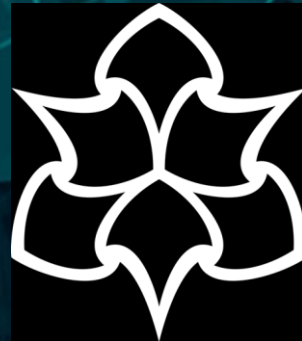


*Daily Rhythms 1: Population
Denominators and Spatio-Temporal
Crime Hotspots*

**Manchester Metropolitan University Crime & Well-Being Big Data
Centre**

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Contents

1. Population denominators and crime hotspots – literature
2. Our Contribution
3. Results
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Population measures

1. Residential Population
2. Workplace Population (Malleon & Andresen, 2016)
3. Ambient Population (Mburu & Helbich, 2016; Andresen, 2011; Bogomolov et al., 2014)

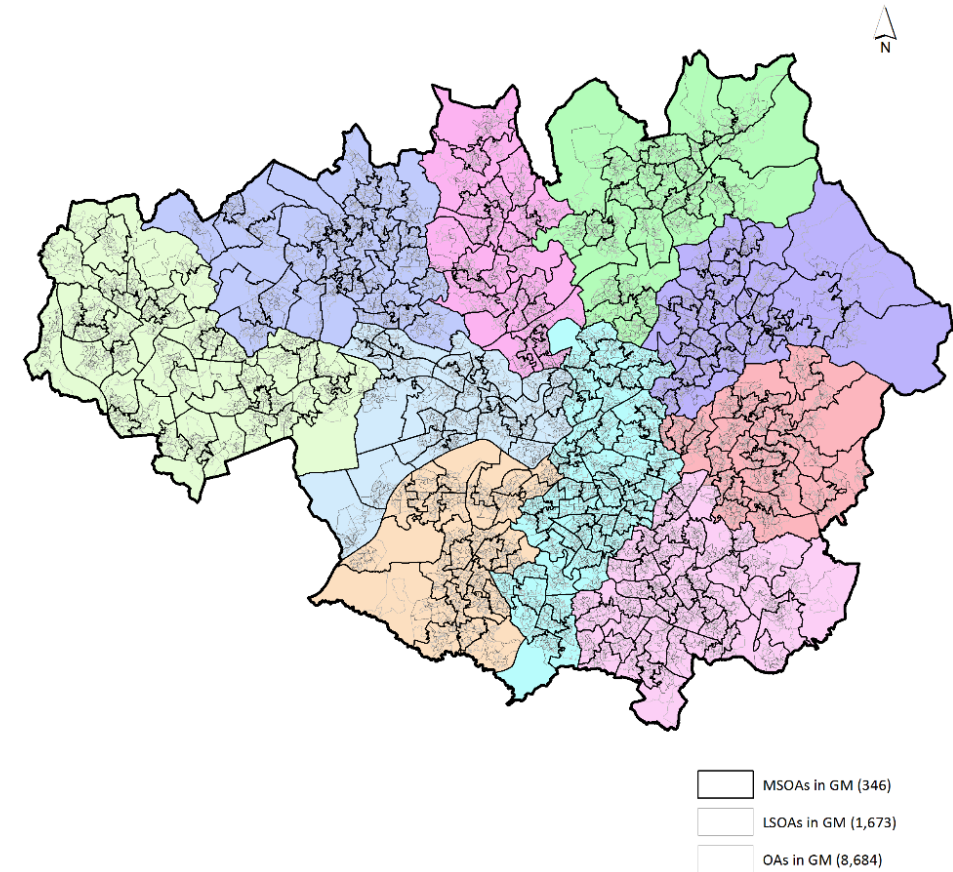
Research Questions

Q1. Do these population estimates capture the daily rhythms of the city?

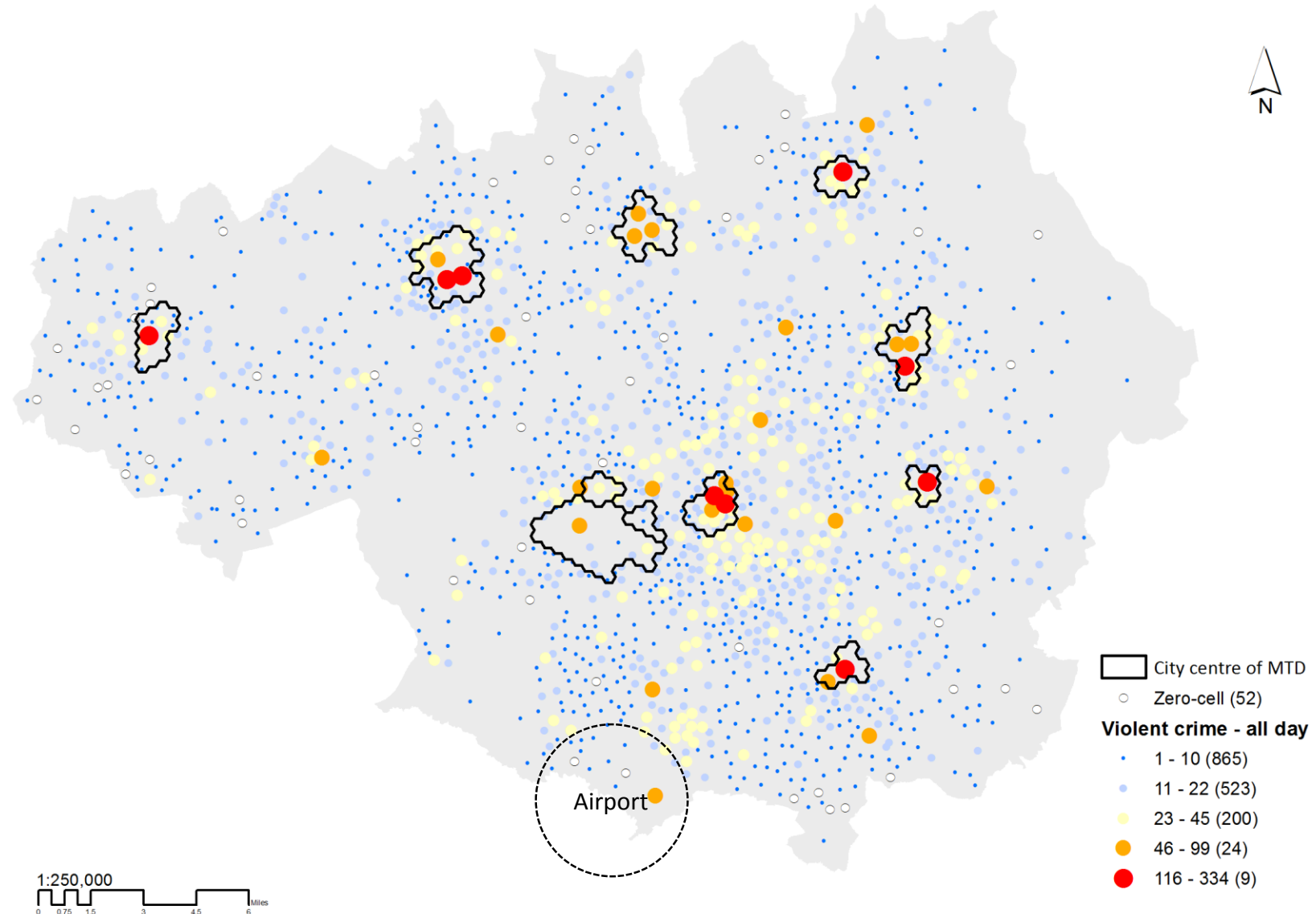
Q2. Do they represent the population-at-risk?

Our Greater Manchester data

- **Mobile Phone Origin Destination (MPOD)** matrices from Transportation of Greater Manchester (TfGM)
 - Average number of persons travelling per trip from Origin A to Destination B in different time bins (T_0, T_1, \dots, T_N)!
 - Each trip assigned with a flag to represent final Destination (1- YES, 0-No)
- **Crime Data** from Greater Manchester Police (GMP)
 - Attributes: Crime records with spatial coordinates and time-stamps / (non domestic)



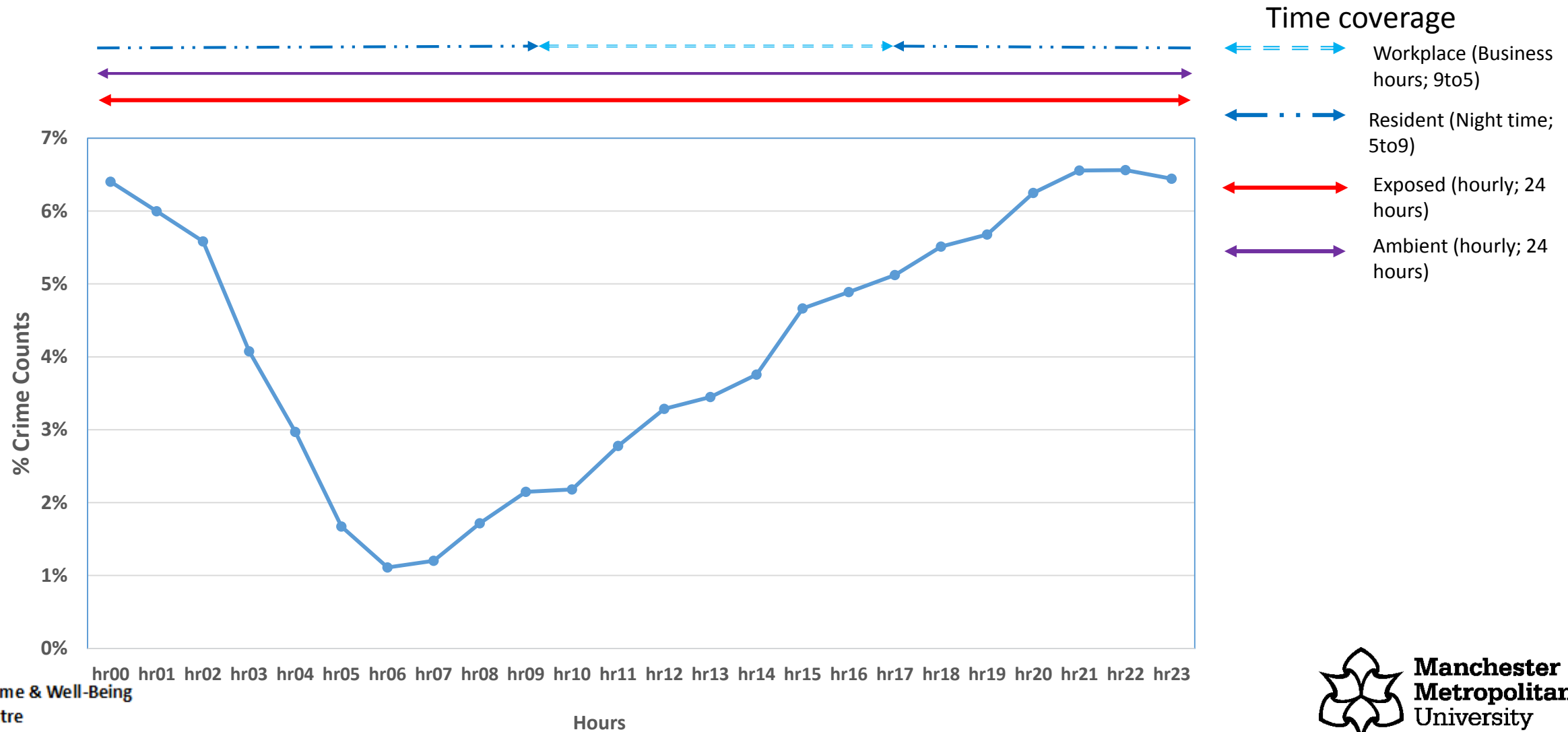
The spatial patterning of violent crime counts



Conventional correlation analysis (daily population measure)

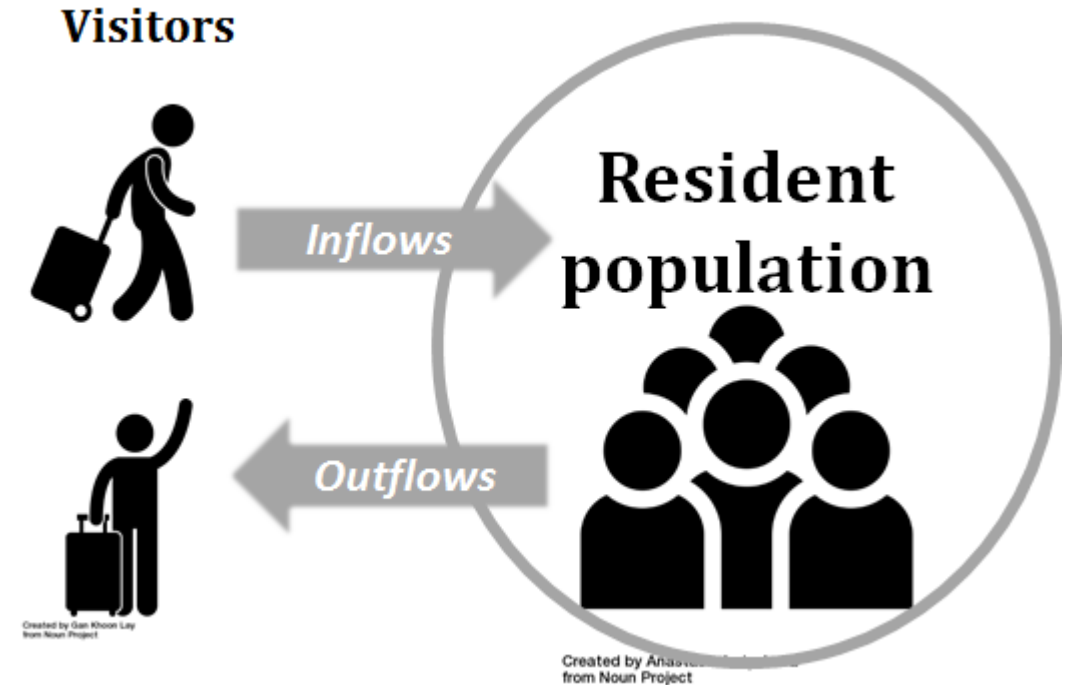
Total crime	Residential pop	Workplace pop
Pearson Correlation	.202**	.626**
Sig. (2-tailed)	.000	.000
N	1673	1673
**. Correlation significant at 0.01 level (2-tailed).		

The temporal patterning of violent crime



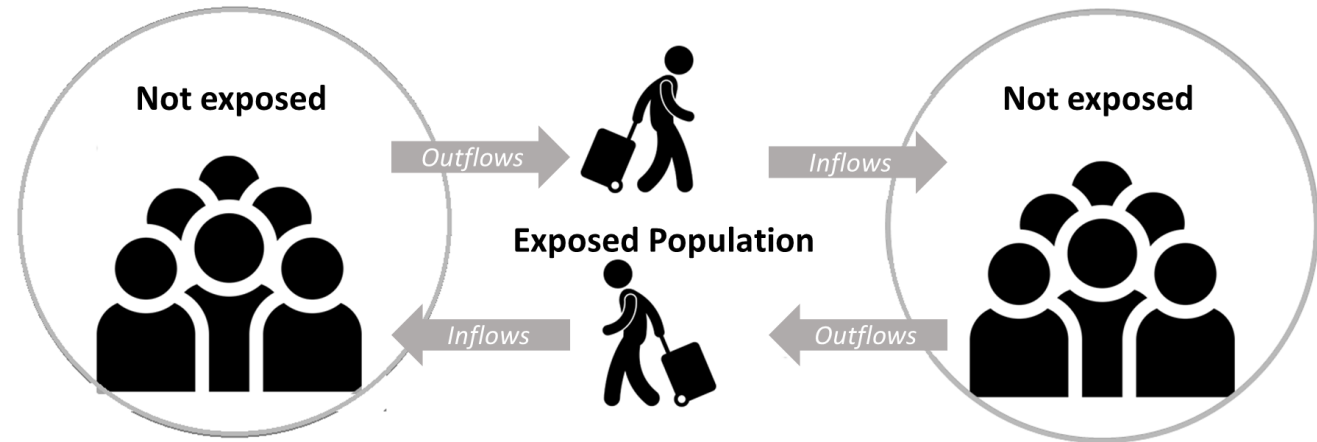
Ambient (total) population

- Based on the incremental flows of the residential population in different time bins (T_0, T_1, \dots, T_N)!
- Assuming the ambient population to be equivalent to residential population at T_0 (midnight)
 - $Amb_pop_{T_0} = Resid_pop + Inflows_{T_0} - Outflows_{T_0}$
 - $Amb_pop_{T_1} = Amb_pop_{T_0} + Inflows_{T_1} - Outflows_{T_1}$



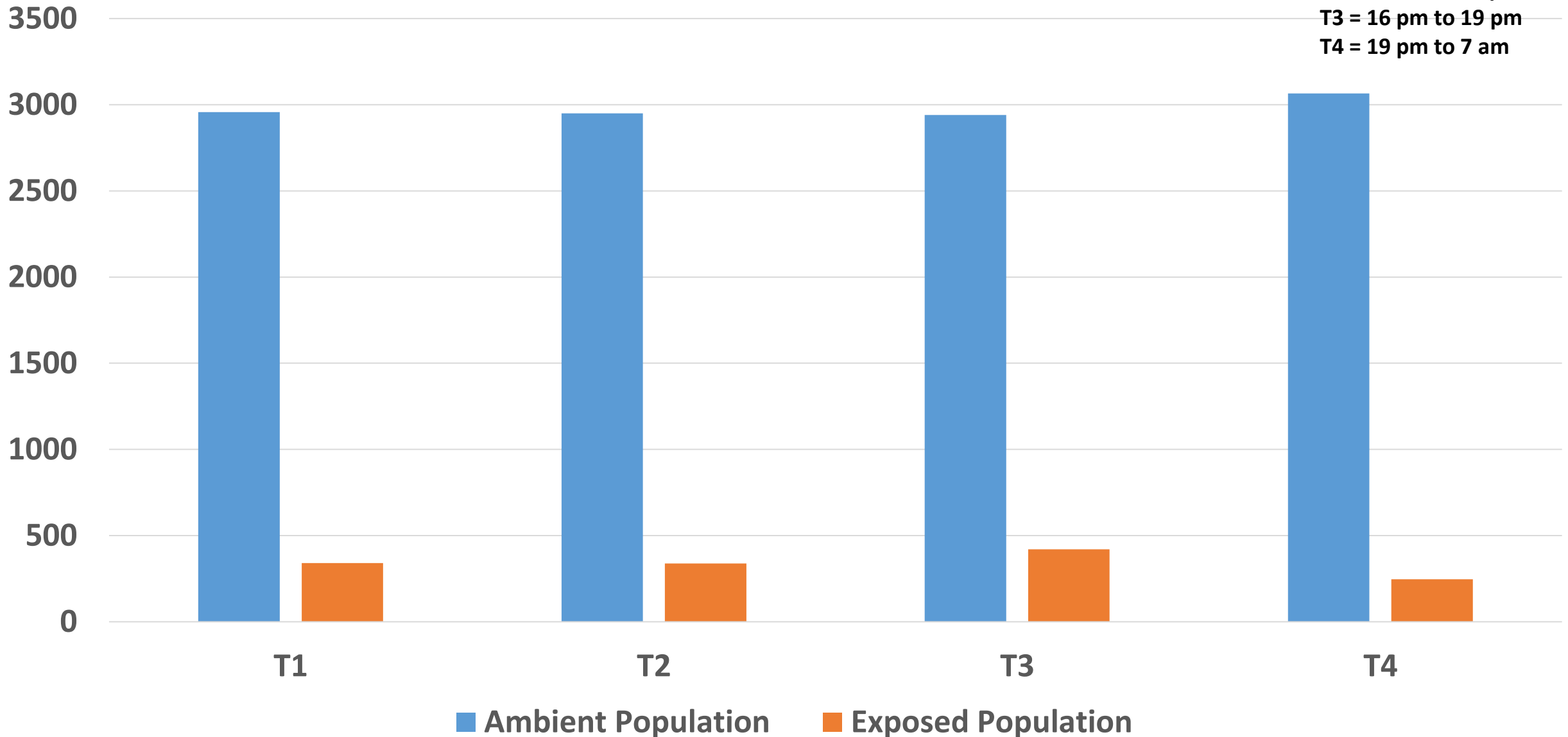
Exposed (mobile) population

- Based on determining the street based population at different time bins (T_0, T_1, \dots, T_N)!
- This excludes the population who have reached their **final destination** or have left the area (**initial origin**) at particular time
- $$\text{Exp_pop_}T_0 = \text{Inflows_}T_0 + \text{Outflows_}T_0 - \text{Inflows_}T_0\text{_{FD}} - \text{Outflows_}T_0\text{_{FO}}$$

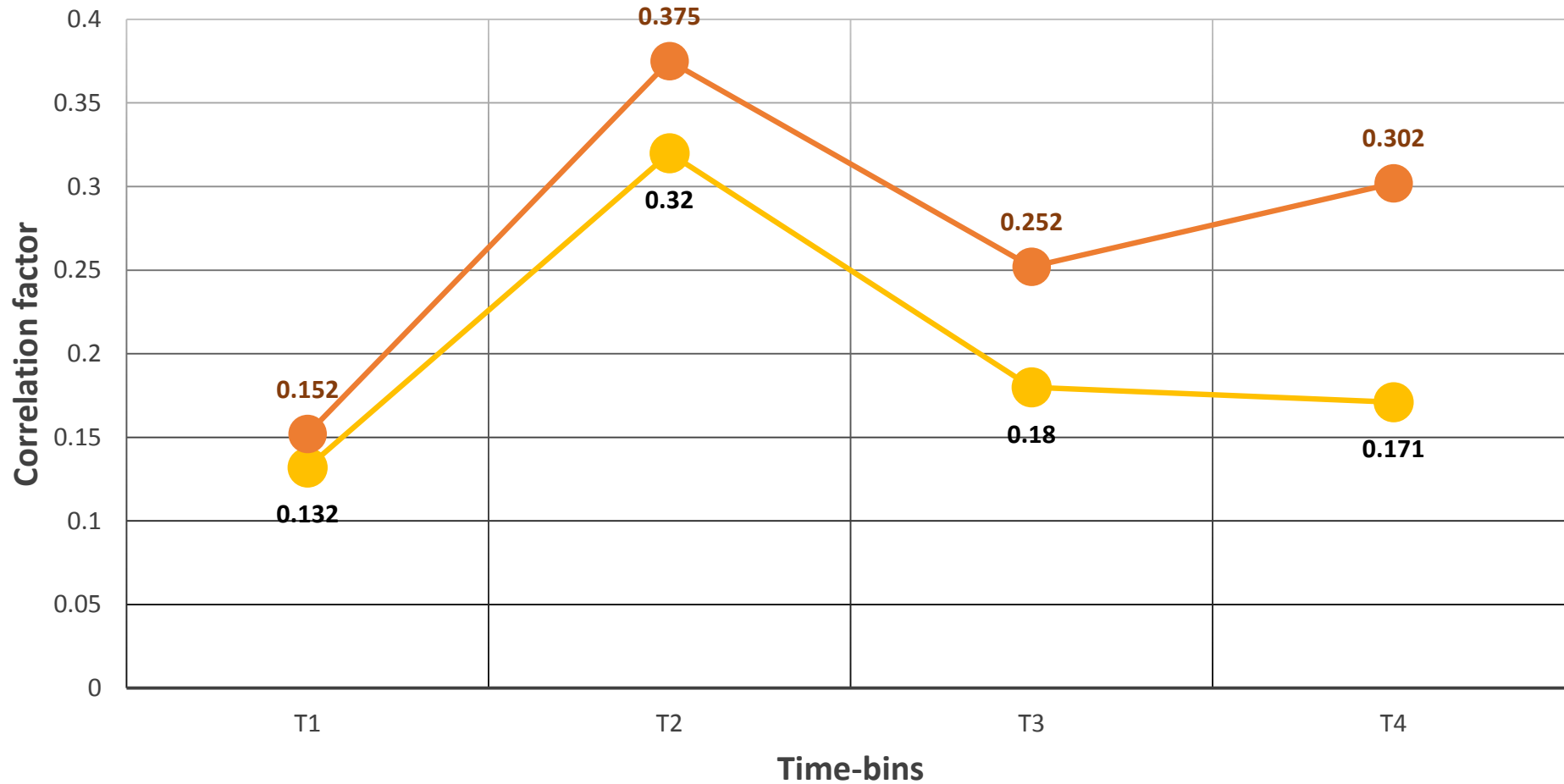


Time sensitive ambient and exposed populations

T1 = 7 am to 10 am
T2 = 10 am to 16 pm
T3 = 16 pm to 19 pm
T4 = 19 pm to 7 am



Time sensitive correlation analysis



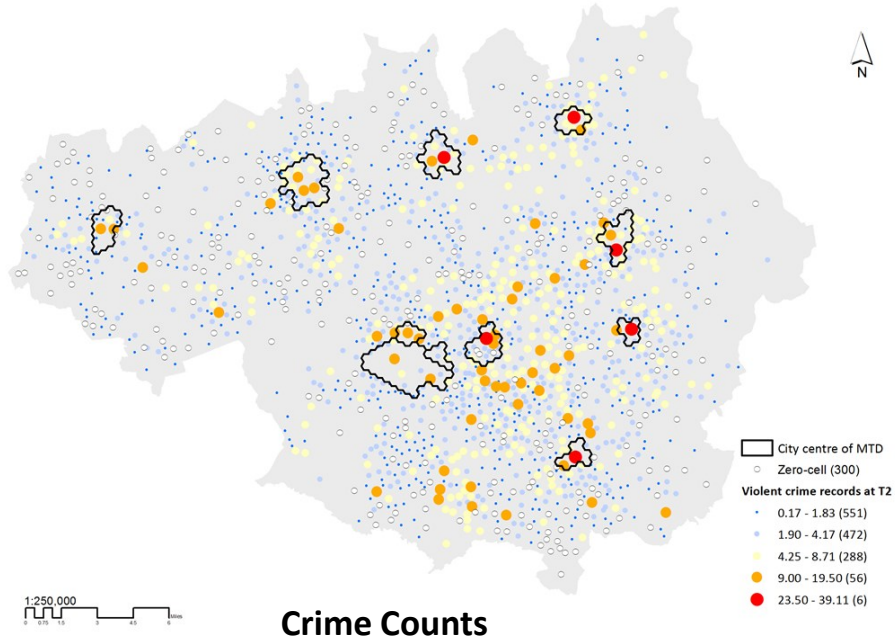
T1 = 7 am to 10 am
T2 = 10 am to 16 pm
T3 = 16 pm to 19 pm
T4 = 19 pm to 7 am

Hotspot analysis

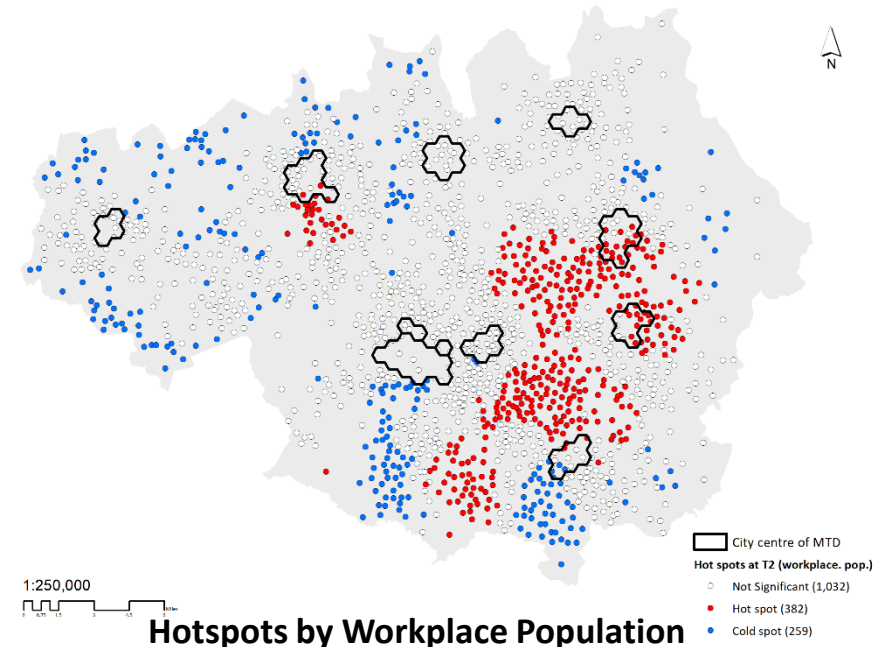
- G_i^* statistics (Getis-Ord's G_i^*)
- Z-score of G_i^* (p-value <0.05) enables identification of high or low value clusters of spatial units.
- Question – to what extent does the method of spatio-temporal clustering impact on the detection of hotspots?

Violent Crime Count and Crime Rate Hotspot Analysis by Different Population Denominators in Different Time Bins

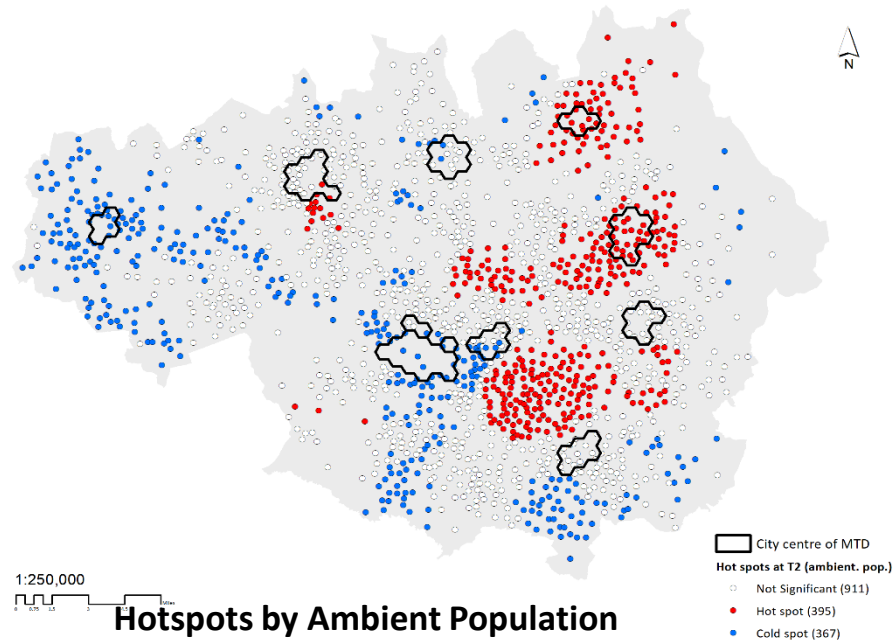
T2 = 10 am to 16 pm



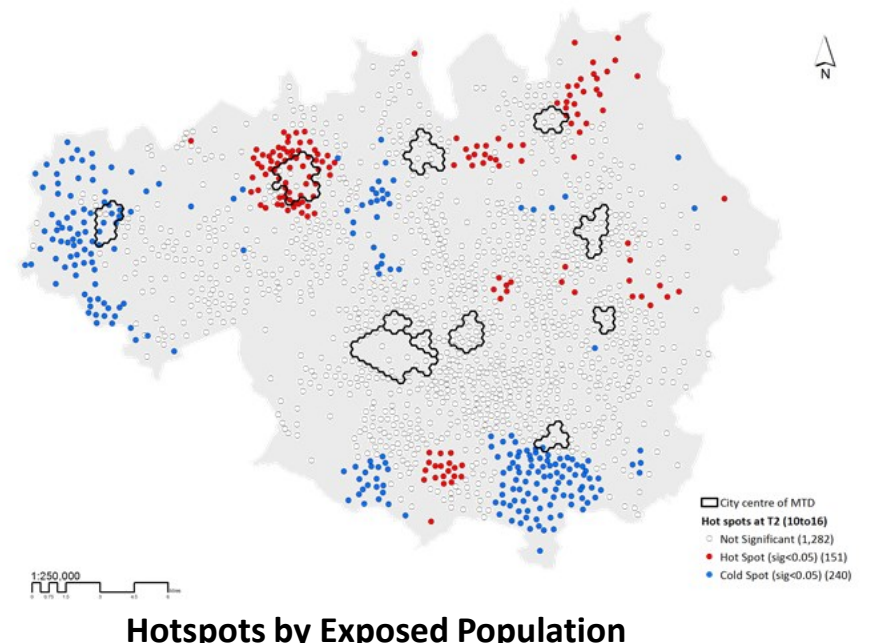
Crime Counts



Hotspots by Workplace Population

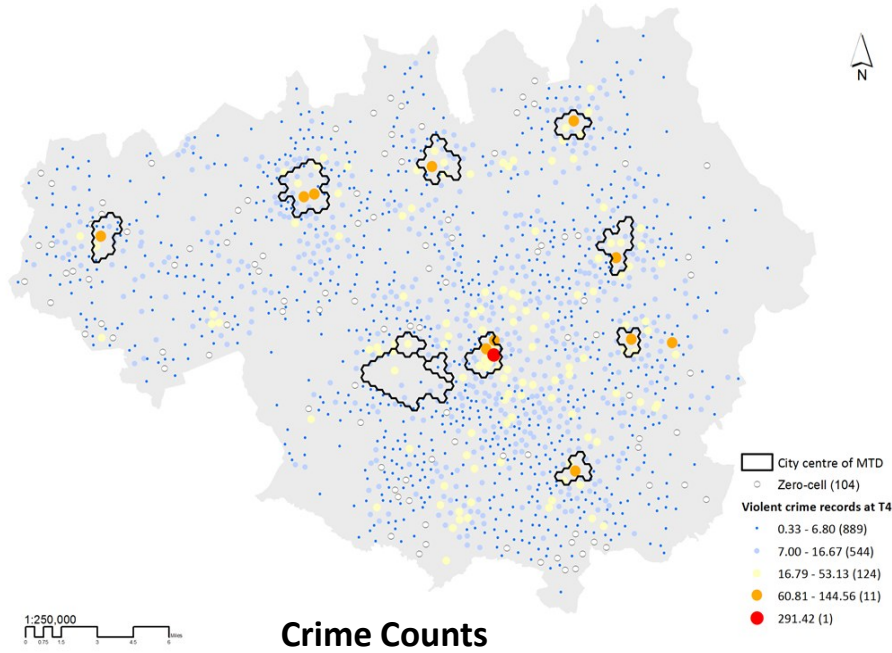


Hotspots by Ambient Population

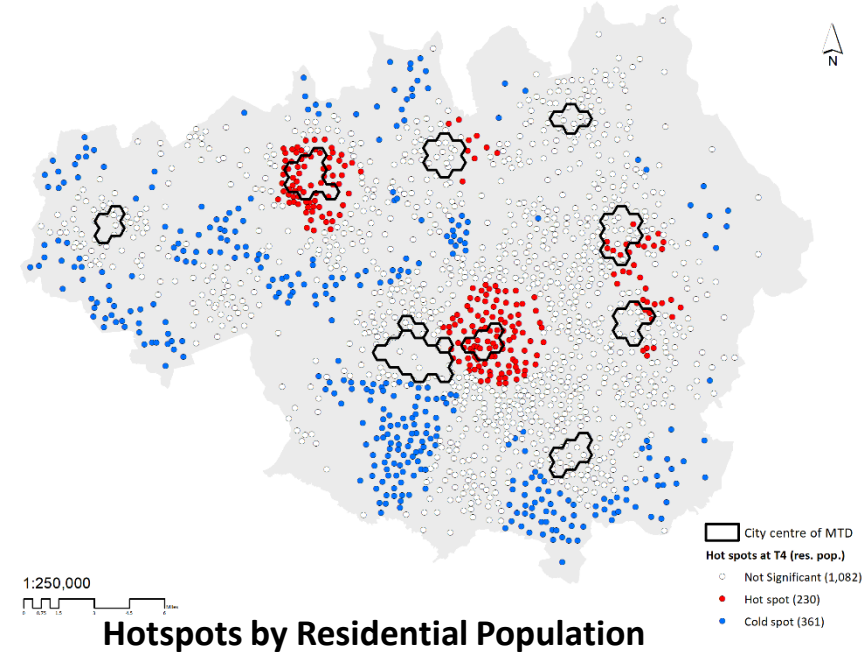


Hotspots by Exposed Population

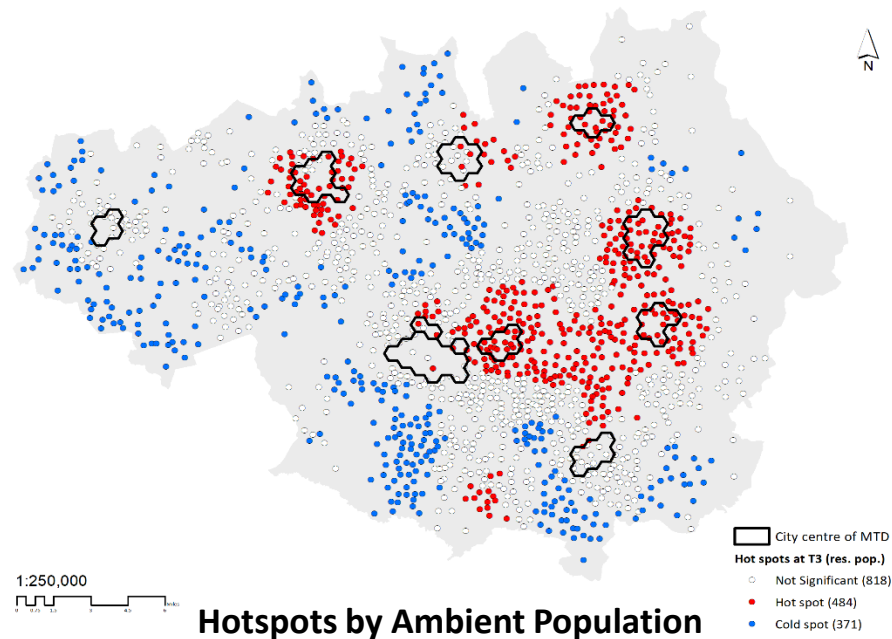
T4 = 19 pm to 7 am



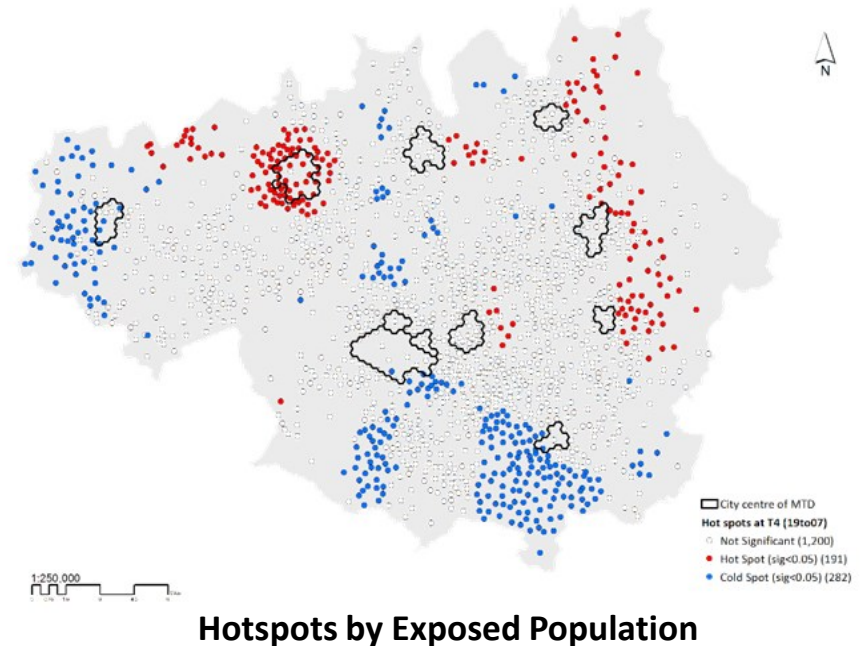
Crime Counts



Hotspots by Residential Population



Hotspots by Ambient Population



Hotspots by Exposed Population

Conclusion

- The daily rhythms of the city – ambient and exposed populations.
- The exposed (theoretically correct?) population holds a higher correlation with the violent crime than the ambient population across multiple time bins.
- Different population denominators generate markedly different hotspots.
- Population denominators require to be sensitive to crime type.

Questions?

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