Social and Spatial Inequalities Research Group



Creating a Geodemographic Classification

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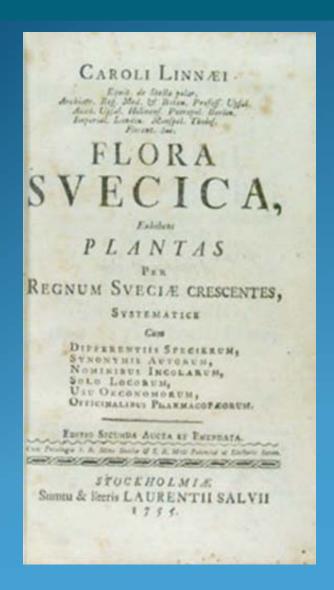
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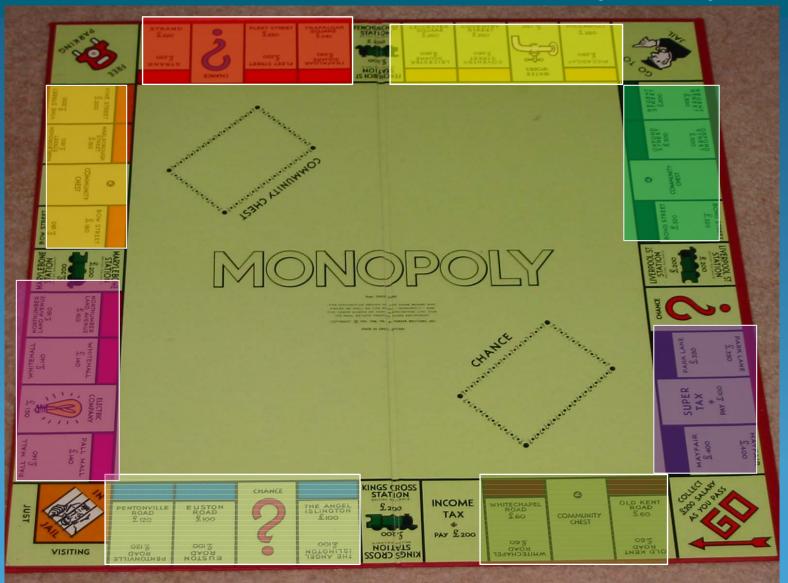
www.areaclassification.org.uk

Why classify areas?



"All the real knowledge which we possess depends on methods by which we distinguish the similar from the dissimilar. The greater number of natural distinctions this method comprehends the clearer becomes our idea of things. The more numerous the objects which employ our attention the more difficult it becomes to form such a method and the more necessary." (Linnæus 1737)

Why classify areas?



The Myths of Geodemographics

- > The more data the better?
- The smaller the areas the more meaningful?
- These are distinct groups?
- >A true depiction of reality?

The Objective of Creating OAC



- To provide a publicly accessible classification, free at the point of use for researchers, policy makers and the public, avoiding the expense of commercial classifications.
- To provide a transparent and replicable methodology with the input data as well as the classification output.
- To stimulate value added research either using the classification or the raw data.

The Seven Steps of Cluster Anlysis

- 1. Clustering elements (objects to cluster, also known as "operational taxonomic units")
- 2. Clustering variables (attributes of objects to be used)
- 3. Variable standardisation
- 4. Measure of association (proximity measure)
- 5. Clustering method
- 6. Number of clusters
- 7. Interpretation, testing and replication

(adapted from Milligan 1996)

Milligan, G. W. (1996) Clustering validation: Results and implications for applied analyses. in Arabie, P., Hubert, L. J. and De Soete, G. Eds., *Clustering and Classification*. Singapore: World Scientific.

What Goes into the Classification

- Output Areas are the smallest area for general census output.
 - >223,060 in the UK
 - ➤ England & Wales
 - ➤ Number of OAs: 174,434
 - ➤ Min size: 40 households, 100 people
 - ➤ Mean size: 124 households, 297 people
 - ➤ Scotland
 - ➤ Number of OAs: 42,604
 - ➤ Min size: 20 households, 50 people
 - ➤ Mean size: 52 households, 119 people
 - Northern Ireland
 - ➤ Number of OAs: 5,022
 - ➤ Min size: 40 households, 100 people
 - ➤ Mean size: 125 households, 336 people

What Goes into the Classification

- 41 Census Variables covering:
- > Demographic attributes
 - Including age, ethnicity, country of birth and population density
- > Household composition
 - Including living arrangements, family type and family size.
- > Housing characteristics
 - Including tenure , type & size, and quality/overcrowding
- Socio-economic traits
 - Including education, socio-economic class, car ownership & commuting and health & care.
- > Employment attributes
 - Including level of economic activity and employment class type.

How many data inputs are involved? 223,060 Output Areas, 41 Variables = 9,145,460 data points

Standardising the Variables

- Log Transformation
 - Reduce the effect of extreme values
- ➤ Range Standardisation (0-1)
 - ➤ Problems will occur if there are differing scales or magnitudes among the variables. In general, variables with larger values and greater variation will have more impact on the final similarity measure. It is necessary therefore to make each variable equally represented in the distance measure by standardising the data.

Clustering the Data

K-means clustering

K-means is an iterative relocation algorithm based on an error sum of squares measure. The basic operation of the algorithm is to move a case from one cluster to another to see if the move would improve the sum of squared deviations within each cluster (Aldenderfer and Blashfield, 1984).

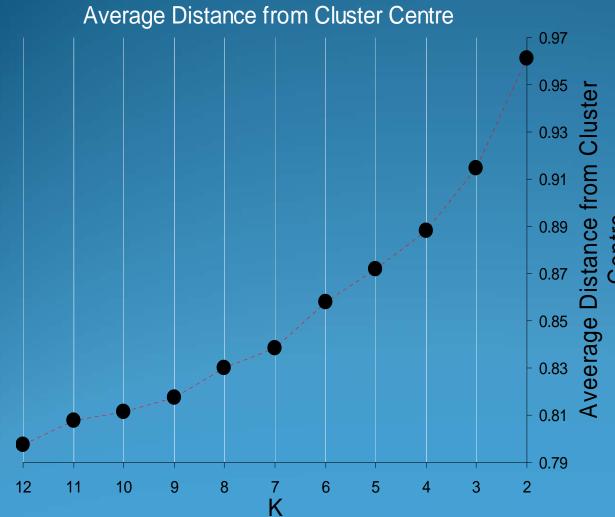
The case will then be assigned/re-allocated to the cluster to which it brings the greatest improvement. The next iteration occurs when all the cases have been processed. A stable classification is therefore reached when no moves occur during a complete iteration of the data. After clustering is complete, it is then possible to examine the means of each cluster for each dimension (variable) in order to assess the distinctiveness of the clusters (Everitt *et al.*, 2001).

Issues of Cluster Number Selection

- When choosing the number of clusters to have in the classification there were three main issues which need to be considered.
- ➤ **Issue 1**: Analysis of average distance from cluster centres for each cluster number option. The ideal solution would be the number of clusters which gives smallest average distance from the cluster centre across all clusters.
- ➤ **Issue 2**: Analysis of cluster size homogeneity for each cluster number option. It would be useful, where possible, to have clusters of as similar size as possible in terms of the number of members within each.
- ➤ **Issue 3**: The number of clusters produced should be as close to the perceived ideal as possible. This means that the number of clusters needs to be of a size that is useful for further analysis.

Issues of Cluster Number Selection

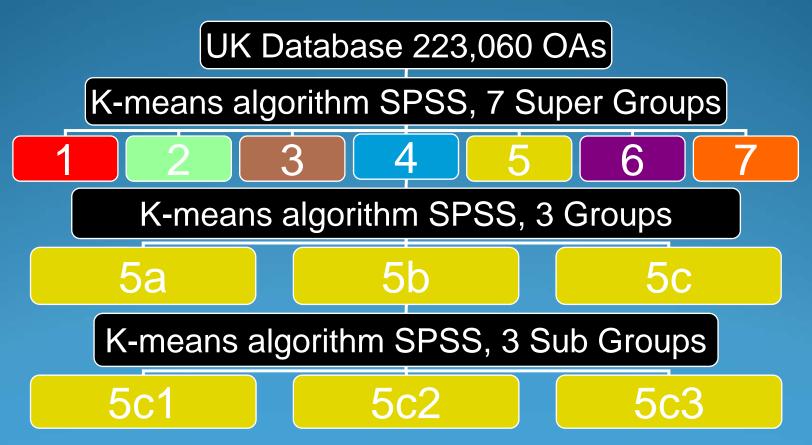
A three tier hierarchy 7, 21 & 52 clusters



- First Level target 6, 7 selected based on analysis of, average distance from cluster centre and size of each cluster.
- Second Level target 20, 21 selected based on analysis of, average distance from cluster centre and size of each cluster.
 - ➤ Third Level target 50, 52 selected based on size of each cluster. Split into either 2 or 3 groups

Clustering: Using k-means to Create a Hierarchy

- First level run as standard k-means
- ➤ Second level, first level is split into separate files and each file is clustered separately
- >Third level, second level is split into separate files and each file is clustered separately

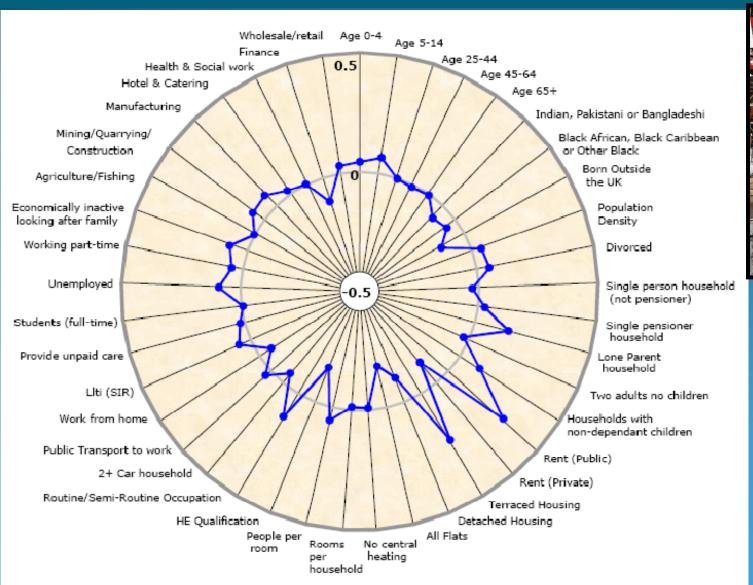


Cluster Names

7b: Afro-Caribbean Communities

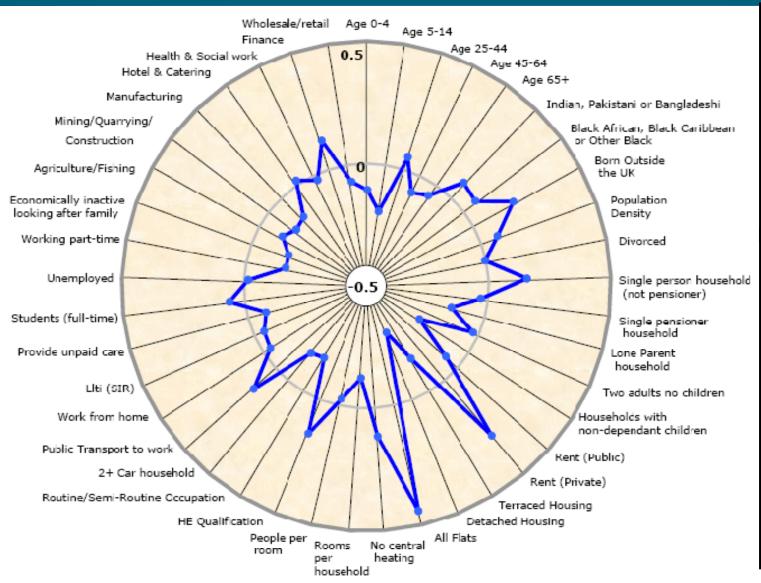
1: BLUE COLLAR COMMUNITIES	1a: Terraced Blue Collar
	1b: Younger Blue Collar
	1c: Older Blue Collar
2: CITY LIVING	2a: Transient Communities
	2b: Settled in the City
3: COUNTRYSIDE	3a: Village Life
	3b: Agricultural
	3c: Accessible Countryside
4: PROSPERING SUBURBS	4a: Prospering Younger Families
	4b: Prospering Older Families
	4c: Prospering Semis
	4d: Thriving Suburbs
5: CONSTRAINED BY CIRCUMSTANCES	5a: Senior Communities
	5b: Older Workers
	5c: Public Housing
6: TYPICAL TRAITS	6a: Settled Households
	6b: Least Divergent
	6c: Young Families in Terraced Homes
	6d: Aspiring Households
7. MIII TICIII TURAI	7a: Asian Communities

1: Blue Collar Communities





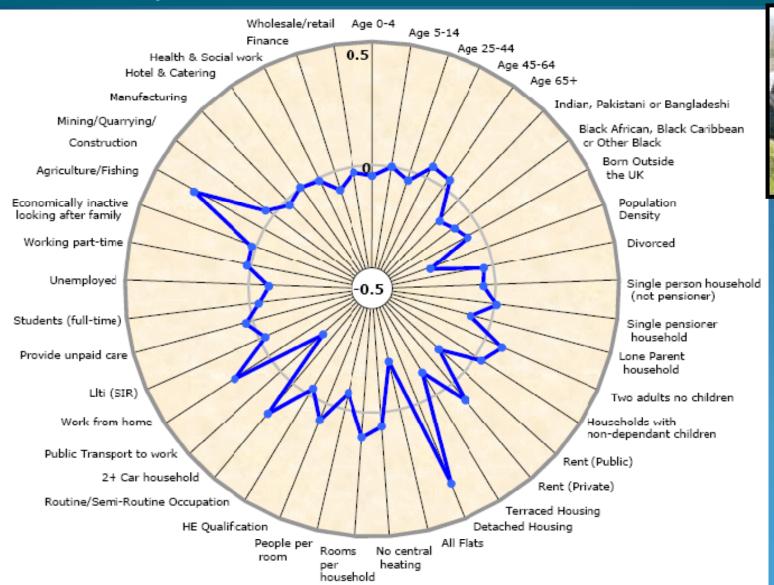
2: City Living





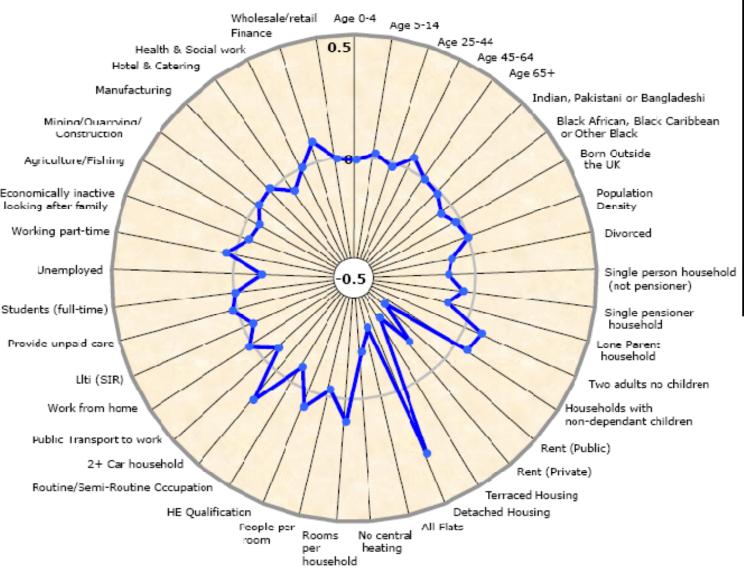


3: Countryside



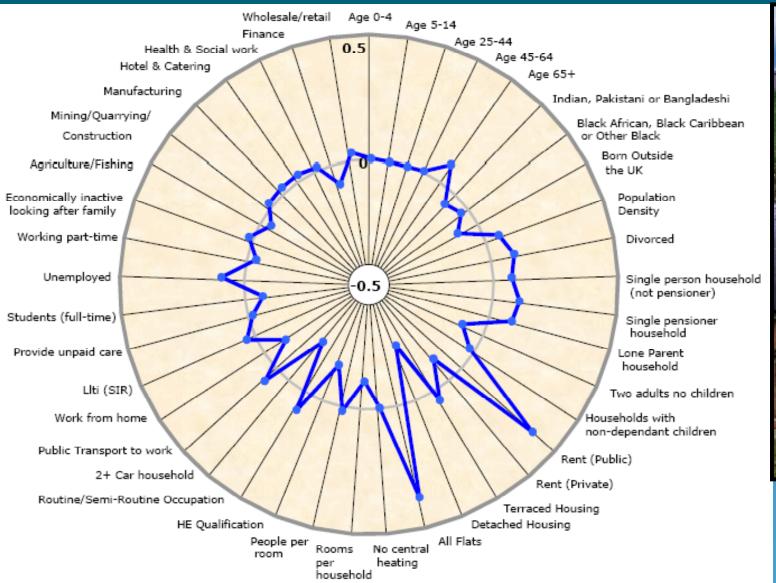


4: Prospering Suburbs



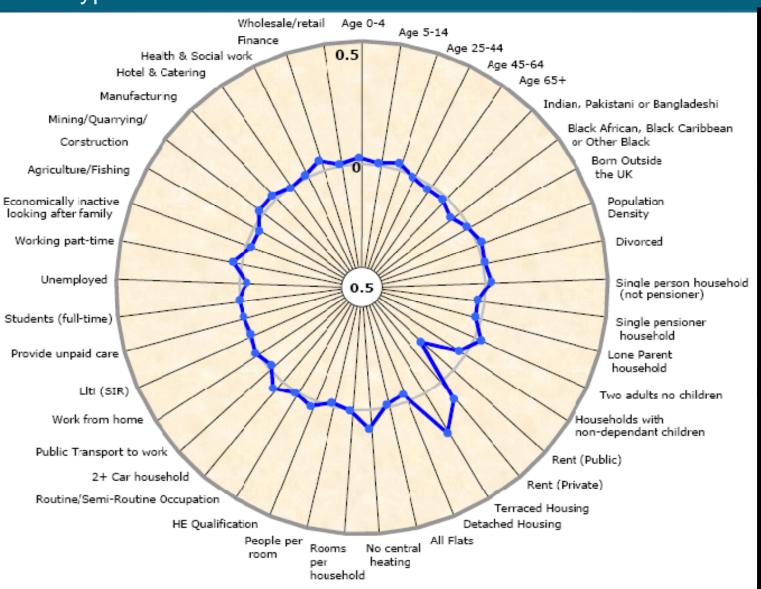


5: Constrained by Circumstances





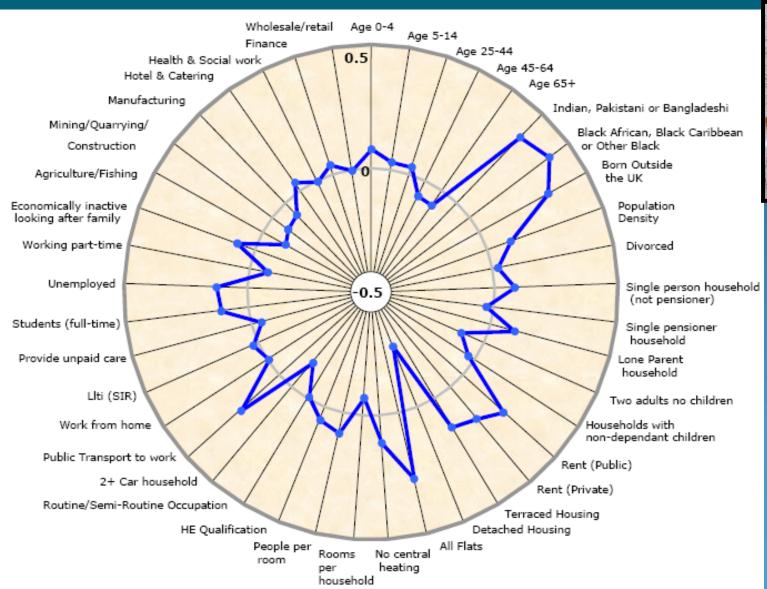
6: Typical Traits



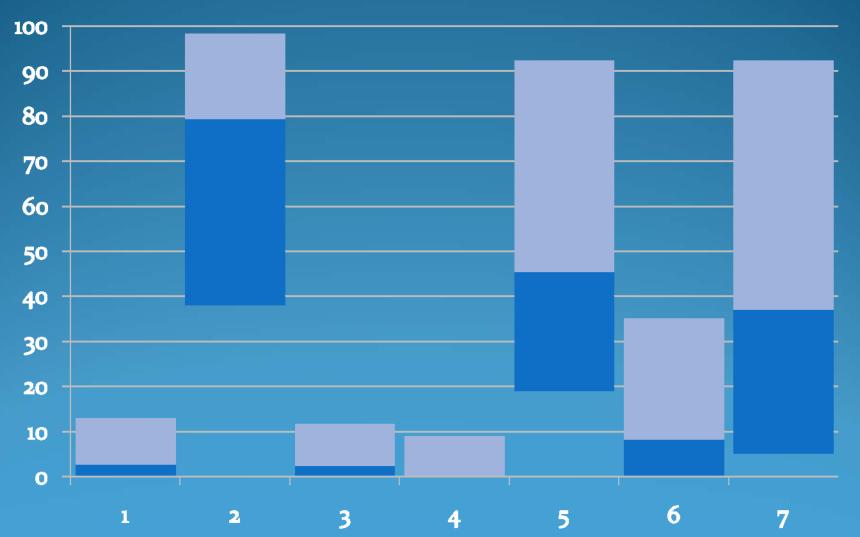




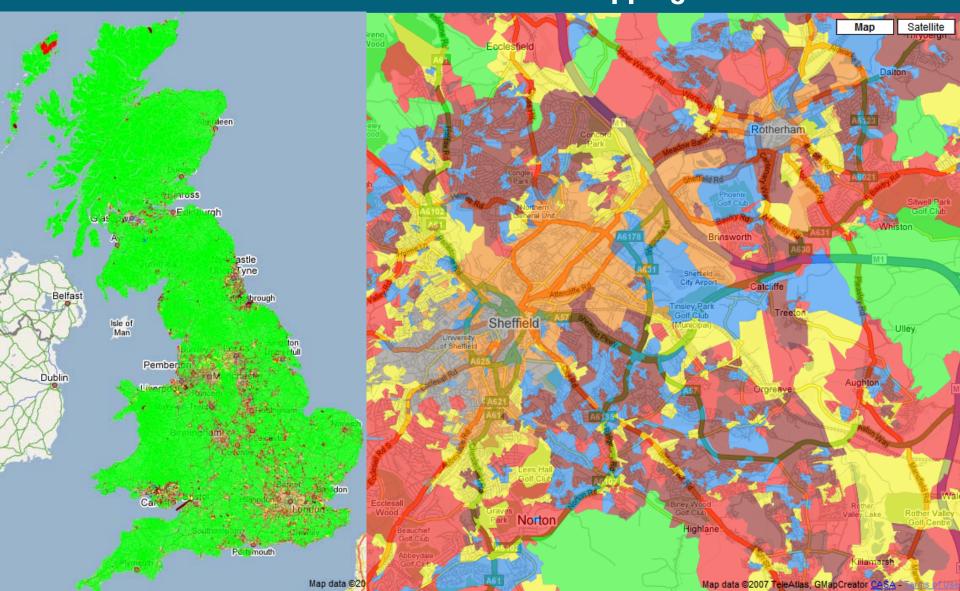
7: Multicultural







Mapping the Classification



Further Information

