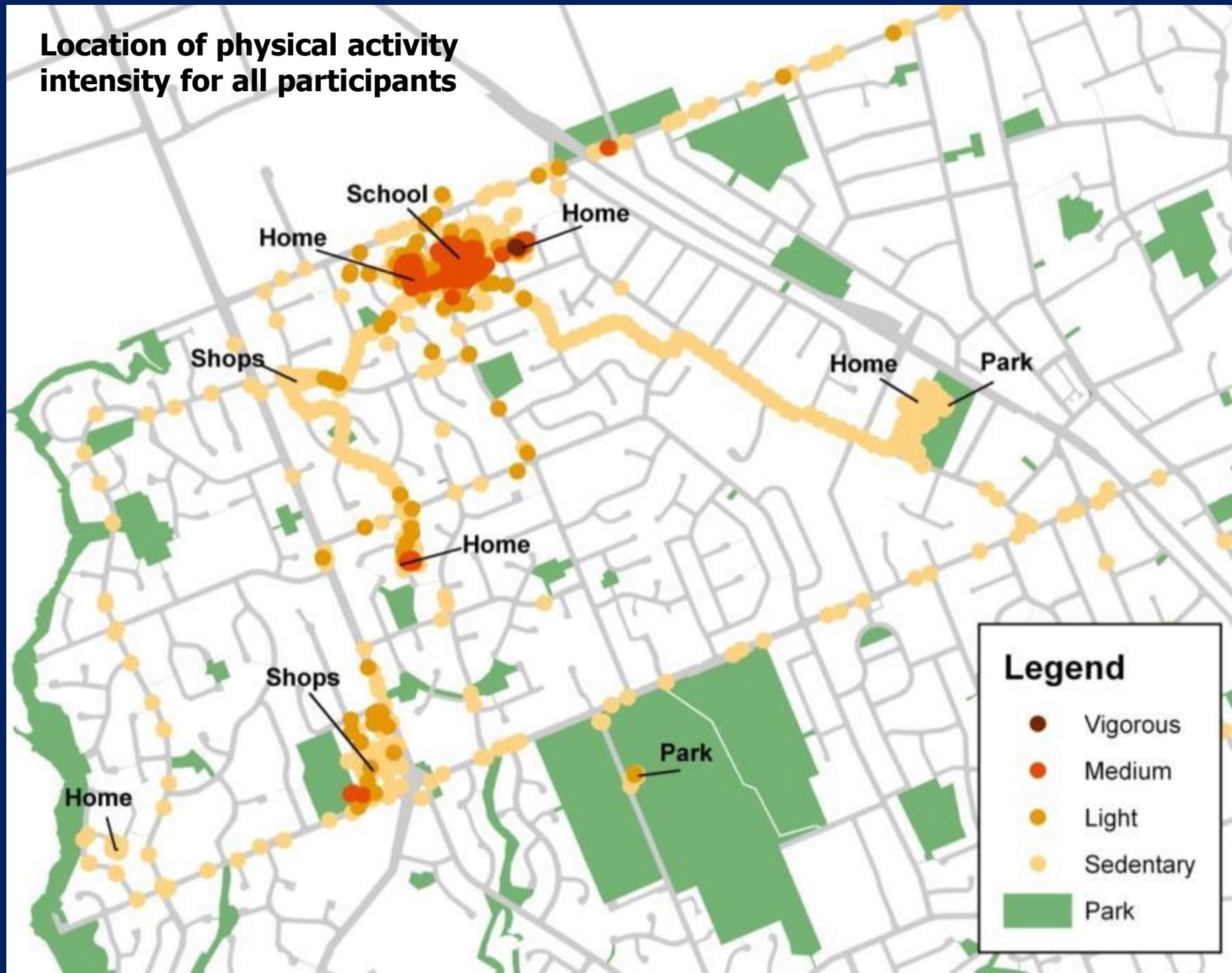
# GIS applications in health – part 1

- Geographic distribution and variation of disease (incidence/prevalence)
- Surveillance and monitoring of disease (infectious and chronic)
- Identifying at risk populations
- Identifying spatial patterns, e.g. clusters of disease
- Spatial statistics, spatial regression, spatial interpolation
- Measure the environment (generate/test hypotheses )
- Assess resource allocation/accessibility (e.g., health services, schools, water)
- **Estimate exposure to the environment (e.g, using GPS)**





## Location of physical activity intensity for all participants



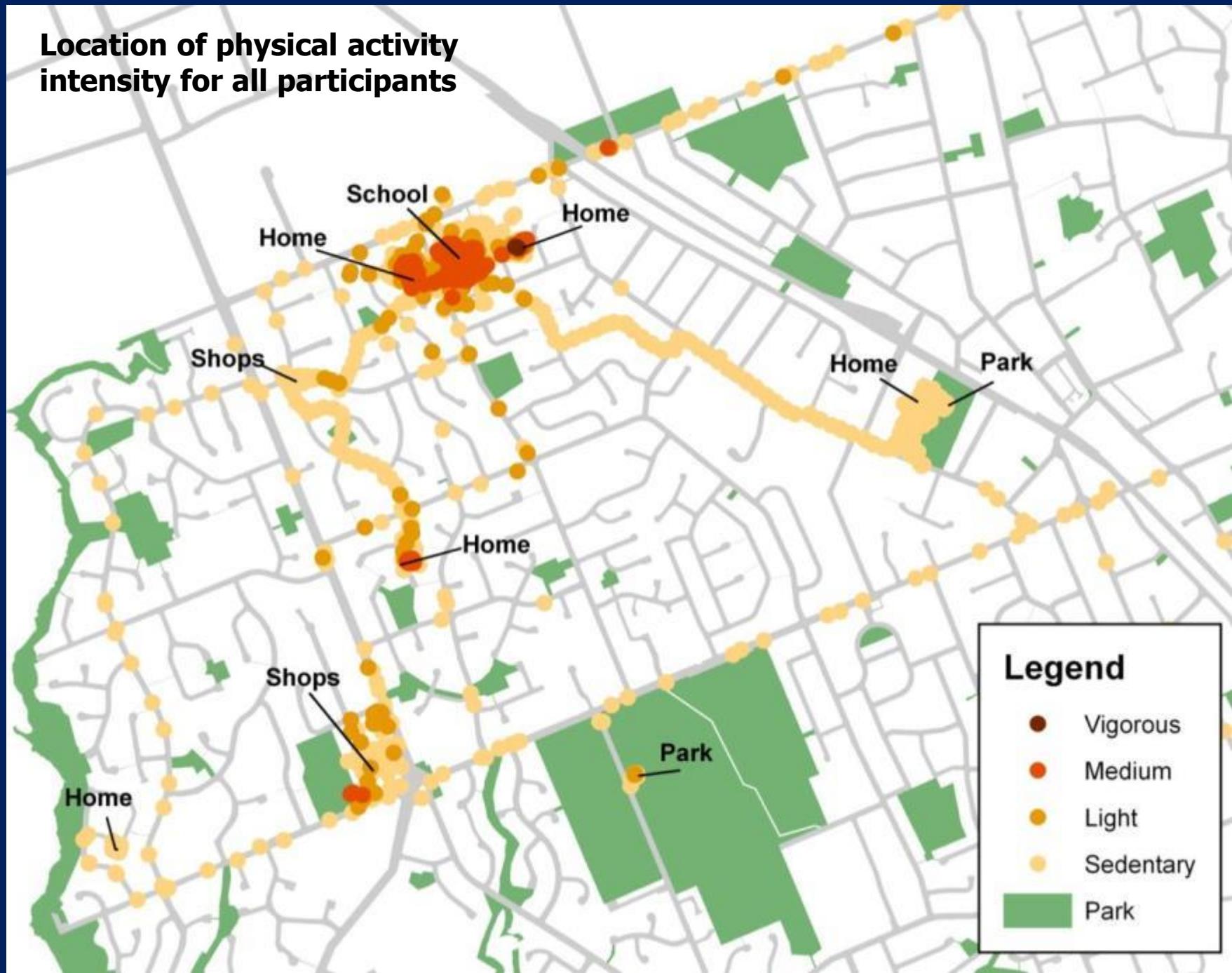
### Legend

- Vigorous
- Medium
- Light
- Sedentary
- Park

# GIS applications in health – part 2

- Mapping perceptions/qualitative data

## Location of physical activity intensity for all participants



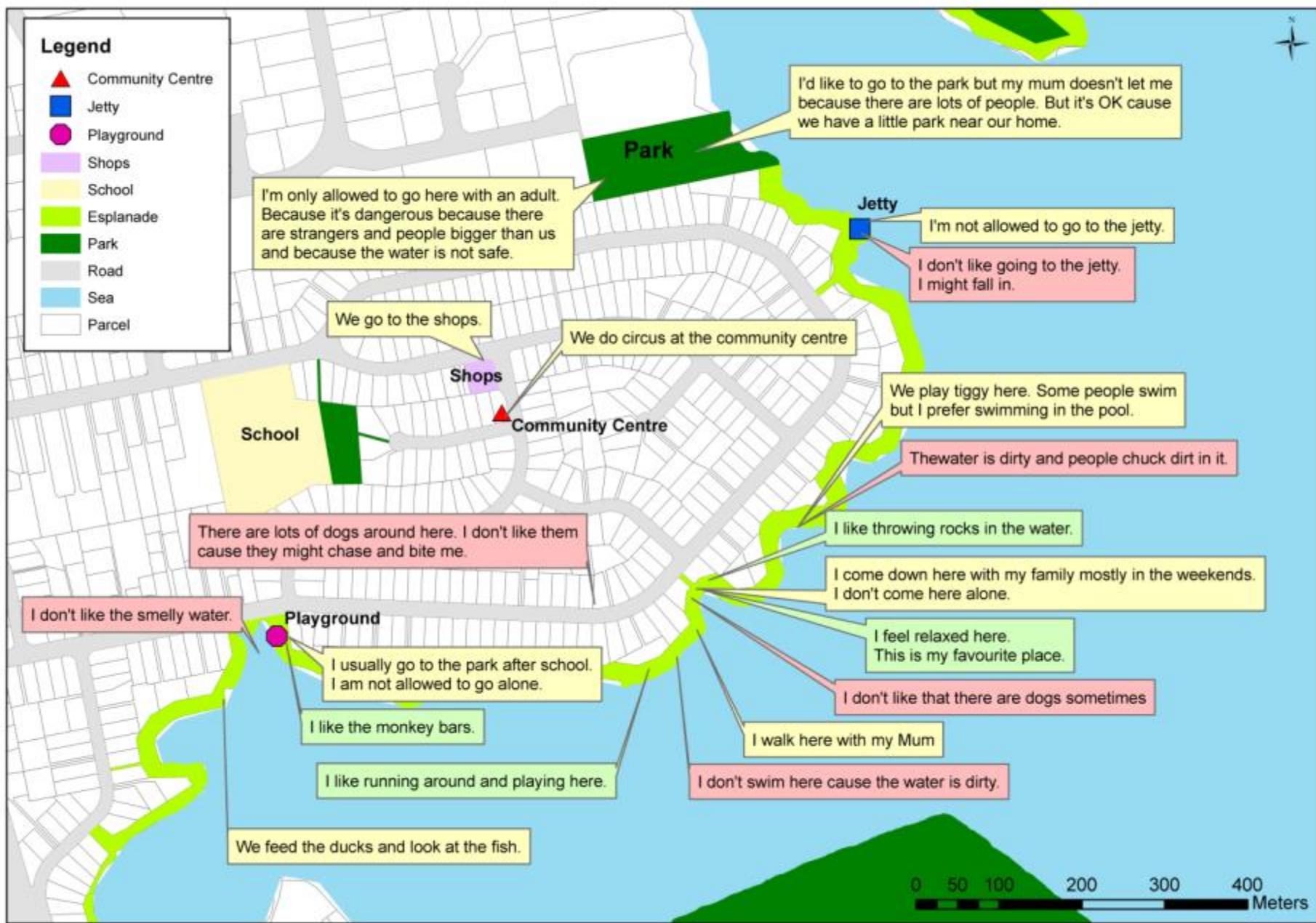
### Legend

- Vigorous
- Medium
- Light
- Sedentary
- Park



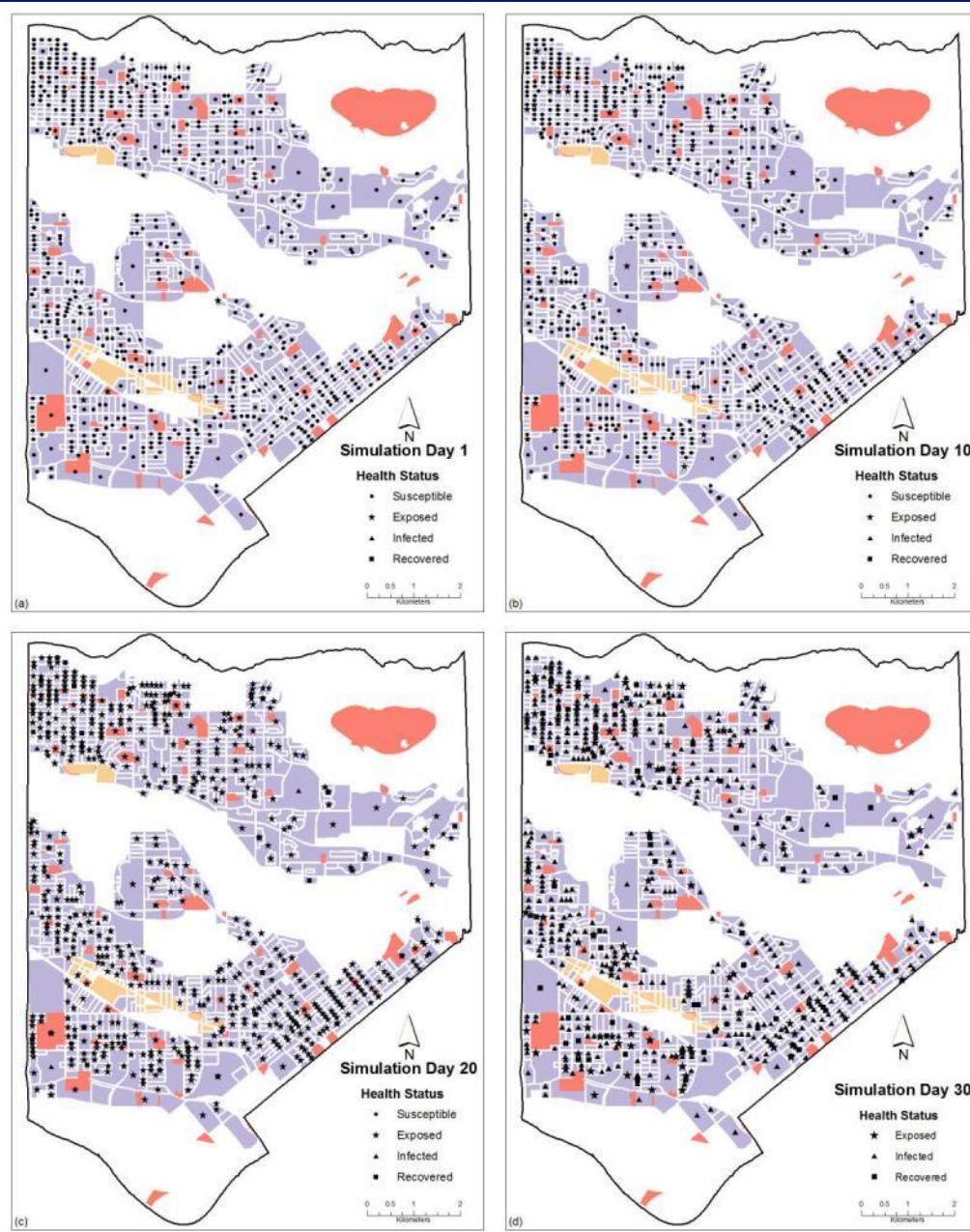
### Legend

- ▲ Community Centre
- Jetty
- Playground
- Shops
- School
- Esplanade
- Park
- Road
- Sea
- Parcel



# GIS applications in health – part 2

- Mapping perceptions/qualitative data
- **Simulating spatial behaviour (e.g, disease transmission, walking)**



Perez, L. and Dragicevic, S., 2009. An agent-based approach for modeling dynamics of contagious disease spread. *International journal of health geographics*, 8(1), p.50.

# GIS applications in health – part 2

- Simulating spatial behaviour (e.g, disease transmission, walking)
- Mapping perceptions/qualitative data
- **Measuring health-related policy implementation/progress**

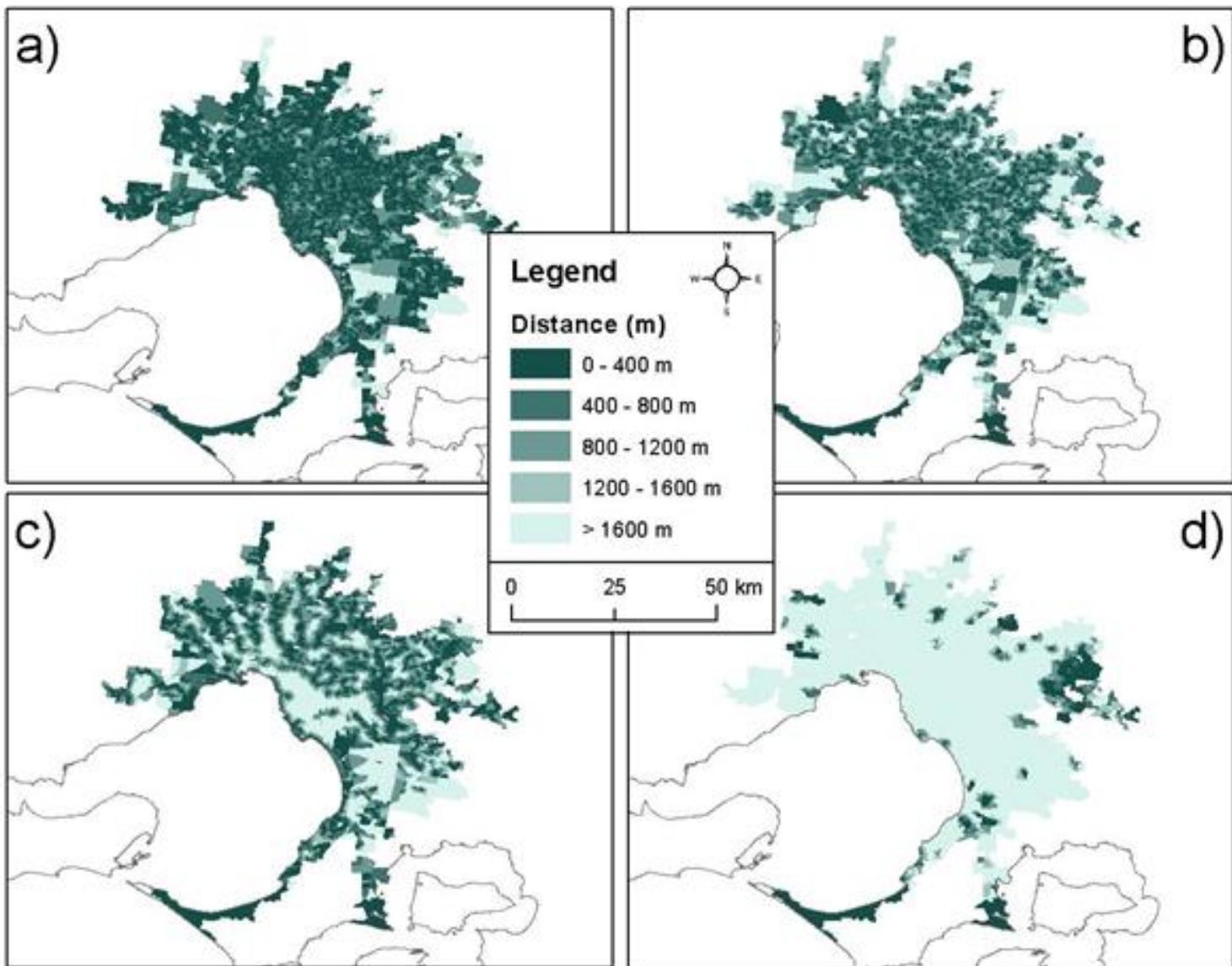
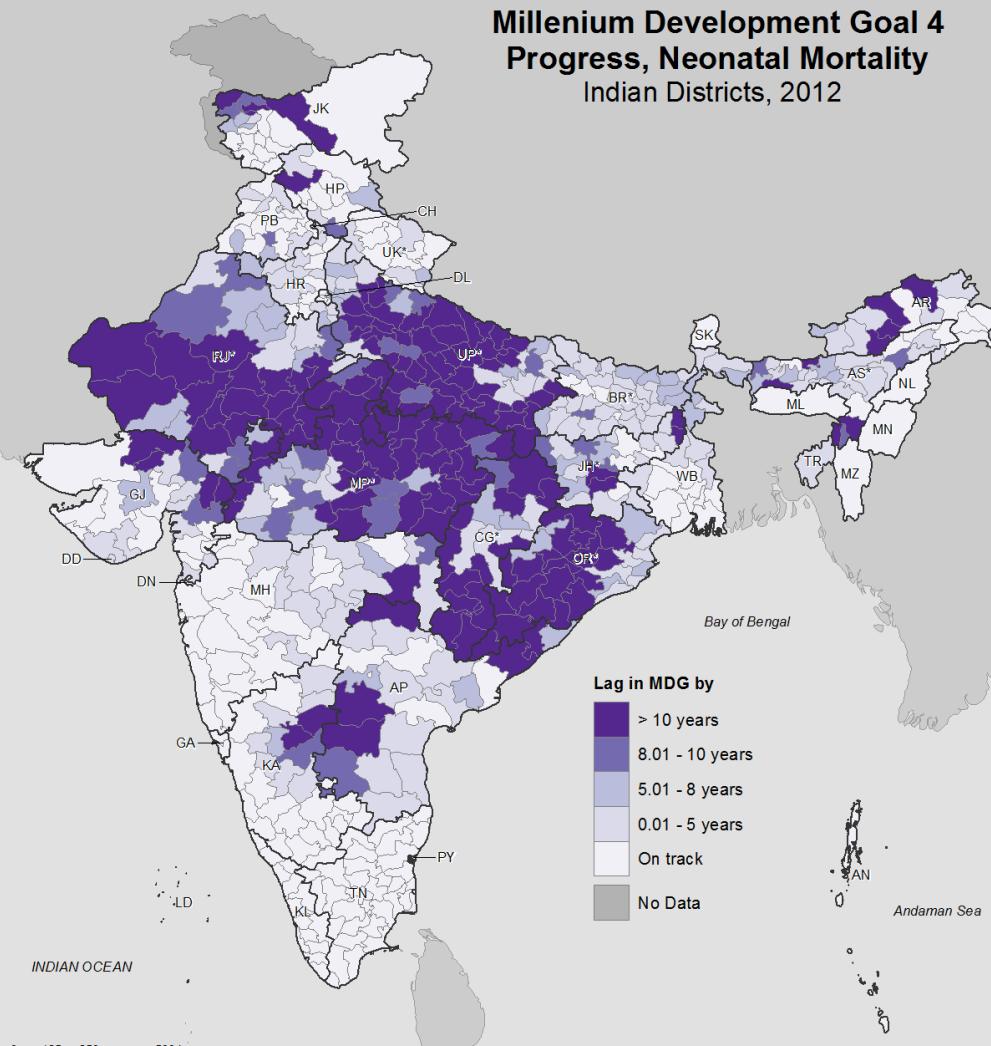


Figure 3. Distance (m) from SA1 population weighted centroid to closest a) park or garden, b) organised recreation area, c) natural or semi-natural area, and d) protected area in metropolitan Melbourne.

## Millenium Development Goal 4 Progress, Neonatal Mortality Indian Districts, 2012



St. Michael's  
Inspired Care. Inspiring Science.

MDG-4 progress per district in 2012, assessed by the number of years a district lagged behind the MDG of 38 per 1,000 live births by 2015 according to current trends. Richer states are the 20 states of Andhra Pradesh (AP), Arunachal Pradesh (AR), Delhi (DL), Goa (GA), Gujarat (GJ), Haryana (HR), Himachal Pradesh (HP), Jammu and Kashmir (JK), Karnataka (KA), Kerala (KL), Maharashtra (MH), Manipur (MN), Meghalaya (ML), Mizoram (MZ), Nagaland (NG), Punjab (PB), Sikkim (SK), Tamil Nadu (TN), Tripura (TR), West Bengal (WB), and six small union territories. Poorer states are the nine states of Assam (AS), Bihar (BR), Chhattisgarh (CG), Jharkhand (JH), Madhya Pradesh (MP), Orissa (OR), Rajasthan (RJ), Uttarakhand (UK), and Uttar Pradesh (UP).

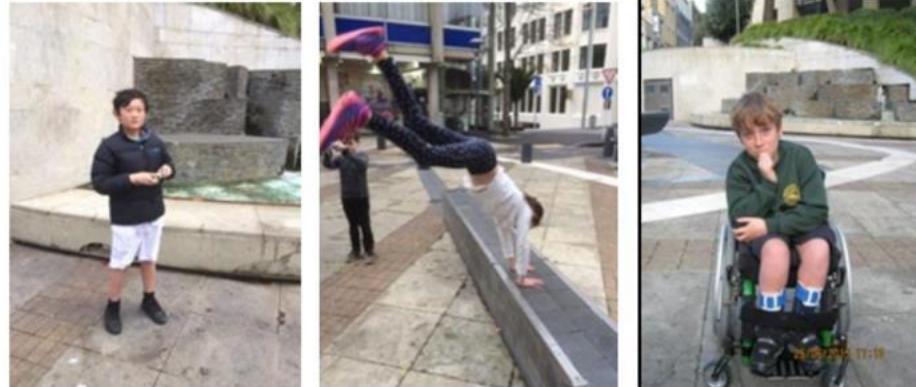
# **GIS applications in health – part 2**

- Simulating spatial behaviour (e.g, disease transmission, walking)
- Mapping perceptions/qualitative data
- Measuring health-related policy implementation
- **Targeting interventions**
- **Assessing the impact of interventions**
- **Assisting in research design and data collection (E.g., identifying study area, design of sample frames )**

# GIS applications in health – part 2

- Simulating spatial behaviour (e.g, disease transmission, walking)
- Mapping perceptions/qualitative data
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- Targeting interventions
- Assessing the impact of interventions
- Assisting in research design and data collection (E.g., identifying study area, design of sample frames )
- **Communication/knowledge translation/community engagement**

# Kids in the City – child friendly urban design audit



"The water feature was the first thing I saw and the first thing I went to. It was an interactive feature – it wasn't just taking up a chunk of space. It was fun climbing to the top."



"No cars should be allowed to drive in."

Aerial perspective view of Pioneer Women's & Ellen Melville Hall and Freyberg Square showing the removal of the Freyberg Place roadway to create an integrated public space and community hall. Concrete seating terraces and steps extend up the bank towards the Metropolis, interspersed with native tree and shrub plantings and an interactive water feature, creating a destination public space in the city.

CONCEPT DESIGN

"A group of kids could have like a scavenger hunt around the statue and it could be like an adventure."

"I like the water, how it comes down, the noise it makes."

"A drinking fountain."

"I'd come on a warm day with one of my friends, scooter around, go up into the tree, play on a playground – and I'd probably feed the birds."

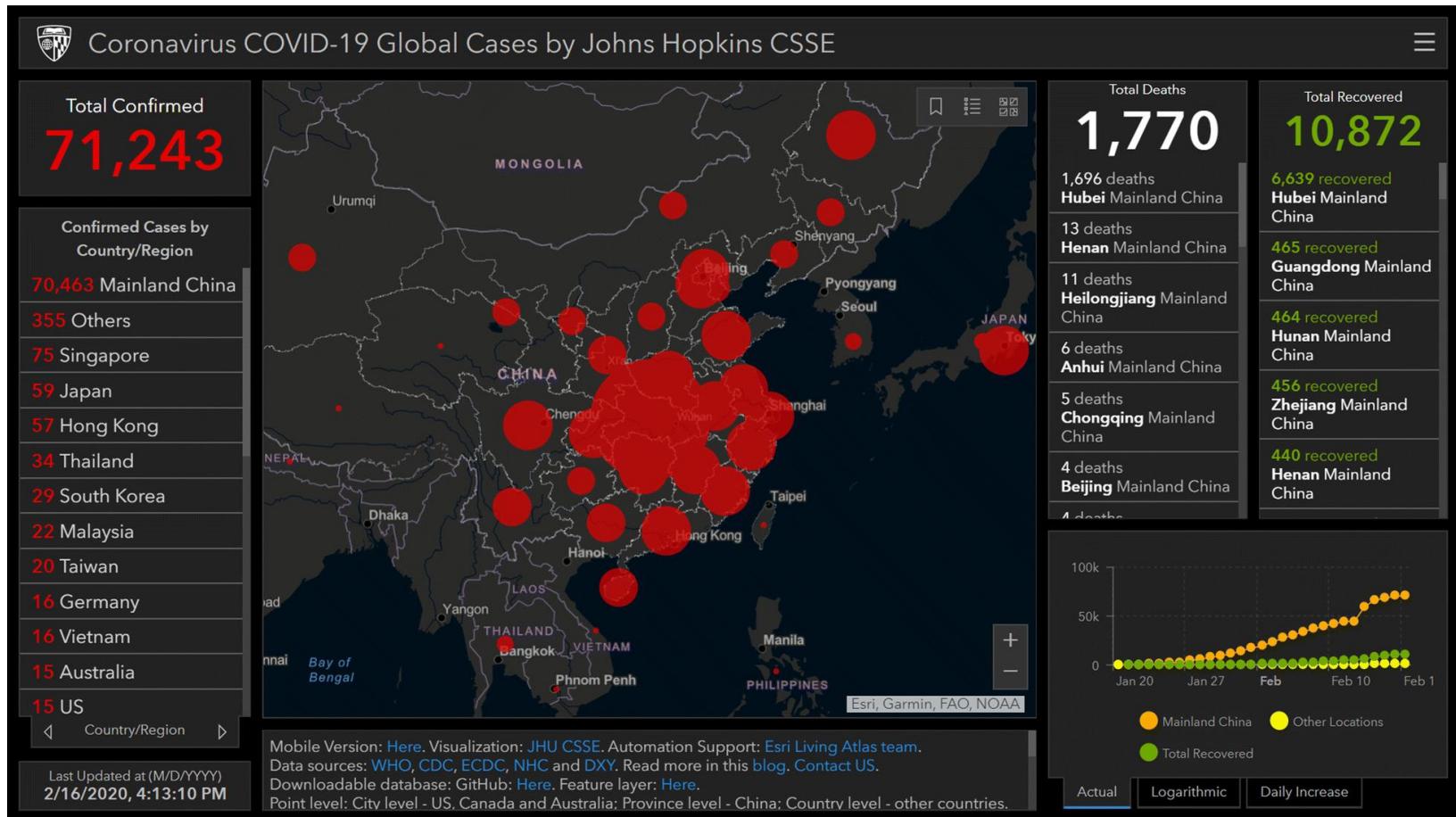
PIONEER WOMEN'S AND ELLEN MELVILLE HALL AND FREYBERG SQUARE UPGRADE

AERIAL PERSPECTIVE

# **How 21st century GIS technologies are supporting the global fight against outbreaks and epidemics**

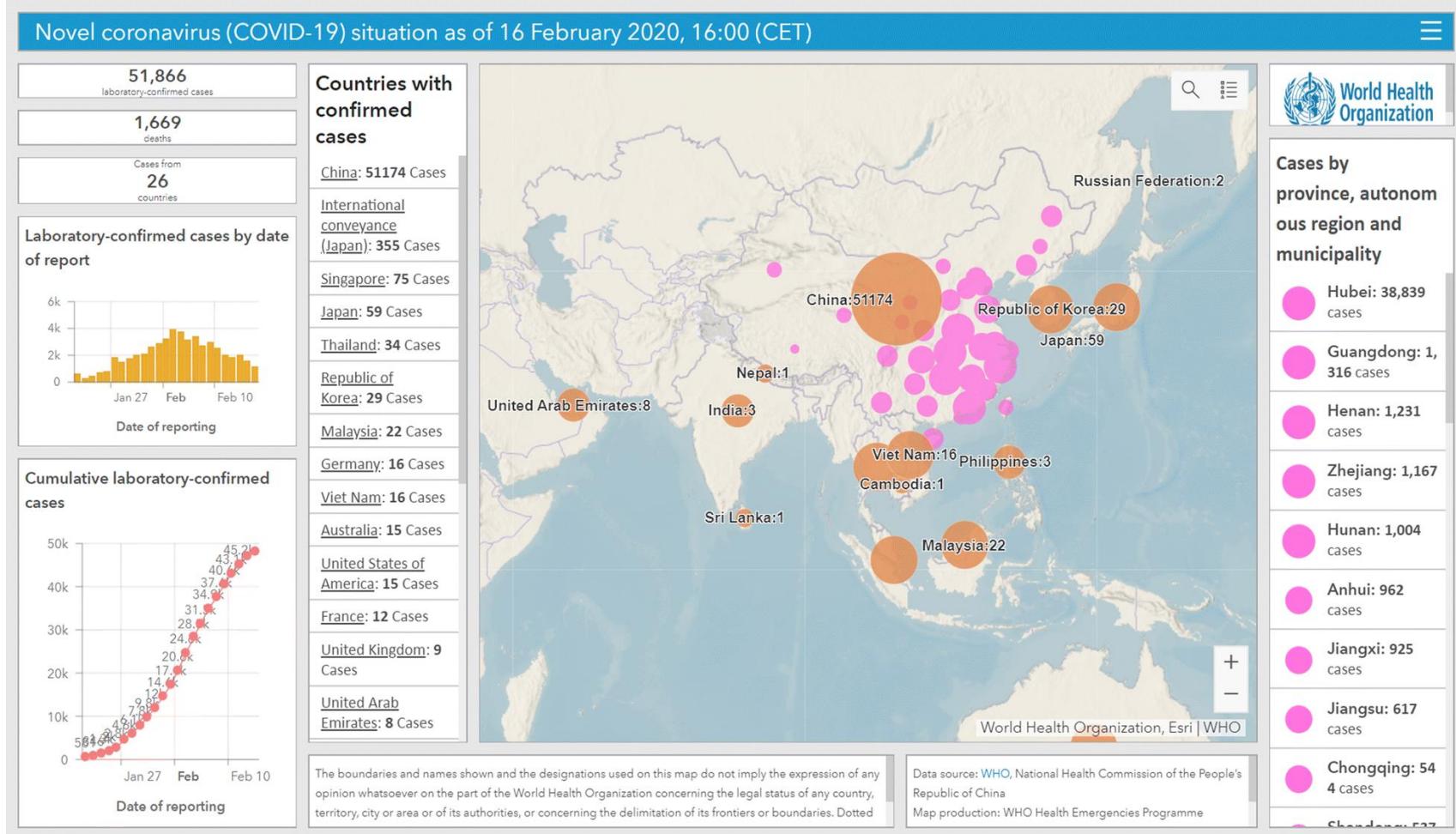
Boulos, M.N.K. and Geraghty, E.M., 2020. Geographical tracking and mapping of coronavirus disease COVID-19/severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) epidemic and associated events around the world: how 21st century GIS technologies are supporting the global fight against outbreaks and epidemics.

# Map-based dashboards



integrating disparate information, visualisation/communication

# Map-based dashboards



integrating disparate information, visualisation/communication

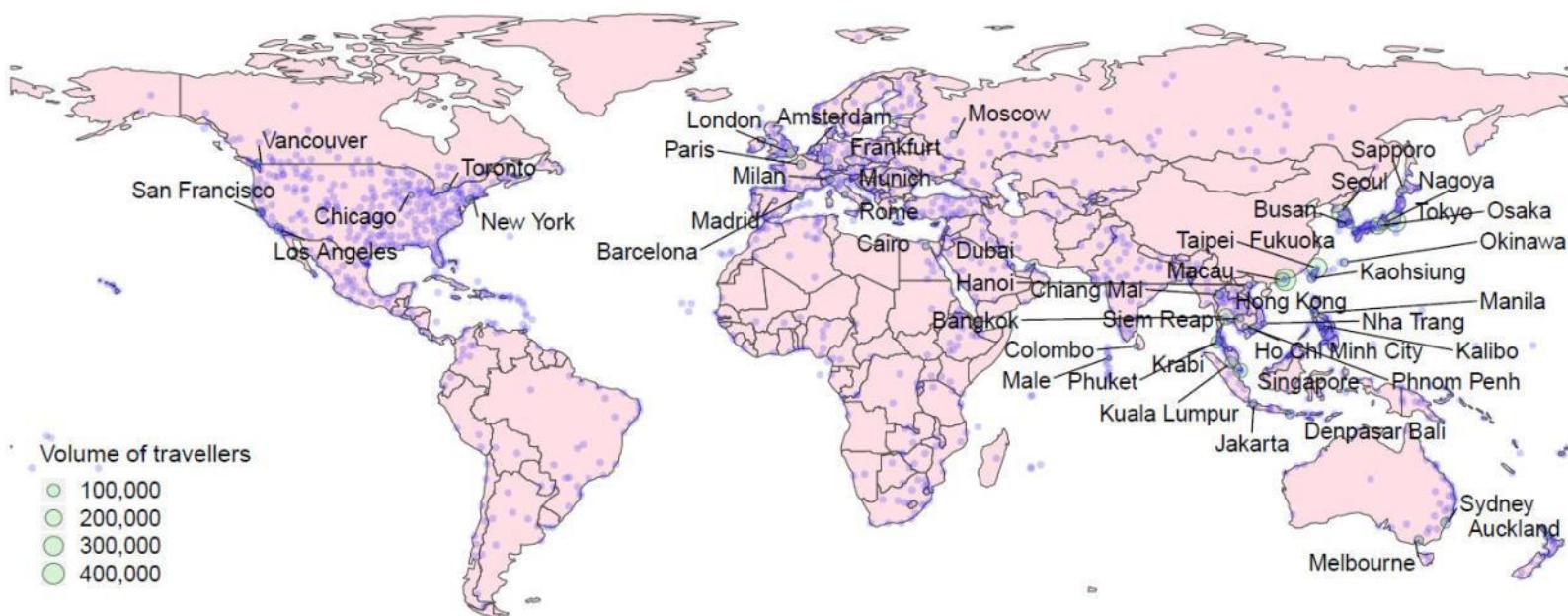
# Geospatial apps ‘close contact detector’



“Functions include: close contact inquiry, including train journey number and plane flight number checking for diagnosed cases, and location information about the activity spaces of nearby confirmed cases (no individual names are ever displayed in returned results)”

integrating disparate information, data collection/creation,  
spatial analysis, visualisation/communication

# WorldPop and EpiRisk predictive global risk analytics and maps for SARS-CoV-2 based on population movements out of Wuhan and travel destinations

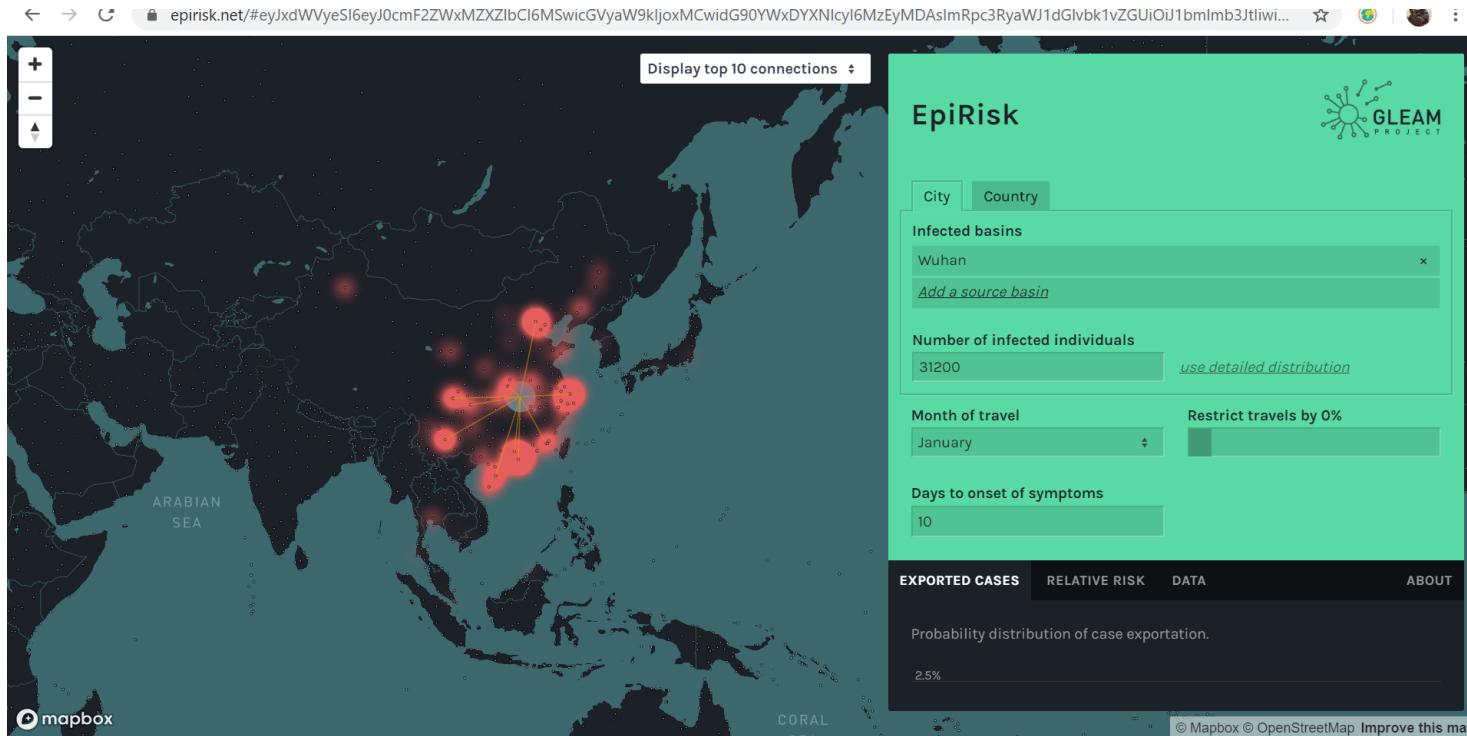


Geographic distribution of cities travellers from high risk cities in China.

Source: Lai et al. 2020: <https://www.worldpop.org/events/china>

Lai, S., Bogoch, I.I., Ruktanonchai, N., Watts, A.G., Li, Y., Yu, J., Lv, X., Yang, W., Hongjie, Y., Khan, K. and Li, Z., 2020. Assessing spread risk of Wuhan novel coronavirus within and beyond China, January-April 2020: a travel network-based modelling study.

# WorldPop and EpiRisk predictive global risk analytics and maps for SARS-CoV-2 based on population movements out of Wuhan and travel destinations



epirisk.net  
Interactive risk map

# Mapping the worldwide spread of misinformation about coronavirus

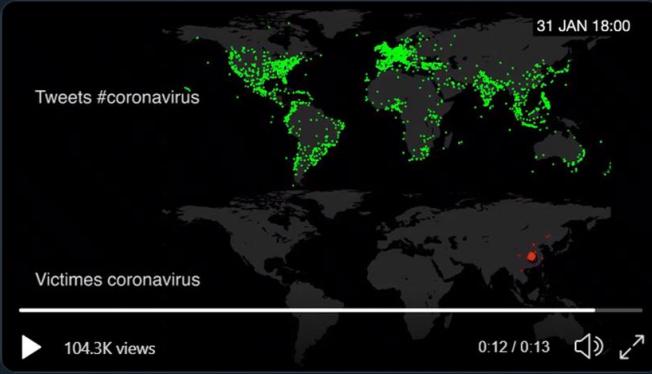
 Search Twitter

 **Mehdi Moussaid**   
@Mehdi\_Moussaid

Propagation du hashtag **#coronavirus** sur Twitter (en vert) et des victimes du coronavirus (en rouge), entre le 24 et le 31 janvier.

"La rumeur et la panique se propagent plus vite que le virus", disais Margaret Chan (directrice de l'OMS)

**#Fouloscopie**



31 JAN 18:00

Tweets #coronavirus

Victimes coronavirus

▶ 104.3K views 0:12 / 0:13

3:46 pm · 3 Feb 2020 · Twitter Web App

1.4K Retweets 3.2K Likes

## **Other ways GIS technologies can help in combat infectious disease outbreaks and epidemics**

### **Outbreak source**

Spatial analyses integrating phyloepidemiological methods can be used to identify the likely sources of new outbreaks.

### **Public events**

Dashboards and Web maps that bring together location and time-sensitive events to help assess and reduce exposure.

### **Site selection**

Site selection is a common GIS application...relevant to epidemics.

### **Supply chain**

GIS can assist in supply chain planning and management.

### **Resource locators**

Residents of affected areas can use apps and maps to locate aid and resources (including navigation)

### **Drones**

Transporting crucial medical supplies and patient lab sample. Reduce human contact.  
Used for broad disinfecting. Integrated drone and GIS technologies can help target and speed efforts in places they are needed most.

# Key issues applying GIS in public health

- Data
- 2. Delineation of exposures
- 3. Self selection & scale
- 4. Other

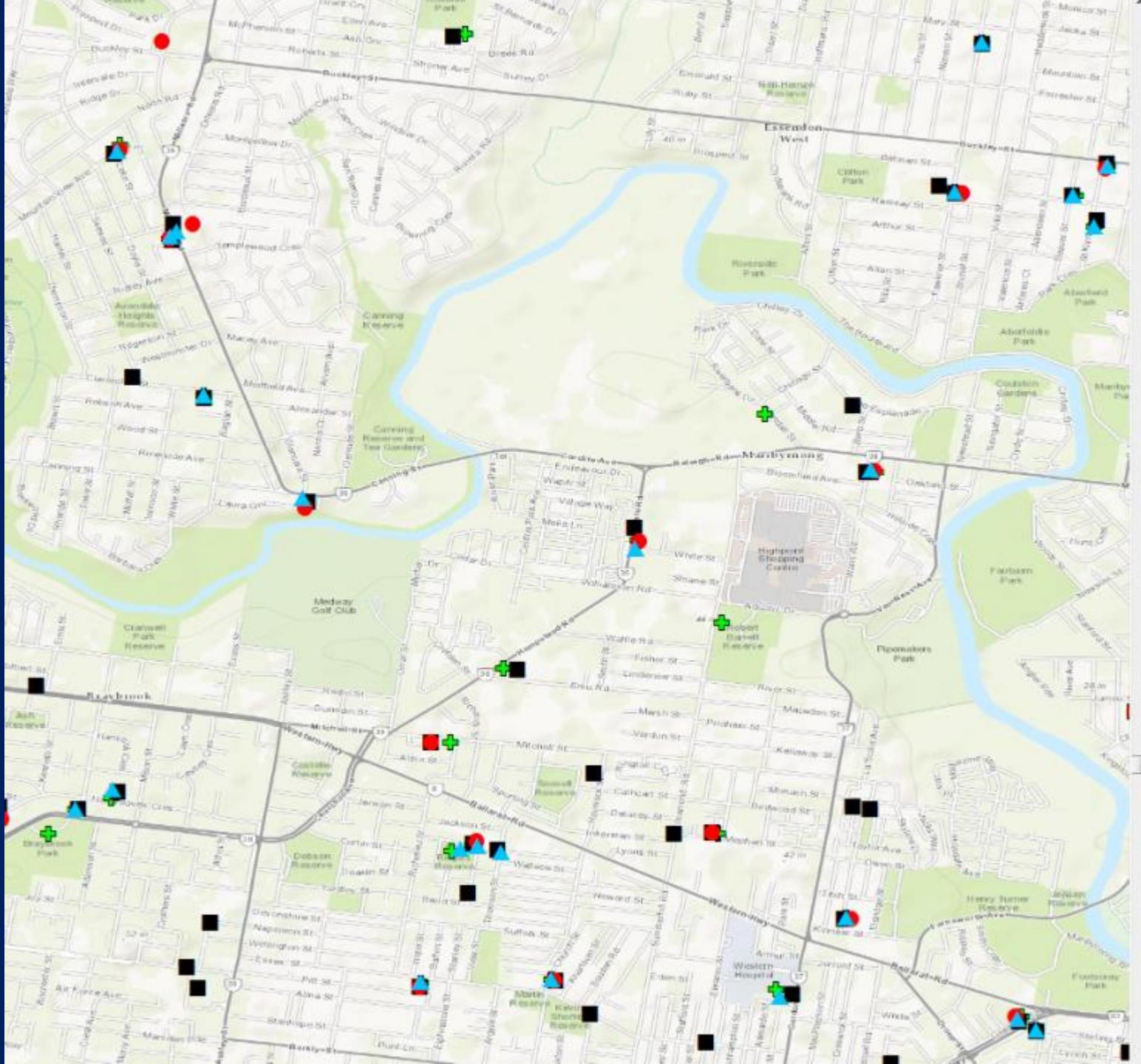
# Data challenges

- Sourcing
- Licenses
- Confidentiality/Privacy/Ethics
- Cost
- Error/accuracy
- Resolution/scale/extent
- Temporal matching
- Large datasets -> computationally intensive analysis

# Inconsistencies across data sources

Example: Childcare facilities in Victoria

Data source	Count
A	3966
B	4183
C	2799
S	2094



Papua New Guinea

# Supermarkets in the ocean



Duncan S, Stewart TI, Oliver M,  
Mavoa S, MacRae D, Badland HM,  
Duncan MJ. Portable global  
positioning system receivers:  
static validity and environmental  
conditions. American journal of  
preventive medicine. 2013 Feb  
28;44(2):e19-29.

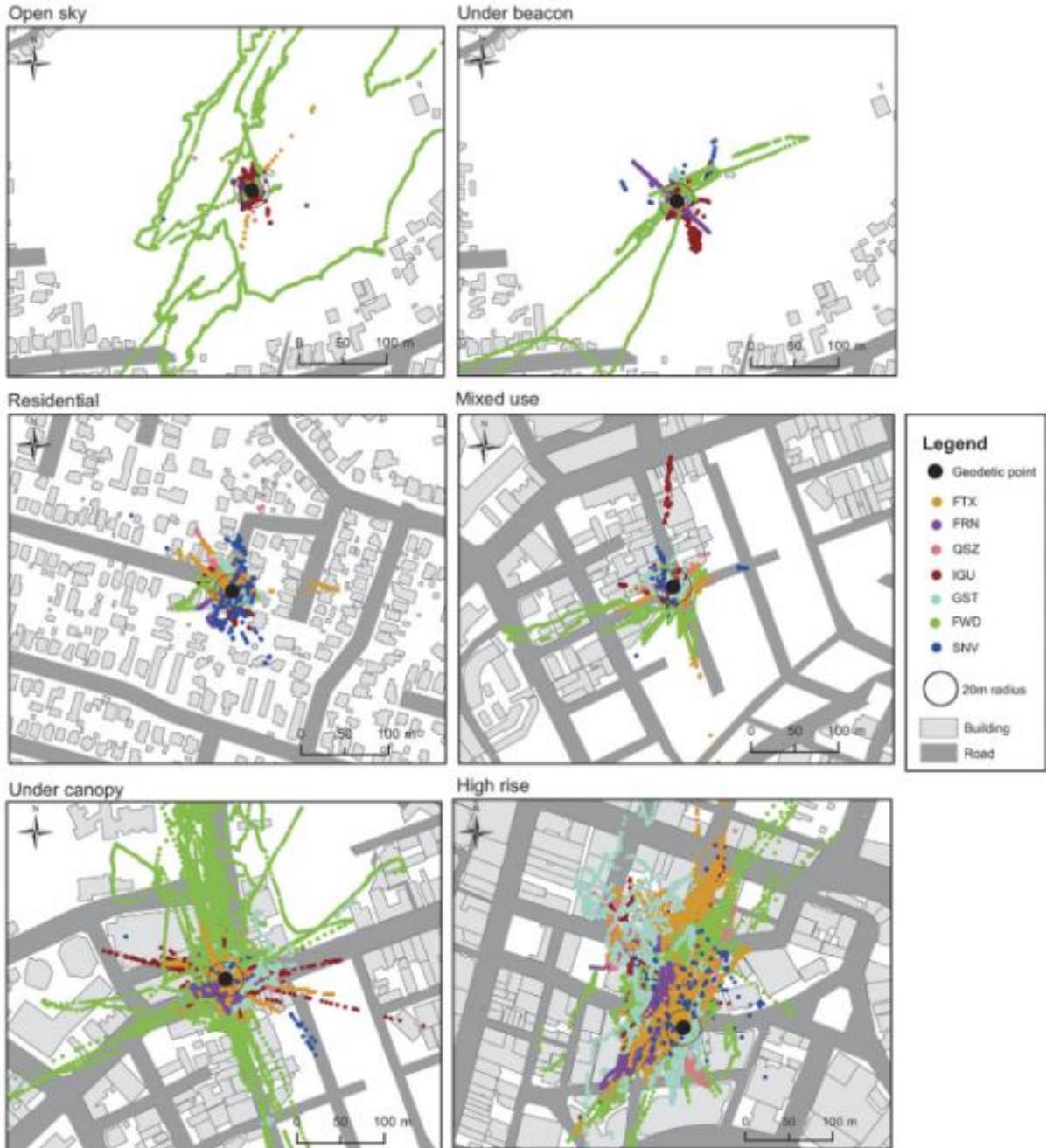


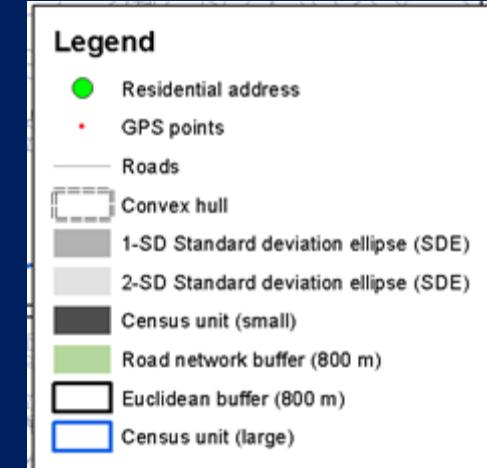
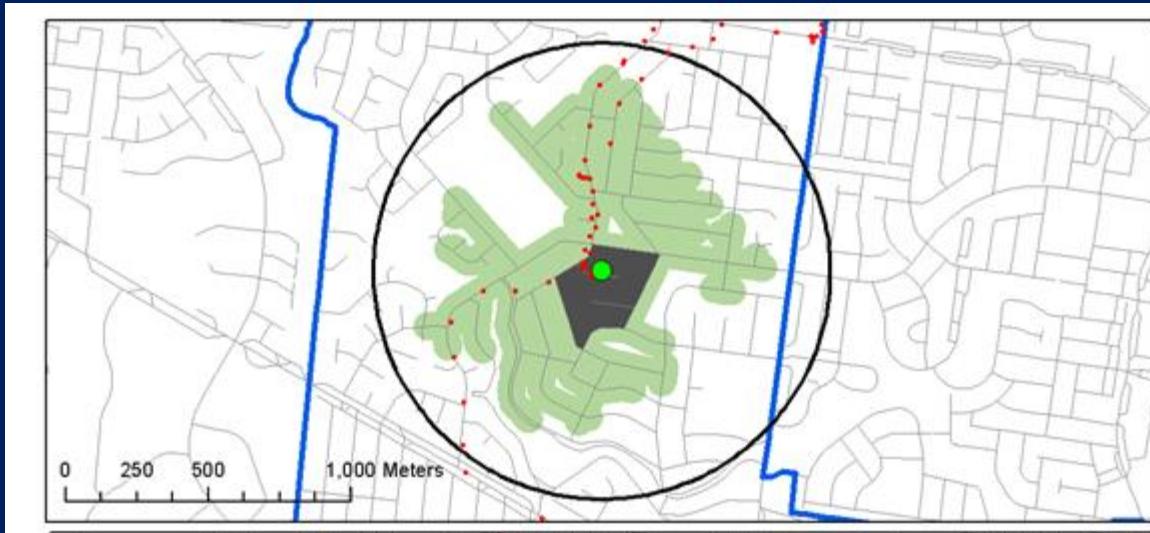
Figure 2. Geospatial distribution of GPS observations for each geodetic site

Note: FRN, Garmin Forerunner 205; FTX, Garmin Foretrex 201; PWD, FRWD B100; GST, GlobalSat TR-203; IGU, i-gotU GT-600 GPS Travel Logger; QSZ, BT-Q1000XT; SNV, StarsNav BTS-110

# Issues: Delineating exposure

What environment?

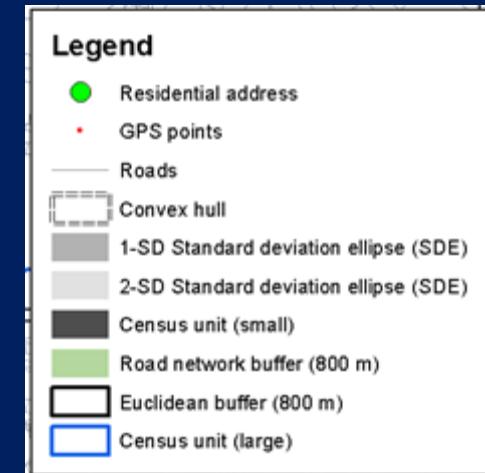
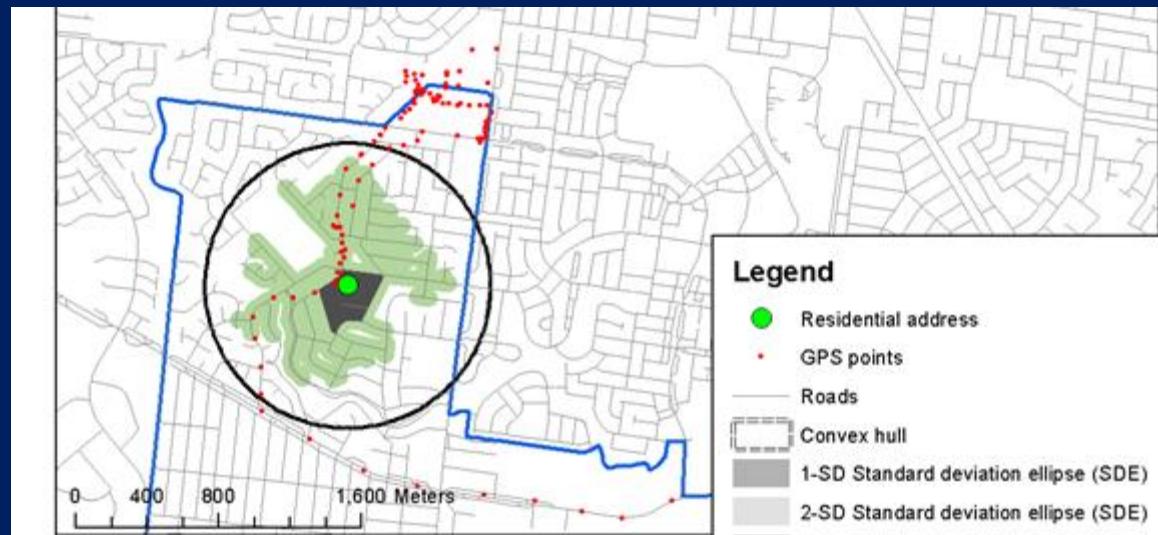
How to delineate that environment?



# Issues: Delineating exposure

What environment?

How to delineate that environment?



# Issues: Self-selection and scale

- Self-selection
  - Studies neglect issue of residential mobility and selection as alternative explanations to causation
  - Need to study movers or adjust for preference
- Scale
  - Scale and extent of spatial analysis needs to match the scale and extent of the actual phenomenon or error will be introduced.

# Issues: other

- Secondary data
  - Do they measure what is being examined?
  - Insufficient variation in the variables of interest between areas?
  - Lag time between exposures and disease manifestation?
- Arbitrary administrative boundaries (e.g. ABS)
  - Exposure misclassification of area level variables
  - Readily available data set area boundaries do not represent meaningful geographic communities of interest/ spatial patterning of phenomena
- The modifiable areal unit problem (MAUP)
  - Different results on how data are spatially distributed depending on the size and shape of the areal units.
- Ecological fallacy
  - Inferring individual relationships from relationships observed at a ecological level.

Thank you

[suzanne.mavoa@unimelb.edu.au](mailto:suzanne.mavoa@unimelb.edu.au)