Learning about *walkability* in Wellington with Python and open data

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What can you expect?

The slides contain

- An introduction to doing spatial data science
- A smattering of Python code
- Lots of graphs and figures
- Some insights about walkability in Wellington

Motivation

The global open data movement

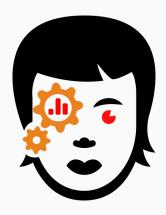
- Data sources
 - OpenStreetMap
 - Data.gov*: data.gov, data.gov.nz, data.gov.uk, data.gov.in
- Inspiring people generating tools / insights with open data
 - Kuan Butts
 - Geoff Boeing

Limited analysis of NZ

I want to learn about the coolest little capital in the world

- How are we being served?
 - housing
 - public transport
 - amenities
 - infrastructure
- Can things be better?

.. but I'm still just a Data Scientist



Audiens Cave

Walkability

Importance

Reducing car reliance and encouraging more transportrelated physical activity are now recognised as beneficial objectives from health, social and environmental perspectives. Evidence is accumulating that a number of built environment attributes are associated with the likelihood of residents using active transport.

- Measuring neighbourhood walkability in NZ cities

Need



Unlike cars, pedestrians are sensitive to their environment; changes to it can impact the *walking experience* or the *decision to walk*.

Task

We'll explore the impact of hilly terrain on walkability Specifically, on walkability to council playgrounds - an amenity that should be locally accessible on foot.

Python package set up

Main packages

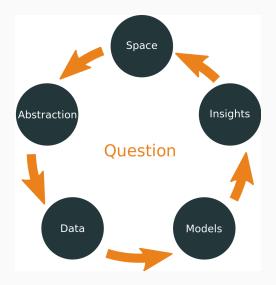
```
# Geoprocessing
import osmnx as ox  # Processing OSM as graphs
import networkx as nx  # Graph structure processing
import pandana as pa  # Efficient accessibility computing
import geopandas  # Processing geodataframes
from shapely.geometry import \
Point, Polygon, LineString  # Core geometric objects
```

Supporting packages

```
# Plotting
import matplotlib.pyplot as plt
                                      # Classic Python plotting package
import seaborn as sns
                                      # Ggplot2-like plotting in Python
import folium
                                      # Interactive, web-ready maps
                                      # Geometric patches for matplotlib
from descartes import PolygonPatch
# General utilities
import yaml
                                      # Reading stored API keys
                                      # Processing arrays and matrices
import numpy as np
import pandas as import pd
                                      # Processing dataframes
# Bayesian analysis
import pystan
                                      # Running Bayesian models
```

Spatial data science

Overview



Question

What is the impact of hills on walkability to playgrounds in Wellington?

Space

Wellington



Abstraction: spatial primitives

Points



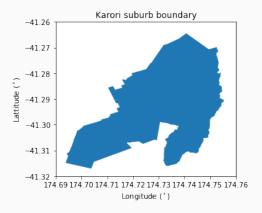
- Point coordinates of playgrounds
- Overlaid on map of Wellington

Lines



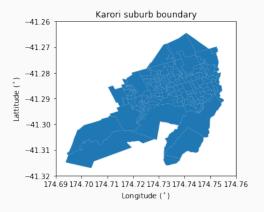
• Line segments that define a street or in this case, a route.

Polygons



Poylgon boundary of suburb

Complex abstractions



- Poylgon boundary of suburb
- Polygon boundaries of meshblocks within suburb

Abstraction: map to graph

Creating a street graph



Map represented as street edges with intersections as nodes

Data: spatial primitives

Spatial entities in geodataframes

```
(wlg_meshblock_suburbs_ov
.query('suburb == "Karori"')[['suburb', 'postcode', 'MB2019_V1_00', 'geometry']]
.head())
```

	suburb	postcode	MB2019_V1_00	geometry
917	Karori	6012.0	2104100	POLYGON ((174.7527410567269 -41.27200381936174
934	Karori	6012.0	2106100	POLYGON ((174.7533828776446 -41.28300381209932
942	Karori	6012.0	2105503	(POLYGON ((174.7530386517898 -41.2800340600980
944	Karori	6012.0	2149500	POLYGON ((174.7535317353109 -41.28234654562732
947	Karori	6012.0	2149400	(POLYGON ((174.754264330901 -41.27988740170357

Points, lines and polygons can all be compressed in a geodataframe.

Data: graphs

Graph nodes in geodataframes

```
nodes_gdfs, edges_gdfs = ox.graph_to_gdfs(G_sub)
nodes_gdfs.head()
```

	elevation	highway	osmid	x	у	geometry
1259077823	196.755	NaN	1259077823	174.792882	-41.227920	POINT (174.7928822 -41.22792)
1259077824	218.696	NaN	1259077824	174.791983	-41.229385	POINT (174.7919835 -41.2293852)
1259077827	163.804	NaN	1259077827	174.805433	-41.213698	POINT (174.8054327 -41.2136978)
3619684648	12.692	NaN	3619684648	174.780604	-41.276563	POINT (174.7806038 -41.2765628)
3619684652	12.344	NaN	3619684652	174.781234	-41.276037	POINT (174.7812341 -41.2760368)

Graph edges in geodataframes

edges_gdfs[['name', 'grade', 'osmid', 'maxspeed']].head()

		name	grade	length	osmid	maxspeed
1259077823	1259072929	Truscott Avenue	0.1319	66.800	110175609	50
	1259072943	Truscott Avenue	-0.0475	65.443	110175609	50
	6083853567	John Sims Drive	-0.1116	177.292	110176112	50
1259077824	6083853567	John Sims Drive	0.1650	13.022	110176112	50
1259077827	465611807	Cambrian Street	0.0396	71.272	107284021	50

Data used for analysis

- Street graph: with street gradient attribute for edges
- WCC playgrounds represented as points
- Suburb boundaries defined by WCC as polygons

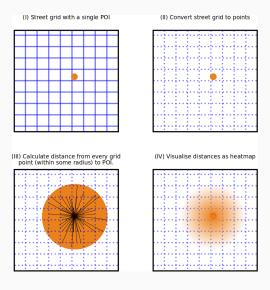
Model 1

Approximating walkability as accessibility

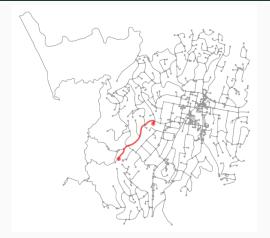
Just to make life confusing, there are several definitions of accessibility. For the following analyses, accessibility is:

- an objective metric
- calculated with a street graph and points of interest (POIs)
 - e.g. Wellington street graph and playground locations
- calculated with a specific unit of interest
 - e.g. distance, travel time, total travel time etc.
- limited to nearest POI

How to calculate accessibility

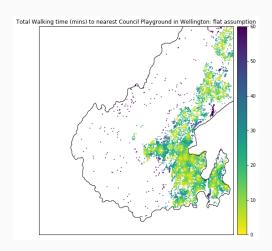


Accessibility on streets



- Find closest street graph nodes to: start and park
- Find shortest part between start and park nodes
- Sum edge weights of shortest path

Efficient accessibility with Pandana

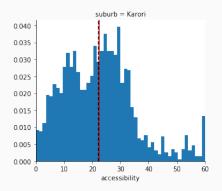


Model 2

Bayesian modelling

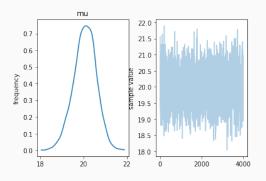
From the observed accessibility data, what is the average accessibility to a playground across the different Wellington suburbs?

Set up Bayesian model



- Model individual suburb accessibility (A_s) as a lower value truncated normal distribution.
- Normal distribution: $A_s \sim N(\mu, \sigma)$
- Truncation condition: $A_s \in [0, \inf]$

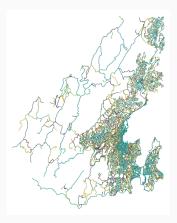
Efficient Bayesian modelling with Stan



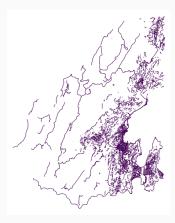
- Stan model output for μ (labelled as mu)
- Samples of μ drawn by Stan

Insights (mostly visual)

Street graph with gradients

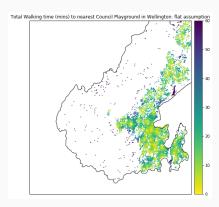


All edges (green ~ flat gradient)

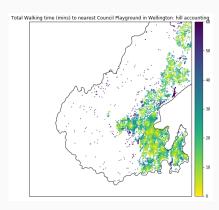


Edges within 5% absolute gradient

Hills vs. flat land

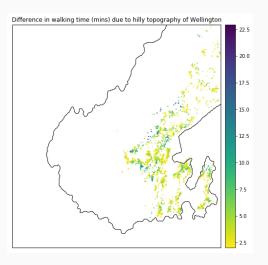


Assuming single speed

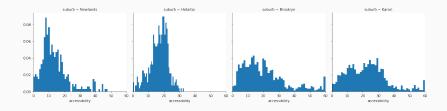


 Accounting for speed variability due to hills

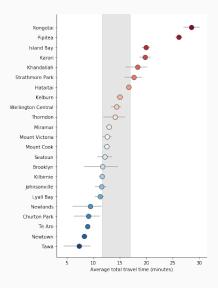
Impact of hills on playground accessibility



Accessibility by suburb



Modelling playground accessibility by suburb



What have we learned?

- Hills have an impact on total travel time. Differences in total travel time can be up to 20 minutes.
- Wellington suburbs average 12-17 minutes in total travel time to nearest playground.
- But, there is a large variation within suburbs.

Lots more work to do!

- Making graphs better
- Removing the 60 minute spikes from the Bayesian modelling
- Exploring the heterogeneity in accessibility within suburbs
- Impact of including school playgrounds in the analysis
- Impact of adding a new council playground (e.g. Berhampore playground coming in \sim 2020)

Resources

- Write up on https://shriv.github.io
- Code in https://github.com/shriv

Image Credits

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