Department of Geography

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Online exploration of cultural regions, migration and ethnicity using the geography of personal names

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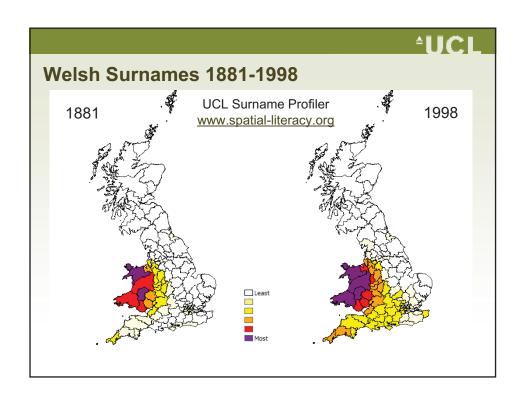
www.onomap.org www.publicprofiler.org

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Surname Geography

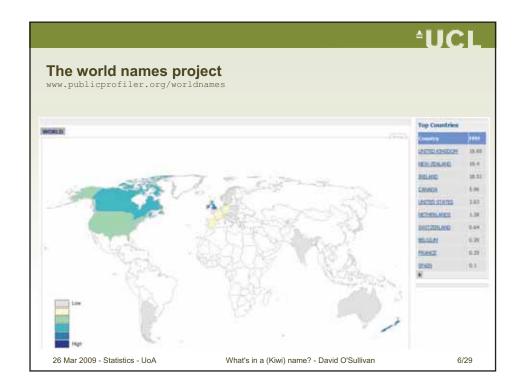


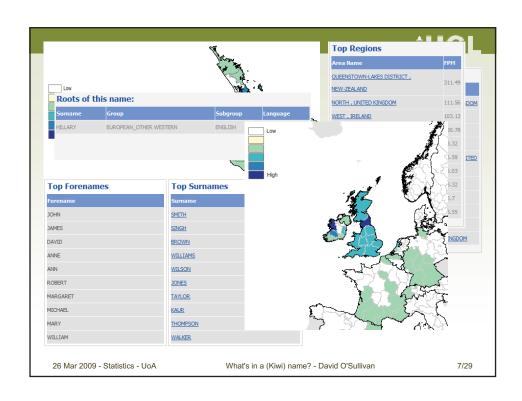
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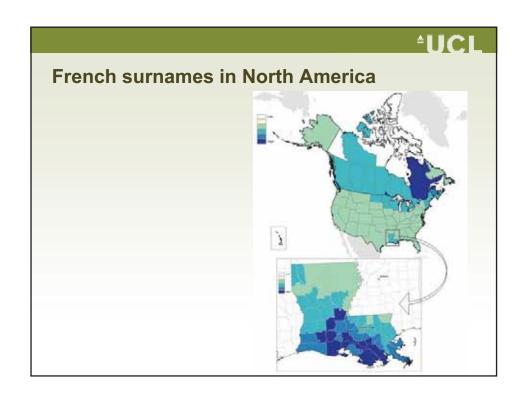
Publicprofiler WorldNames

www.publicprofiler.org/worldnames

- 26 countries in America, Europe, Asia and Oceania
- Covering a total population of 1 billion people
- Individual level data for 300 million people (full name and address)
- 8 million unique surnames
- 5 million unique forenames
- · Postcode geographies







Welsh surnames in Argentina Community of Welsh speakers in Patagonia Welsh migration and settlement in the 19th century.

Mapping names

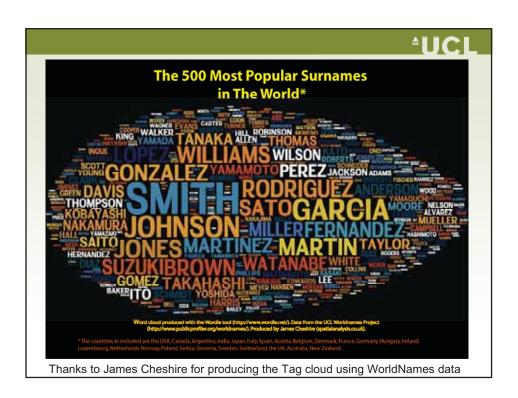
- · The whole database is available on-line
- · Mapping historic and contemporary migrations

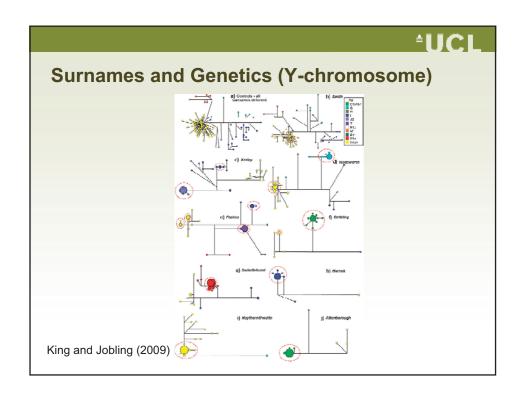
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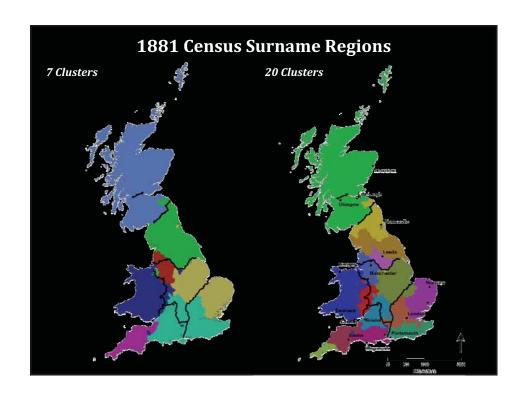
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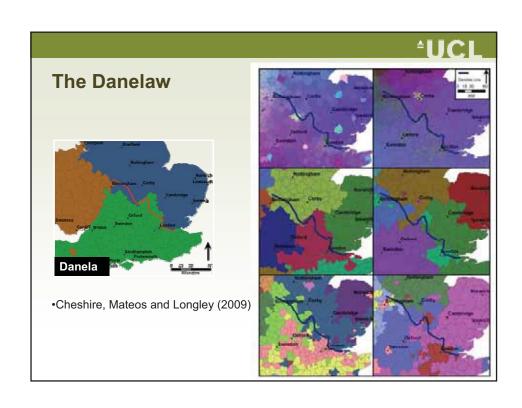
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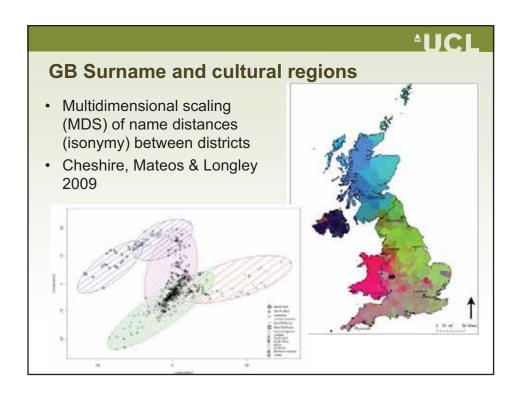


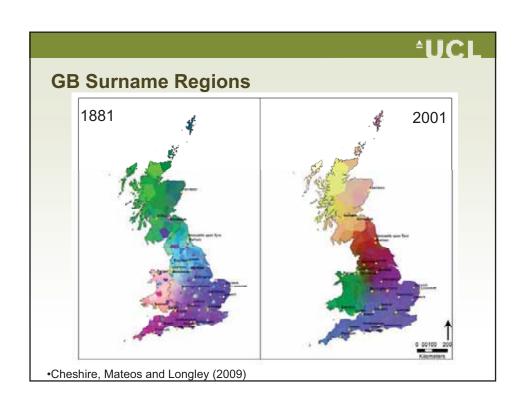


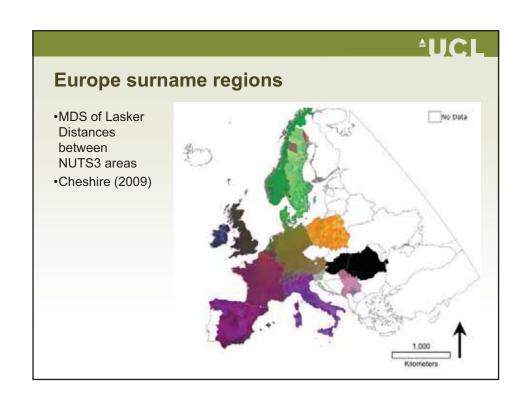


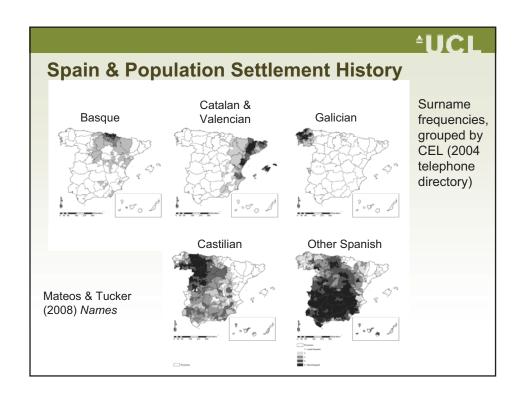




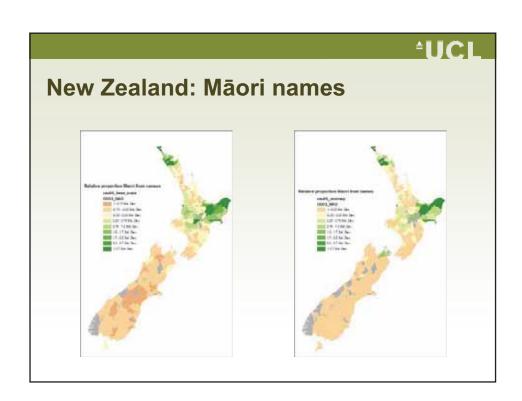


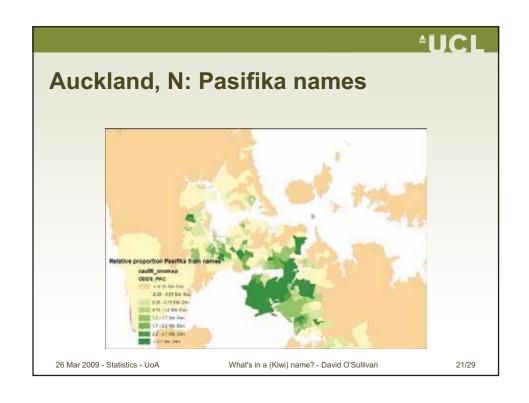


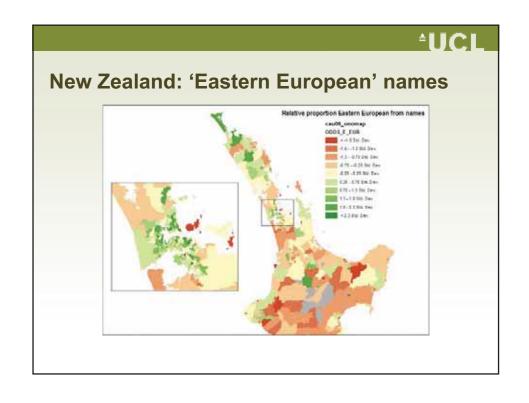




New Zealand: Scottish names Passenger arrivals at Por Chainers New Zealand: Scottish names PHILIP LAINS Breakie (Hod) (22) (sem.) Brown, James (23) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (23) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (23) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (23) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (23) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (24) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (24) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (24) (selako printer, ag., Andersons Bay; d. 27/11/1891); Brown, James (24) (seaver, famer, (46) (selako printer); d. (46) (selako printer); d. (46) (selako printer); d. (47) (selak





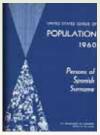


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Ethnicity Names & Networks



Research on names and identity





- Demography and epidemiology; subdivision of populations by ethnicity
 - US Census Hispanic names list (Passel and Word since 1950s)
 - Asian surnames in US (Lauderdale, 2004)
 - South Asian names in UK (Nam Pechan & SANGRA)
- Genetics; Population structure and geography, endogamy and gene mutations
- Economics; Name discrimination in labour, housing, and credit markets
- Geography and Sociology; cultural transmission, migration and spatial diffusion

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Decoding ethnicity from names

Names can potentially provide information about:

Aspect	Etimology/ Onomastics	Space-time Distribution
Surname &	Language	Geographic Origin
Forename	Religion	Migration flows
Forename	Gender	Age

- Review paper of name-based classifications of ethnicity;
 Mateos (2007) Population, Space and Place
- · Primarily public health applications
- Main groups: Hispanic, South Asian, Chinese, and Muslim

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Creating a name-based ethnicity classification

· Objective:

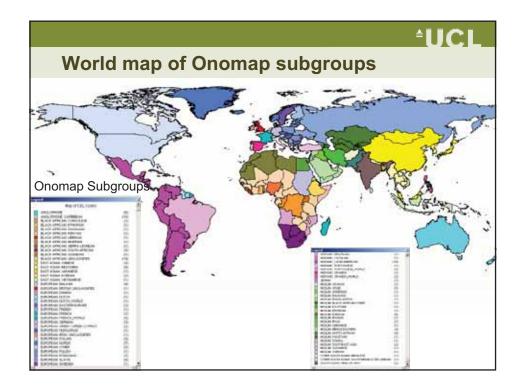
- To create a classification of forenames and surnames by fine ethnic groups, covering the whole of the population in 26 countries
- EU-21, North Am., AU & NZ, Japan, India, & Argentina

Data sources:

- UK Electoral Register 2001-2006 (46 million adults)
- Telephone directories (300 million subscribers)
- · Covering a total population of 1 billion people
- Individual level data (full name and address) for:
 - 8 million unique surnames
 - 5 million unique forenames

• Analysis:

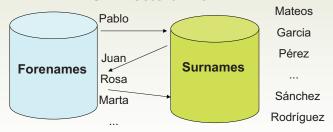
- Over 1.5 million unique names coded into "Onomap classification"
- 80 Onomap Subgroups and 14 Groups



Onomap classification

 Forename-Surname clustering (based on Hanks and Tucker, 2000)

UK Electoral Roll



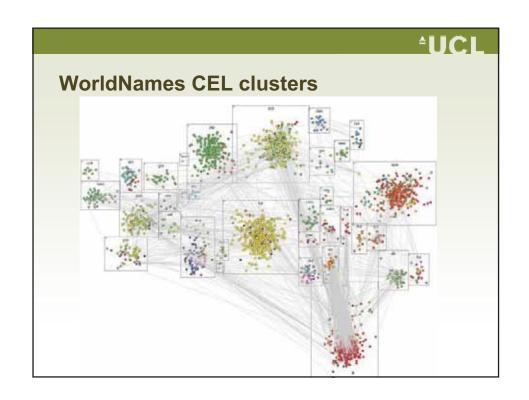
- Several iterations until self-contained cluster is exhausted
- Cluster assigned a cultural, ethnic & linguistic Onomap type
- Probability of ethnicity assigned to each name

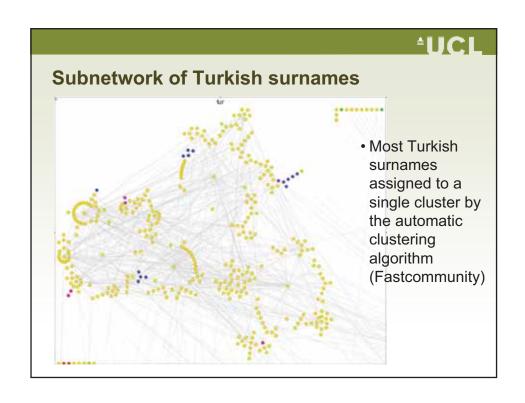
Mateos et al (2007) CASA Working Paper 116

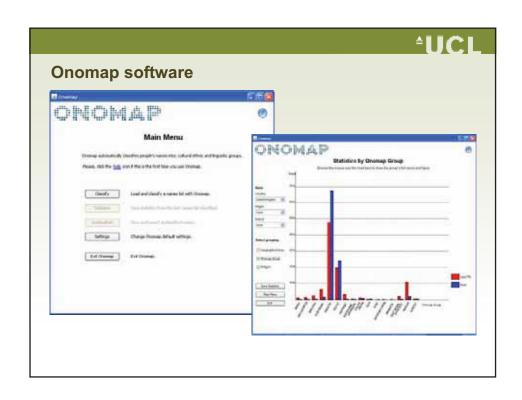
Naming networks • 5000 surnames in WorldNames • Nodes = unique surnames • Edges = common forenames between a pair of surnames • Weights = relative importance of the forename to each surname (% of pop) • Colours = clusters assigned by network clustering algorithms

Social networks parallel

- Granovetter (1973) The Strength of Weak Ties
- Weak ties play an essential role in the diffusion of information and innovation
- Cliques of highly related names are separated by bridges or weak links (sparse links)
- Key: identify bridges removing 'the weak ties' through filtering on edge weights
- Network clustering algorithms (Newman)
- Forthcoming paper with David O'Sullivan on naming network clustering, using Fastcommunity (Clauset, Newman and Moore, 2004)

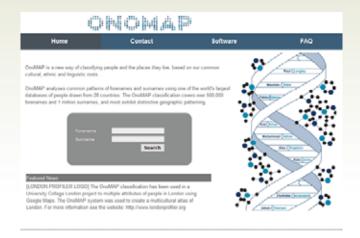






Onomap online

 Individual names can be searched on-line www.onomap.org



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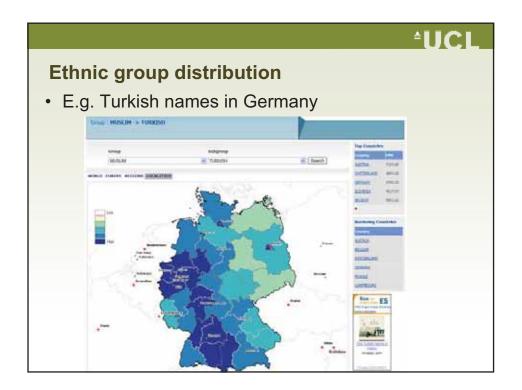
Evaluation of Onomap at the individual level

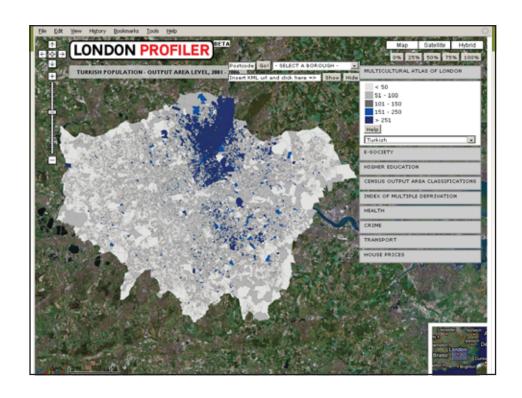
- Evaluated against individual's ethnicity (hospital admissions, birth register)
- On average: 80% accuracy
- Least indicated for:
 - Long established groups (i.e. Irish, Jewish)
 - Groups with non-unique names (i.e. Caribbeans)
- Highly indicated for all other groups

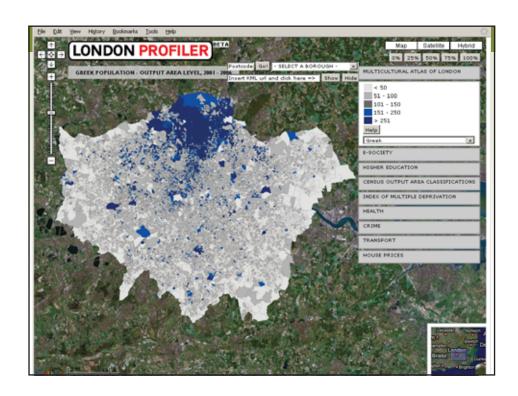
Mateos, Longley and Webber (2007) Lahka, Gorman and Mateos (forthcoming)

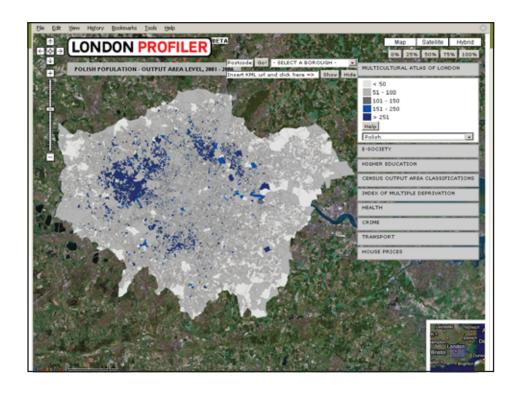
Current applications of Onomap

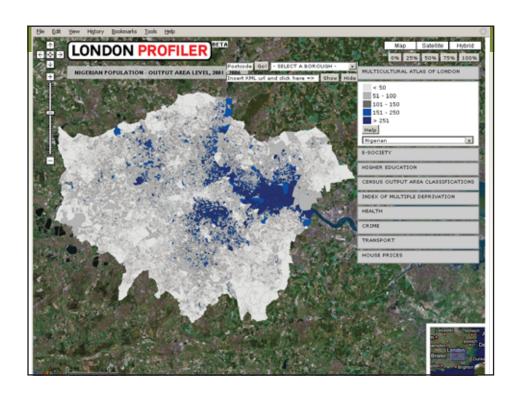
- Cancer research
 - LSHTM (5 m. patients)
- Public Health
 - PCTs (Camden, Islington, Southwark)
 - University of Edinburgh & GROS (Onomap evaluation)
 - University of Essex
- Political party representation
 - ANU, Australia; Princeton Univ.
- · Residential Segregation
 - Europe, US, UK

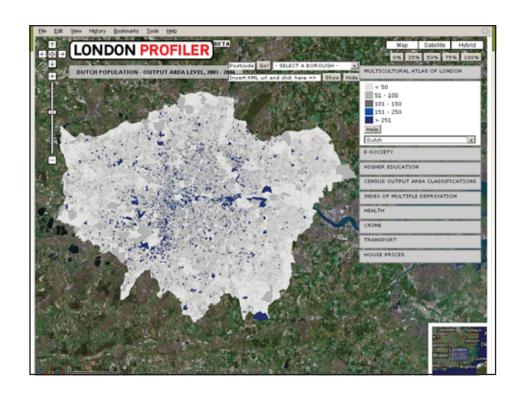


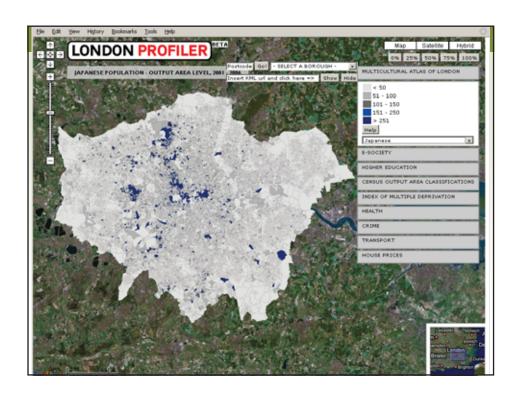






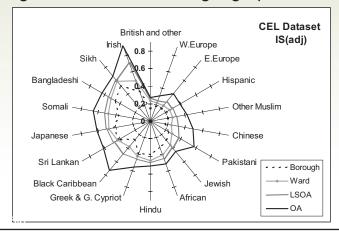


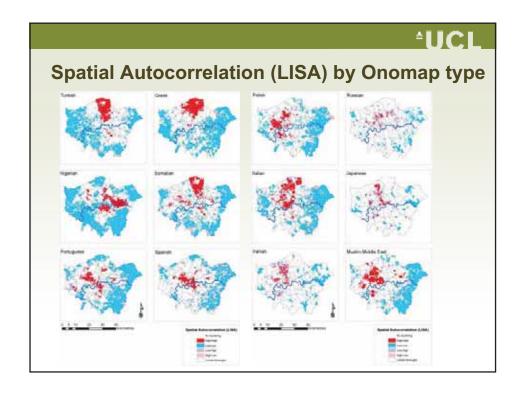




Index of Dissimilarity – Onomap

 Index of Dissimilarity in London, using Onomap categories at four different geographical scales





Lack of ethnicity data in Europe

- · Incompatible definitions and measurements:
 - Migrant / Native
 - Foreigner / National
 - Migrant ancestry
 - Race / Skin colour
 - Cultural / National identity
- Lack of ethnicity data collection, or poor quality and comparability
- · Lack of sufficient geographical granularity
- · Low frequency of update
- Need for complementary methodologies to study segregation
 - There is a fragmented, nation-specific research perspective on spatial segregation in Europe (Musterd, 2005)

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European cities segregation analysis

- Names from telephone directories coded by Onomap Subgroup
- · 31 largest cities selected
- Counts per Onomap Group and small area (postcodes) calculated within each city
- · Index of Dissimilarity calculated for each city

$$D = \frac{1}{2} \sum_{i=1}^{n} \left| \frac{x_i}{X} - \frac{t_i - x_i}{T - X} \right|$$
 (Duncan and Duncan, 1955)

xi = Population of group X in area i

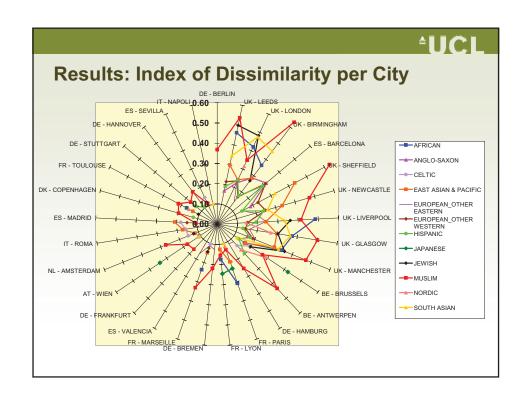
X = Total population of group X in the city

ti = Total population of area i

T = Total population of the city

Precautions taken

- · To prevent small groups driving results
- Analysis done at Onomap Group level (14 groups) not Onomap Subgroup (80)
- Presentation of results filtered, after ID was calculated:
 - Remove Onomap Groups with average population per area in a city < 15 individuals
 - Remove Onomap Groups with ID < 0.1



Conclusions: European Segregation

- · UK cities much more segregated than rest of countries
 - could be a result of finer small area geography
 - followed by cities in Belgium and France
- · Muslim group most segregated across Europe
 - followed by South Asian, and African
 - at larger distance by Japanese, Jewish & East Asian

Issues

- High variability between cities: population sizes, extent and number of areas
- Differential population coverage (tel. dir. & elec. roll)
- Onomap Group level may not reflect the full reality in each country (i.e. historic minority groups)
- Dissimilarity index on its own not ideal (Simpson, 2005; Massey & Denton, 1988)

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Conclusions: Onomap methodology

Advantages

- Facilitates ethnicity analysis using finer spatial, temporal, and nominal granularity
- Cost-efficient alternative when ethnicity data is missing/ low quality
- Ethnicity categories can be re-aggregated in different ways
- Probability scores; tailor classification to specific applications

Disadvantages

- Only reflects patrilineal heritage (problem of mixed ethnicity)
- Different histories of surname adoption, naming conventions & name change rules in each language and country
- Name normalisation decisions are required
- Publicly available registers of names have biases
- Not appropriate for reporting ethnicity at individual level
- Ethical considerations and privacy issues

Thanks for listening

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www.onomap.org www.publicprofiler.org/worldnames www.londonprofiler.org