



# Using NTS data to explore frequency and rhythms of food shopping

Giulio Mattioli

Centre for Transport Research, University of Aberdeen

# “Gross polluters for food shopping travel - a practice-based typology”

Two **methodological challenges**:

1. How to **investigate the frequency** of activities
2. How to **highlight meaningful patterns of variation** in energy-demanding (mobility) practices

# Why National Travel Survey data?

- Theme 1 ambition to investigate the timing and rhythm of energy-demanding practices (Walker, 2014).. But what about **activities with less than daily frequency?**
- 2 days in British 2000 TUS clearly inadequate
- 6 weeks ideally needed for travel behaviour (Schlich & Axhausen, 2003)
- **British NTS**
  - **advantages: 7 consecutive days**, all household members, continuous survey, large sample, includes information (distance, vehicles) allowing to compute energy / emissions estimates
  - **limitations:** only trips reported, short walks (<1 mile) only on 7<sup>th</sup> day

# Why food shopping?

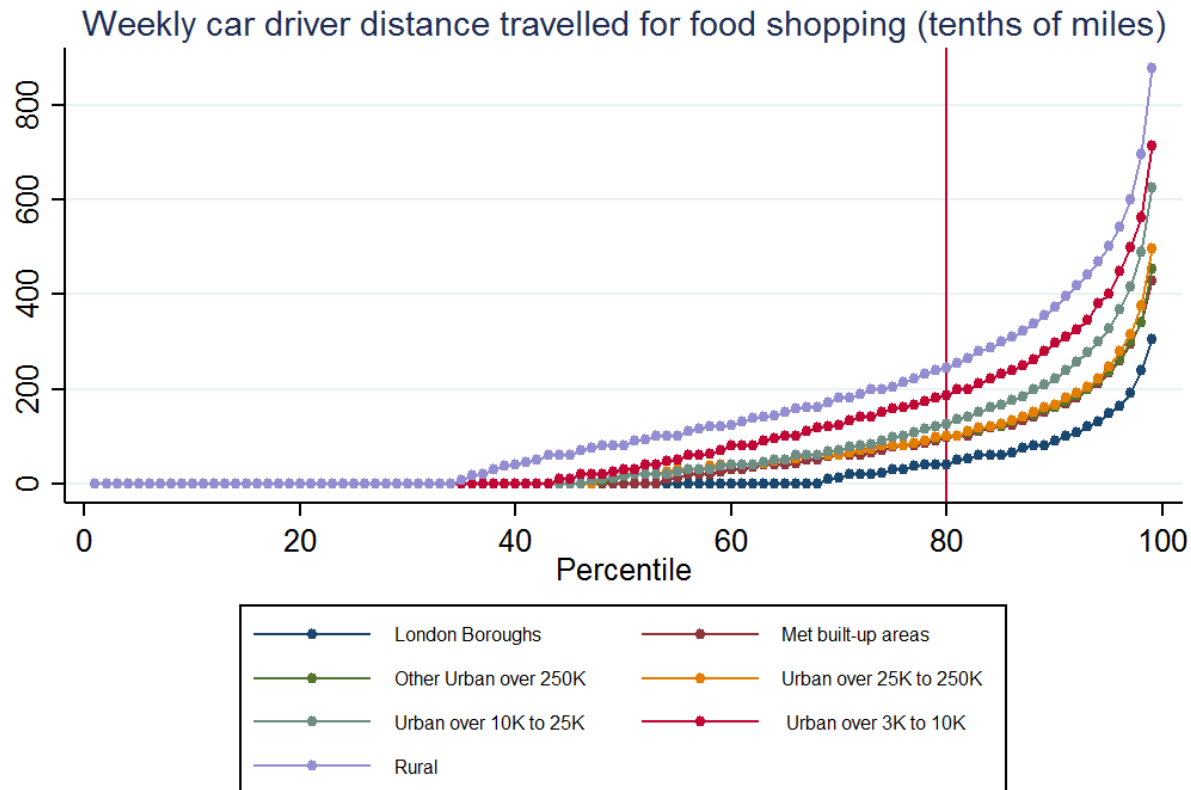
- generally **less-than daily frequency but at least once a week** (Bhat et al., 2004): can assume that behaviour in travel diary is a **representative of a more general pattern**
- best analysed at the **household level**
- **substantive interest:**
  - substantial amount of travel
  - particularly and increasingly car dependent (until recently)
  - mostly studied from a built environment perspective – what about frequency?

# Beyond aggregate averages – looking for variation

Three sequential steps:

1. Focus on a **single practice**: food shopping (travel)
2. Focus on '**gross polluters**' = households in the top 20% of weekly car driver distance for food shopping (proxy for CO<sub>2</sub> emissions) (cfr. Brand & Boardman, 2008)
3. **Cluster analysis** to highlight variation within the top 20%

# 'Gross polluters' for food shopping travel



NTS 2002-2010 - own elaboration

# Cluster analysis

- Classification and segmentation techniques **common in 'ABC' research**
- CA has been **used to investigate 'clusters of practice'** based on own-survey data (Browne et al., 2014)...
- **...but conceptually tricky** to cluster practices based on information at the individual / household level
- notably when using existing datasets
- degree of arbitrariness in **choice of input variables**
- **NTS travel diary week data rarely used for clustering** (Mattioli, 2014)

# Input variables

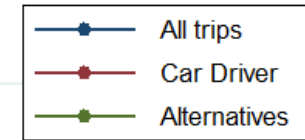
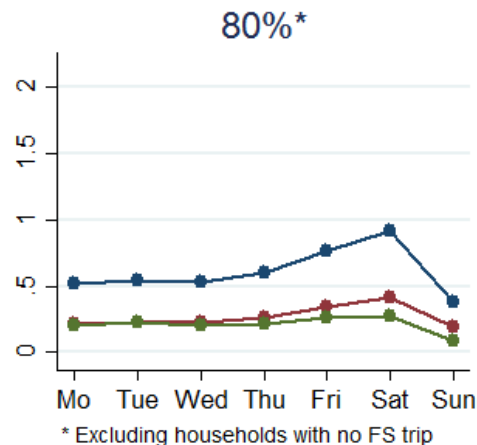
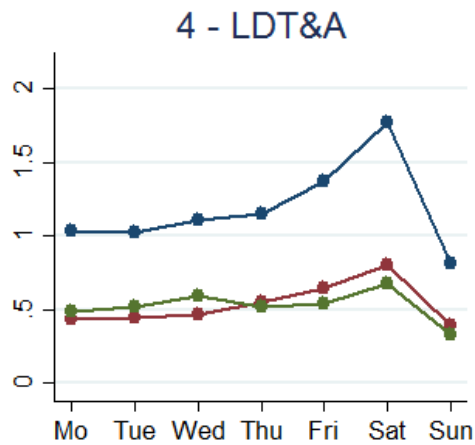
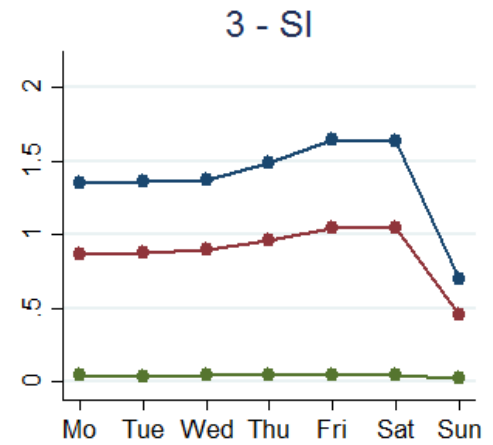
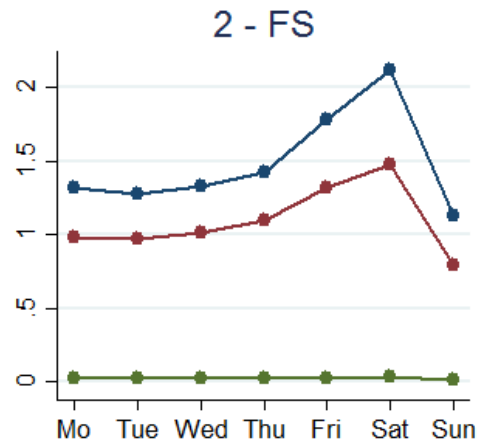
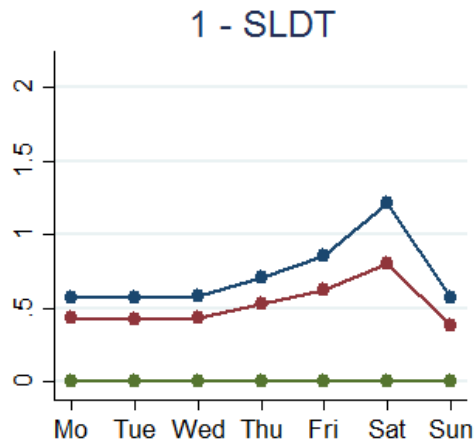
- |  |   |
|--|---|
| <b>1 Distance</b>                      | Percentile of car driver distance travelled (within type of area)       |
| <b>2 Concentration</b>                 | % of distance accounted for by longest trip                             |
| <b>3 Frequency</b>                     | Total number of car driver trips in travel week diary                   |
| <b>4 Alternative modes</b>             | % of trips by modes other than car driver / passenger                   |
| <b>5 Shopping intensity (distance)</b> | % of total household car driver distance accounted for by food shopping |
| <b>6 Shopping intensity (time)</b>     | % of total household travel time accounted for by food shopping         |



# 4 cluster solution

	Single long distance trip Cluster 1	Frequent shopping Cluster 2	Shopping intensive Cluster 3	Long distance trip & alternatives Cluster 4
Size (%)	44.9%	37.1%	11.7%	7.3%
Distance (percentile)	36	<b>64</b>	<b>64</b>	43
Concentration	<b>50%</b>	28%	34%	<b>49%</b>
Frequency	3.3	<b>6.8</b>	<b>5.5</b>	3.3
Alternatives	0.4%	1.3%	2.3%	<b>40.3%</b>
Shopping intensity – distance	12%	16%	<b>57%</b>	17%
Shopping intensity – travel time	9%	13%	<b>43%</b>	16%

# Trip rates by day of the week



NTS 2002-2010 - own elaboration

# Socio-demographic profile

	1 - Single Long Distance Trip	2 - Frequent Shopping	3 - Shopping Intensive	4 - Long Distance Trip & Alternatives	MEAN
<b>DEMOGRAPHICS</b>					
Household size	2.6	2.8	2.0	2.7	<b>2.6</b>
HRP over 60 years old	26%	28%	68%	31%	<b>31%</b>
HRP non-employed	19%	20%	67%	26%	<b>25%</b>
Pensioner household	17%	17%	57%	19%	<b>21%</b>
<b>INCOME</b>					
Lowest or second quintile	26%	26%	54%	31%	<b>29%</b>
<b>Mobility difficulties</b>					
Mobility difficulties (foor or bus)	16%	20%	43%	17%	<b>20%</b>

# Conclusions

- Frequency (not just distance) as a problem. No strong impact of accessibility: built environment matters, but it is only part of the story
- Focusing on **top 20% for a specific practice brings to light partly different social groups** than Top 20% in terms of overall transport emissions (high income, rush hour of life, etc., cfr. Brand & Boardman, 2008)
- **Who should be the target for policy?**
- Clustering based on patterns of activity over a week, then description based on characteristics of individuals / households. Is this an **useful approach?**

**Thank you for your attention!**

[www.demand.ac.uk](http://www.demand.ac.uk)

[giulio.mattioli@gmail.com](mailto:giulio.mattioli@gmail.com)