

Situating social policy analysis: possibilities from quantitative and qualitative GIS

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Introduction

Geographic Information Systems (GIS) use maps and spatial analysis as a way of investigating real-world problems. Although traditionally associated with the geographical disciplines, it has since been adopted and used throughout the social sciences, including urban and environmental planning, public and community health, political and social studies, crime prevention and transport to name a few (Longley et al., 2015). GIS originally developed out of computerised cartography and it still uses maps and mapping as a way of structuring, managing, visualising and analysing data. A GIS stores a map as a series of separate map layers, with each layer containing data on a particular theme. A theme could relate to the population of an area, such as population density or percentage of children of school age; physical infrastructure, for instance the road network or public transport routes; or the environmental quality of an area, such as the location of parks and green space or the concentration of air pollution. By layering these themes one on top of the other within the GIS digital mapping, it is possible to see how they relate to each other across an area and to produce composite maps that can reveal spatial patterns and relationships. GIS contains tools that allow the data to be manipulated and analysed spatially and statistically so it is possible to determine, for example, how many people live within five minutes' walk of a bus stop or how many children in low-income households live in a particular school catchment area. This combination of mapping and spatial analysis makes GIS a powerful tool when researchers are investigating how social and economic problems vary across space, how spatial areas relate to one another, or how policy interventions may have different outcomes geographically. And as GIS can map and analyse data at a

1 variety of spatial scales it is possible to investigate problems at the very
2 local level through to the national and international level.

3 Traditionally, GIS has been a predominately static tool – it emphasises
4 variation across space rather than change over time. This has partly been
5 a reflection of data availability – socioeconomic data collection tends
6 be cross-sectional rather than longitudinal and in terms of mapping,
7 data are usually only available aggregated into predefined geographical
8 areas whose boundaries often change, hampering comparisons
9 between time periods. These areas are usually designed for a particular
10 function (for example wards for electing local councillors) and so often
11 their boundaries do not reflect the underlying structure of the local
12 population or the social process being investigated. This can result in
13 the geographic areas distorting or unduly influencing the display and
14 analysis of the populations and social processes of interest. This is called
15 the Modifiable Areal Unit Problem (MAUP) and it continues to be an
16 innate issue in social applications of GIS. The ecological fallacy (EF) is
17 related to MAUP and occurs when data aggregated to areas are used to
18 make inferences about individuals who live in those areas. So although
19 we may observe that areas which have high unemployment rates may
20 also have high crime rates we cannot say that it is the unemployed
21 that is the cause of crime, merely that the two have an association
22 at that geographical scale. Again, EF, like MAUP, is a reflection that
23 socioeconomic and demographic data geographically referenced to
24 individual persons and households has traditionally not been available
25 in GIS and rather is supplied at some level of geographic aggregation.

26 But GIS, socioeconomic data and geographical data are all evolving
27 and there is now increasing capabilities for dynamic mapping,
28 interactive mapping and animated visualisations (for example
29 Andrienko et al., 2016) allowing researchers to investigate how spatial
30 patterns and relationships change over time, and in a finer detail,
31 improving the analysis of the outcomes of policy interventions. These
32 and other issues are addressed and reflected upon throughout the
33 chapter. It will argue why GIS is a useful tool for spatial social policy
34 analysis, looking at how maps and mapping can change the way we
35 see and understand spatial relations between people and places. It
36 will discuss how recent innovations in qualitative GIS are opening up
37 the field to new academic and policy areas that benefit from a mixed
38 method approach to understanding social policy. It will provide many
39 examples from across broad policy themes that illustrate the advantages
40 of exploring the impact of social policy from a geographic perspective
41 within the context of maps, mapping and spatial analysis. It concludes
42 with a reflection and discussion of the issues raised in the chapter and

looks towards the future in terms of how recent innovations and trends could further the use of GIS in the field of social policy.

Why is GIS a useful tool for spatial social policy analysis?

The way in which we ‘see’ and ‘understand’ the spatial relationships that exist within the world has changed over the last 50 years (Davoudi and Strange, 2009) and influenced the evolution and use of GIS. GIS is utilised to overlay different datasets in order to not only represent information, as the discipline of cartography typically does, but also to query and analyse the relationship between those different layers of information. Some of the earliest uses of this more dynamic analysis of spatial data date back to the interpretation of health information collected as part of cholera outbreaks in Paris and London. Most famously, in 1854 the London physician John Snow mapped the locations of cholera victims and nearby water sources in Soho to identify a contaminated water pump where concentrations of victims lived. In doing so, he demonstrated the potential analytical capabilities of overlaying different data sources to identify spatial patterns that may not have traditionally been represented and analysed together on the same map (Snow, 1855; Orford, 2005) and he also provided a timely public health intervention by disabling the pump, preventing its further use.

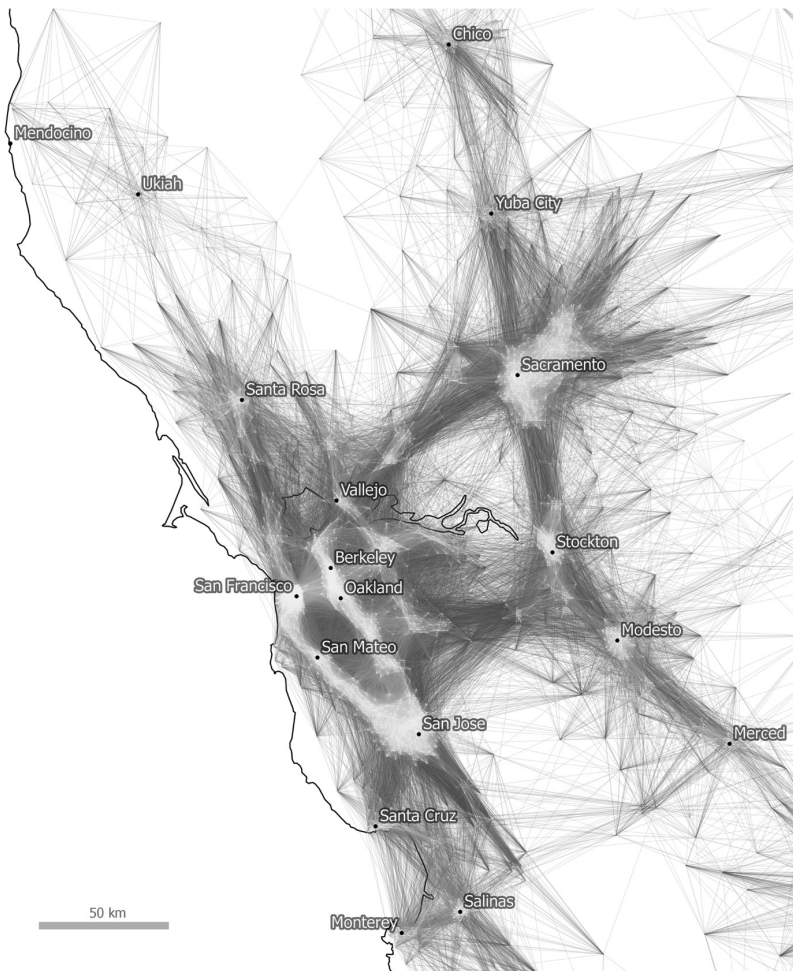
Through the layering of data in computer-assisted environments, modern GIS provides the ability to represent the spatial distribution of people, resources and information in a single, easy to understand map (Wong et al., 2015). These visualisations are useful means to distil large amounts of information into a single image that can be easily understood by a wide range of non-technical audiences, from policy experts, to politicians, to local citizens. Despite improvements in the analytical capabilities of GIS software, it continues to be used more for spatial *information visualisation* rather than as an *analytical tool* to support decision making (Gilfoyle and Wong, 1998; Vonk et al., 2005). This is often the result of a lack of technical expertise of individuals within the public and private sector working in the fields of social science, a point which is discussed later. From a social policy point of view, the layering, analysis and visualisation of data can provide an important means of identifying areas for policy intervention (van der Horst, 2007), allocating resources (Ashby and Longley, 2005), understanding long-term trends (Rebel, 2007) or programme evaluation (Fischer and Nijkamp, 1993). Further innovative uses of GIS beyond these mainstays provide a glimpse of how it might add particular value to

1 social policy analysis. However, GIS remains a strikingly niche method
2 across the social sciences and largely overlooked with social policy
3 research and practice. Yet, as discussed below, both quantitative and
4 qualitative GIS offer considerable potential for such policy making
5 and analysis, examples of which are highlighted in the next section

6 As with any methodological field, GIS techniques, thinking and
7 capabilities continue to evolve, opening up new opportunities for
8 rich spatial insights into social policy issues. Increasingly, for example,
9 more traditional quantitative GIS representations of containerised
10 Cartesian spaces are being challenged through attempts to characterise
11 the relationship between places by mapping flows of data – trade,
12 people, finance, information and so on – rather than displaying static
13 representations of place (Orford and Webb, 2017). Most commonly,
14 flow maps are utilised to map commuting patterns of workers, showing
15 how many people travel from home to work (Rae, 2016) in order to
16 understand the functional geographies of cities and regions –
17 [Figure 9.1](#). Flow mapping has been used to understand the spatial
18 extent of areas that cross local boundaries in order to demonstrate
19 the need for different governments to coordinate public policy,
20 particularly in relation to infrastructure provision. Yet, conversely, it
21 also holds potential for understanding the isolation of communities,
22 nodes of intense activity, and the interaction, or lack thereof, between
23 prosperous and more deprived areas.

24 In a similar vein, GIS is being used beyond its positivist origins to
25 support the creation of more qualitative diagrammatic or conceptual
26 maps that draw together a range of secondary data to construct a
27 single, more artistic image of an overall idea. Such a process is useful
28 for defining potential policy problems that need to be addressed or
29 as a baseline for a visioning process to engage different actors about
30 initiatives (Wong, 2006). The process brings together multiple layers
31 of spatial data about a place and then works to analyse and simplify
32 that information into more basic representations of space, such as by
33 identifying core problem areas, key strategic corridors, or relationships
34 between places. In 1989 French geographer Roger Brunet developed
35 a conceptual metaphor of Europe drawn from a range of spatial
36 socioeconomic data and historical information, identifying a 'Blue
37 Banana' (due to its shape and the colour used to map it) stretching from
38 North West England around France's north-eastern border to northern
39 Italy as a means of highlighting the lack of European socioeconomic
40 integration and inequality between European countries (RECLUS,
41 1989). This simplification of spatial data was not processed through
42 mathematical equations but rather was informed qualitatively by

Figure 9.1: Tract-to-tract commutes of 80km/50 miles or less in the Bay Area, California, USA



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individual experiences, value systems and professional norms which interacted to generate a new way of thinking built through social practices rather than positivist representations of space.

Furthermore, the last decade has seen various innovative attempts to use GIS for qualitative data management and analysis (for example Kwan and Knigge, 2006; Pavlovskaya, 2006 **[[please provide reference]]**; Elwood and Cope, 2009). By adding geo-references to place names within a document, transcript or social media content, for instance, it is possible to use the GIS to map the textual data and add

1 value and additional insights into the understanding and interpretation
2 of the text. Other types of data promoted by qualitative GIS include
3 photographs, video footage and audio clips that can add depth
4 and context to the statistical data traditionally associated with GIS.
5 Moreover, GPS **[[please give in full]]** trace data of the movements of
6 people or people's perceptions of the place in which they live collected
7 through cognitive mapping exercises may provide methodologies and
8 approaches that go beyond the static and Cartesian framework of
9 most GIS. Hence qualitative GIS lends itself to those aspects of social
10 policy research that, for example, emphasise the importance of lived
11 experiences when addressing social problems and the effectiveness
12 of policy interventions, or that seeks to identify differences between
13 objective knowledge provided by quantitative GIS data and people's
14 partial, subjective knowledge and experience of place.

16 **Case studies of indicative examples**

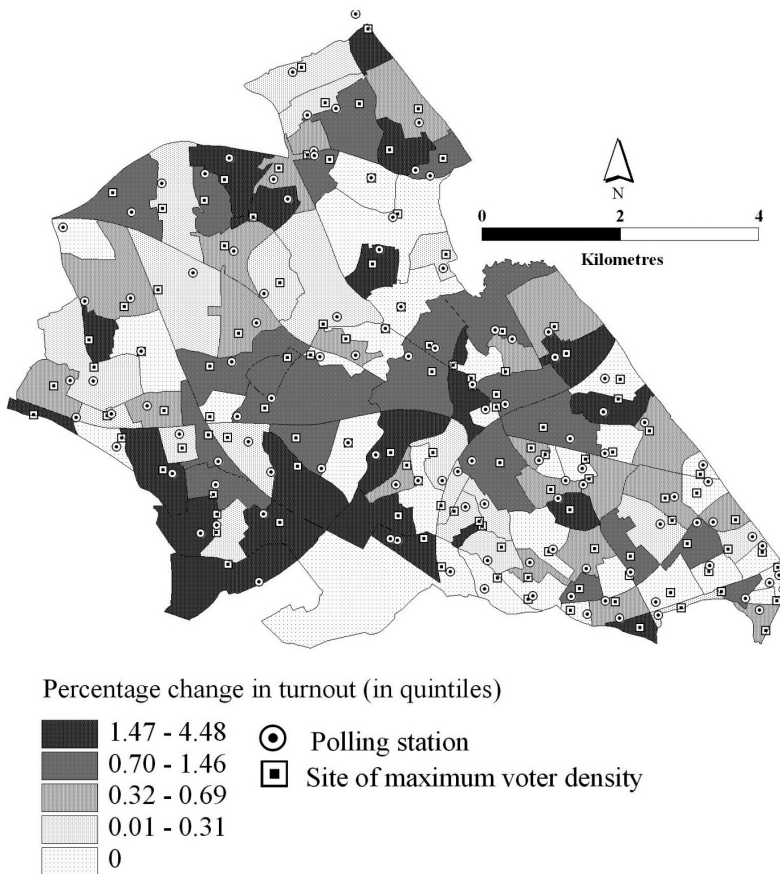
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18 The previous section has provided an overview of some of the major
19 ways in which quantitative and qualitative GIS are used and their
20 potential to enrich applied social and policy analysis. This section
21 moves on to provide a broad overview of different examples of social
22 policy research with reference to both quantitative and qualitative
23 GIS. The aim in doing so is to illustrate some of the specific insights
24 and opportunities that such GIS methods can bring to social policy
25 making and analysis.

26 A common use of GIS in social policy is to research people's ability
27 to access public amenities such as post offices, GP surgeries, libraries,
28 sports facilities, polling stations and good schools (for example Orford
29 et al., 2011; Singleton et al., 2011; Higgs et al., 2015, 2017; Langford
30 et al., 2016). This has important implications for the spatial equity
31 of access with people from poor backgrounds or living in poorer
32 neighbourhoods often finding it more difficult to access essential or
33 good-quality services. It is possible to use GIS to generate a wide
34 variety of accessibility measures from simple straight-line distances, to
35 road network and footpath distances, as well as travel time measures.
36 Increasingly, these metrics include the use of public transport, such
37 as buses and trains, to capture those households which do not have
38 access to private transport. More sophisticated accessibility measures
39 will model the supply as well as the demand for public amenities to
40 reflect competition, congestion and the allocation of scarce resources
41 (for example Langford et al., 2016). These measures have been used
42 as variables in statistical models that have quantified the effects of

accessibility by different groups of people to different types of amenities. For instance, Figure 9.2 illustrates how voter turnout to local elections in London could be improved by siting polling stations closer to where most voters live in terms of voter density.

Another example is to understand the spatial implications of school choice on the educational performance of children in the state school sector. In many Western countries including the UK, parents have a choice as to which state school their children may attend and this can lead to competition for the more popular schools and thus the application of admissions criteria to select students. These criteria usually include catchment areas and distance to school (Singleton et al.,

Figure 9.2: Percentage differences in the predicted probability of turnout at polling district level when re-siting polling stations at the maximum voter density locations for the 1998 local elections, London Borough of Brent, UK



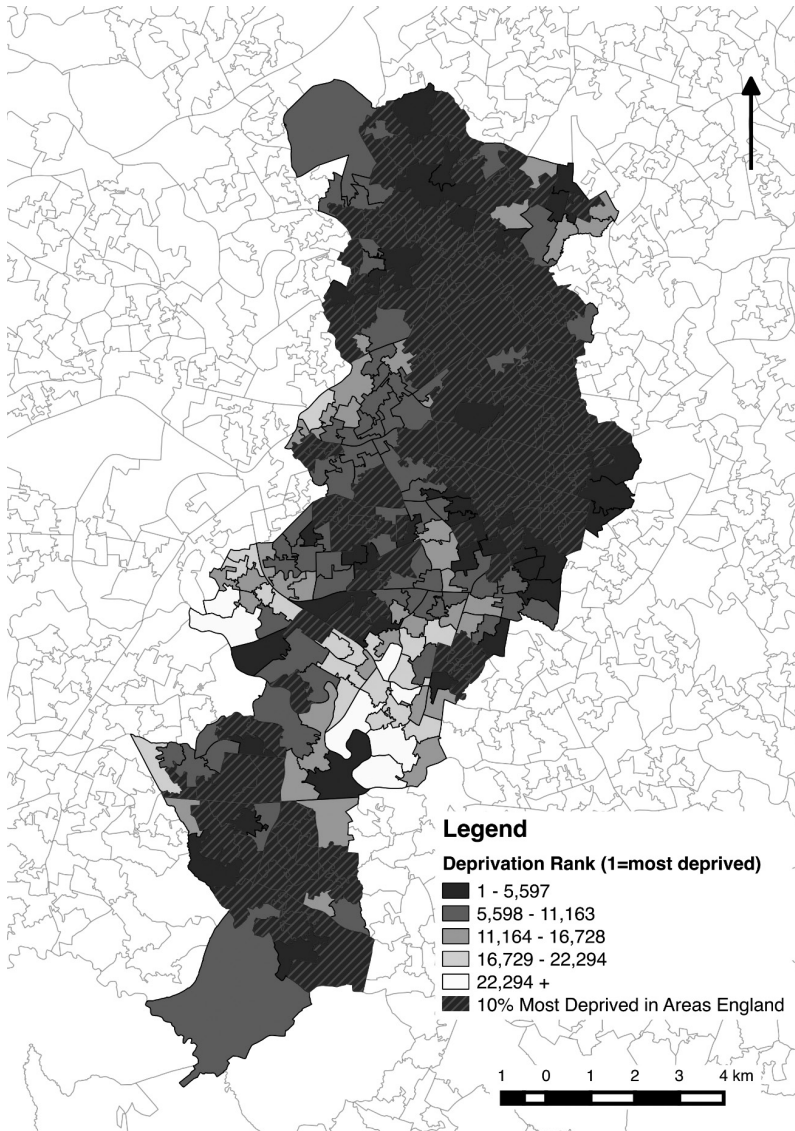
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1 2011) and this can lead to some parents moving close to good schools,
2 pushing up house prices and rents in those local areas (Cheshire
3 and Sheppard, 2004; Glen and Nellis, 2010; Orford, 2018). This
4 'selection by mortgage' (Harris et al., 2016) can lead to residential
5 sorting of neighbourhoods, reinforcing social class differences and, as
6 a consequence, exacerbating existing inequalities in the state education
7 system. GIS can also be used to model both catchment areas and
8 distance and relate this to educational performance (for example
9 Burgess and Briggs, 2010), income identifiers (such as children in the
10 receipt of free school meals – see Chapter **[[chapter 8?]]** and also
11 Hamnett and Butler, 2011) and house prices (for example Glen and
12 Nellis, 2010; Orford, 2018).

13 Accessibility measures have also been used in the construction of
14 the Indexes of Multiple Deprivation (IMD) for England, Scotland
15 and Wales. More broadly, an IMD has been utilised in a number
16 of countries to map and target a range of policy initiatives. This
17 approach divides deprivation into a range of domains that comprise
18 different socioeconomic indicators at different spatial scales. One set
19 of indicators relates to service deprivation measured by GIS analysis of
20 reasonable access to key services using some of the measures outlined
21 above. The domains are then merged and weighted depending on
22 perceived importance to develop a single deprivation ranking of all
23 small areas in a given location (Noble et al., 2006). When mapped,
24 geographical variation in social and economic circumstance between
25 areas becomes apparent and it is often used as a means of understanding
26 the spatial inequality present between places (Figure 9.3). In order
27 to reduce spatial inequality, the measure has been used as a means of
28 targeting funding to places in need as well as for area-based policy
29 initiatives, such as urban regeneration and improvements in service
30 provision (Deas et al., 2003). IMDs have seen particular use to map
31 health inequalities (Mackenzie et al., 1998; Saunders, 1998) and
32 understand factors related to issues such as mortality (Smith et al.,
33 2014), breastfeeding (Brown et al., 2010) and cardiovascular disease
34 (Ramsay et al., 2015).

35 GIS has also been increasingly utilised to develop geodemographic
36 typologies of places in order to better understand their spatial make-up.
37 Geodemographics are developed by drawing together large amounts
38 of spatial socioeconomic data to identify 'types' of similar areas and
39 classify them based on like characteristics (Vickers and Rees, 2011).
40 Identified areas are often then described based on their overriding
41 features, such as 'Aspiring and Affluent Cosmopolitans', 'Urban and
42 Professional Families' or 'Constrained Ageing City Dwellers'. Such an

Figure 9.3: Index of Multiple Deprivation Ranking for Manchester, UK at local level super output area level (2015)



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approach stems from the idea that ‘where you are says something about who you are’ (Harris et al., 2005: 2). Whilst this is to some extent inevitably a simplification of complex neighbourhood dynamics, such classifications have been widely used in policy and practice, including crime and retail analysis, urban policy and regeneration, through to marketing and environmental management (see Singleton

1 and Spielman, 2014). Recently, geodemographics have been extended
2 to explore not just the places that people live but also the ways they
3 travel to work. This approach links demographic data to commuting
4 flows, allowing researchers and policy makers to understand not only
5 the number of people commuting between places but also who those
6 people broadly are (ethnicity, occupation, gender, and so on.) and
7 what mode of transportation they use (Hincks et al., 2018). This data
8 has then been linked into an online interactive GIS system, providing
9 users the ability to visualise the flows between places dynamically.

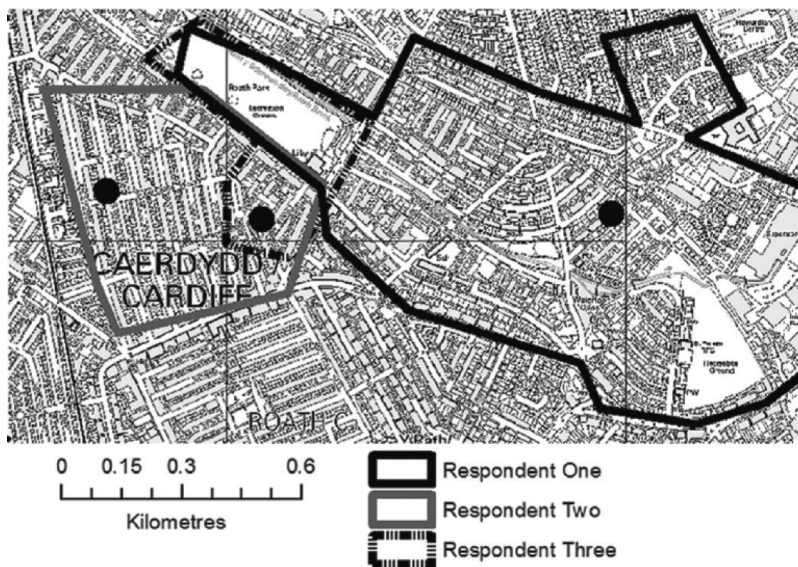
10 At a more technical level, GIS systems have also been developed to
11 help professional planners and other policy makers to make decisions.
12 These 'Planning Support Systems' (PSS) incorporate a wide range of
13 spatial data about places that they can overlay and manipulate to help
14 them make better informed decisions (Geertman and Stillwell, 2004).
15 Unlike traditional GIS, such systems are customised to the task of
16 decision making, incorporating a range of different components with
17 specifically developed tools, models and analytical capabilities unique
18 to the solving of problems. These systems help decision makers in two
19 main ways, first to identify the problem, its causes and what can be
20 done about it, and second to learn about other stakeholders' views by
21 facilitating the exchange of information between professional as well as
22 lay groups (Pelzer et al., 2014). Examples range from systems that allow
23 the impact of transportation interventions on land use to be modelled
24 (Arampatzis et al., 2004), to measuring and targeting sustainability
25 measures and the potential impact of policy changes (Graymore et al.,
26 2009). A key challenge with these systems stems from the wide range
27 of disciplinary clients, which, as they become more developed and
28 complex, must balance software capabilities with the levels of expertise
29 present among potential users.

30 In this respect, there are examples of innovative qualitative and
31 participatory GIS within social policy research and practice, reflecting
32 the importance of engagement and participation of community groups
33 and the public within planning, policy and research processes. There
34 has been a tradition of participatory mapping in local government
35 planning as a means of engaging local communities, particularly in the
36 consultation process of local plan and decision making. These tend
37 to use various 'mental mapping' exercises (such as sketch maps) as a
38 means of getting local people to elucidate their thoughts, perceptions,
39 opinions and local knowledge of an area being affected by a planning
40 decision (Cinderby and Forrester, 2005; Dennis, 2006). The spatial
41 representations created through these mapping exercises are digitised
42 and incorporated into the GIS for further analysis. An example of

participatory mapping is in the analysis of crime and neighbourhood safety. Orford and Leigh (2014) analysed over 700 maps based on residents' perceptions of the neighbourhoods in which they lived and worked in Cardiff, created as part of a wider research project on the role of neighbourhood intelligence in combating crime. The maps were generated by the residents during in-depth interviews using a mobile GIS interface which allowed them to draw neighbourhood boundaries onto a digital base map and annotate the map with insights and knowledge about their neighbourhood (Figure 9.4). This allowed maps of different crimes to be analysed with respect to resident-defined neighbourhoods and revealed the importance of local knowledge in explaining patterns and occurrences of crime at different times of the day and night. Other examples include the work by Pain and colleagues (2006), who found that crime hotspots identified by GIS crime data mapping did not match up with residents' experience of crime or their satisfaction with crime prevention measures, such as good street lighting, which has important implications for future policy interventions.

The adoption of GIS in local planning authorities saw a movement away from information gathering solely using participatory mapping approaches to Public Participation GIS (PPGIS), where local communities and citizens are involved in an iterative process before

Figure 9.4: Self-reported neighbourhoods with residential/work locations in Cardiff, UK



1 the final maps are produced (Kwan and Ding, 2008). Some of the
2 first usages of more community-based PPGIS in the UK were to
3 engage communities in local problem identification. For example, in
4 the 'Shaping Slaithwaite' project residents of the West Yorkshire village
5 were asked at a local fair to contribute to ideas about ways to improve
6 their village. As part of this, computers were provided and an online
7 interactive GIS system of the local area was created. Residents were
8 asked to interact with the online map and leave comments about the
9 different places, such as what areas should be developed or protected
10 or what a particular building means to them, which were then fed into
11 the local planning process (Kingston et al., 2000). At a larger scale, a
12 woodland regeneration in the Yorkshire Dales National Park Authority
13 also highlighted the early potential of PPGIS by allowing users to
14 interact with an online map to select relevant factors to consider for an
15 expansion in woodland tree planting and weighting their importance
16 to generate different possible scenarios and preferred options (Carver
17 et al., 2001).

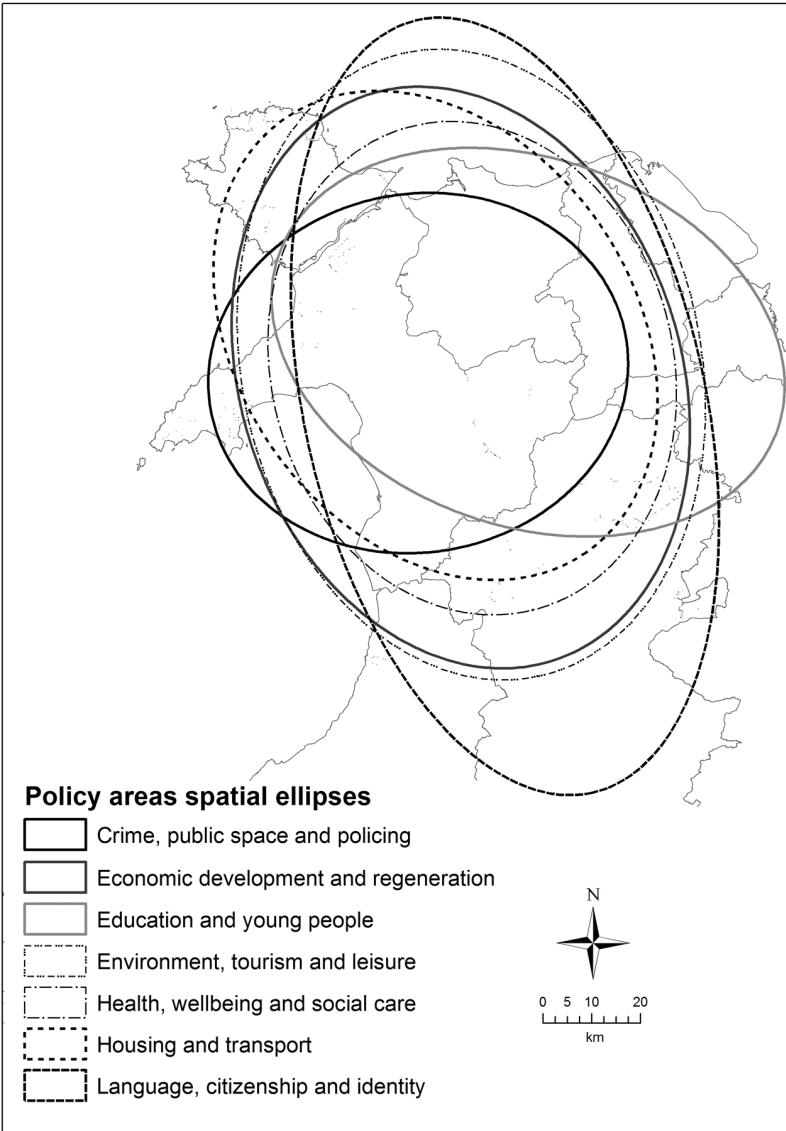
18 Despite their democratic, policy and analytical appeal, there is
19 however a potential danger with such approaches as projects may be
20 based on explanatory information and maps that support the arguments
21 of the planning authority rather than those of the communities
22 involved and does nothing to address the inherent power relations in
23 the decision-making process (Perkins, 2007). It is therefore necessary
24 to carefully consider how such systems are designed and for whom.
25 With the advent of Web 2.0 and neogeography (Haklay et al. **[[ok**
26 **changed to match reference]]**, 2008) citizens increasingly have the
27 opportunity to become more involved with the collection, analysis and
28 presentation of mapping data than in PPGIS in a field referred to as
29 GeoParticipation (Panek, 2016). Here new (and often free) geospatial
30 technologies such as GPS receivers on smart phones allow members
31 of the public to generate and analyse their own maps and upload them
32 onto online community sites to allow others to engage with them
33 without the constraints imposed by official organisations. Probably
34 the best example is OpenStreet Map (OSM) but others include
35 FixMyStreet, by the charity mySociety, which makes it easier for
36 people to report problems in their local community via a map-based
37 platform. Less formal is EmoMap (Ortag and Huang, 2011; Gartner,
38 2012) that allows local people to map their emotions about particular
39 places using a smart phone app and make them publicly available
40 online. This can be used to visualise the way people feel about different
41 parts of their local community which could form an additional layer
42 of information for planners and decision makers. GeoParticipation is

a call for a more humanised approach to mapping information and technology in an attempt to democratise the spatial decision-making process of planners and policy makers (PANEK, 2016). The results of these different types of participatory mapping are often the reflections of the social and cultural backgrounds of the communities and their understanding of spaces and spatial relations (CORBETT and RAMBALDI, 2009).

In a related field, geo-narrative is a GIS-based approach to narrative analysis based on oral histories, life histories and biographies (KWAN and DING, 2008). Here GIS is used to handle, visualise and analyse the chronology of people's experiences and the sequence of events. An example of this is research into the lives of 37 Muslim women in Columbus, Ohio, US in 2002 and the effects of the 9/11 attacks in terms of hostility and hate crimes (KWAN, 2008). The life paths of the 37 women were generated in the GIS based on a variety of multimedia data, including: survey diaries about activities and trips undertaken on designated days; oral histories through in-depth interviews based on how their lives had changed post-9/11 and their perception of safety and risk of where they lived; sketch maps of neighbourhoods including areas which they considered unsafe before and after 9/11; and photos and voice clips to contextualise their experiences. The life paths and associated geo-narratives of the women's lives revealed that space and time played a significant role in shaping the participants' post-9/11 experiences and that specific spatial and temporal experiences and events could be identified that were common to several of the women.

Finally, a slightly different example of qualitative GIS in social policy research comes from ORFORD and WEBB (2017) and their mapping of social policy areas identified from the daily working practices of public policy practitioners in Wales. Here practitioners were interviewed in depth about their daily working lives and were encouraged to talk about the places that were significant in their activities. These places were geo-referenced and then mapped using spatial ellipses (ALEXANDER et al., 2011) as a way of aggregating the individual places into regions. Different maps were produced for different policy areas and these were overlaid to compare and contrast how practitioners working in different fields had different activity spaces even though they may work for the same local authority and be based in the same location (Figure 9.5). The maps identified core and peripheral areas of working in the local authority which did not necessarily match up with the practitioners' official job demarcations and instead reflected historic ties with former administrative areas long since abolished and collaborations with new organisations. They emphasise the fuzziness in

1 **Figure 9.5:** Spatial ellipses of stakeholder interviews by policy area for North
2 Wales, UK



the spatial working practices of policy practitioners and how these are difficult to change even with the implementation of new geographic regimes.

The future of GIS in social policy and analysis

This final section looks towards the future of GIS in social policy and analysis and briefly discusses continuing challenges and some developments that may help address these issues. Traditionally there have been many barriers to the adoption of GIS in social research and policy fields (for example Göçmen and Ventura, 2010). These include the costs of hardware and software, the costs and availability of data, knowledge and awareness of what GIS can do within an application domain, and the training and skills to use the GIS. Over the years some of these barriers have been reduced or removed entirely, whilst others remain in place. GIS software now runs on a standard PC or laptop and the emergence of free and open-source software such as QGIS means that significant financial costs have been removed. Spatial data, such as digital boundary data, are increasingly available under Open Data licence agreements (including UK Ordnance Survey data), whilst many government statistical data sources have standardised geographical references associated with their data records and are available on open platforms (such as data.gov and data.gov.uk).

However, there are many types of data that social policy researchers may find useful, such as survey or administrative data records, where the geographical references may not be of good quality or non-existent and mapping of the data remains problematic (see Fry et al., 2017, and Bright et al., 2018, for recent commentaries on the issues of mapping alcohol licence records collected by local authorities and alcohol outlets available from OSM respectively). Added to this is the general rule of thumb that the more fine-grained the spatial data the less likely it is to be current and the fewer types of data that are available. The quantity, variety and currency of social policy data are far higher for large geographical areas such as government regions and local authorities/municipalities than for small areas such as neighbourhoods, wards and census tracts. This can impact the viability of GIS in social policy research if the objective is to monitor and evaluate the effect of policy interventions in a timely manner at small spatial scales such as within cities or neighbourhoods.

There can be complex privacy and ethical issues associated with mapping data relating to people of interest in social policy research, such as vulnerable children or people with mental health problems, which can either prevent the use of GIS or impede analysis. Maps can hide as well as reveal people and their relationships with places (Dorling and Fairbairn, 1997). Vulnerable or minority groups are either often hidden in the detail of the maps or missing altogether.

1 This is partly because these groups are missing from the data records
2 (for example homeless persons) or they appear in numbers so small
3 that they are redacted due to data disclosure issues or vanish when
4 aggregated into larger populations. Mapping can be a disclosive act
5 in itself – showing where a person or group of people live can often
6 identify them and therefore the publication of a map may be restricted
7 or the data obscured in some way.

8 Nevertheless, despite these technical and ethical challenges, arguably
9 the largest barrier to the uptake of GIS in social policy research
10 concerns the knowledge of what GIS can offer and the training and
11 skills to undertake the research. GIS education and training tends to
12 be focused within particular disciplines in universities (Harris et al.,
13 2014) and not necessarily those disciplines associated with social
14 policy research. Within government organisations, GIS tends to be a
15 function of particular teams (for instance, in Planning) as opposed to
16 being embedded throughout the organisation, and this is particularly
17 true of smaller organisations. The result is that the people with the GIS
18 knowledge and skills are not necessarily working within social policy
19 research teams and this limits the application of GIS in this domain.

20 There are several developments in GIS and the social policy data
21 landscape that could have positive impacts on the use of GIS in social
22 policy research. The first is the emergence of Big Data and linked
23 data especially around administrative datasets. Big Data not only refers
24 to the enormous size of some social datasets that now exist, or their
25 completeness in terms of population-level data, but also to the use of
26 machine learning to undertake predictive analysis of user behaviour
27 to find correlations between different socioeconomic processes in
28 different geographical environments. In social policy this has been
29 used for example to identify patterns in crime rates, disease prevention,
30 and understanding how different types of people move around areas
31 at different times of the day. Coupled to this is the new types of data
32 becoming available gathered by cheap and numerous devices such as
33 mobile devices (such as smart phones), wireless sensor networks (such
34 as for capturing movement of people) and CCTV cameras, as well as
35 social media outputs such as Twitter feeds, that can provide live or
36 near-live information on people and places. These data can have an
37 associated geographical reference enabling them to be mapped and
38 analysed within GIS and have the potential to inform social policy
39 research.

40 The second development is the Open Data movement which
41 encourages organisations and agencies that collect and host data,
42 including government, to make these more accessible to researchers.

In the UK, the Economic and Social Research Council (ESRC) have funded a series of Big Data Research Centres whose aim is to facilitate the promotion, access and use of data collected by government and other organisations by academic and policy researchers. Centres include the Administrative Data Research Centres in the four nations of the UK – which deal with data relating to health and social security records, benefits and tax records and crime and justice records – the Business and Local Government Data Research Centre and the Consumer Data Research Centre which allows access to government and commercial data. The centres provide facilities to link data from different datasets to individuals and provide safe settings in which to analyse the data. It is likely that the types of data made available via the Research Centres would be valuable for social policy research and may address some of the issues discussed earlier associated with ethics, privacy and disclosure as well as the currency and availability of data for small areas. Furthermore, accessing data geographically referenced to individuals and households may be one way to address MAUP as this allows data to be aggregated to bespoke areas that better reflect the underlying population and social processes being investigated.

The final development concerns the increasing pervasiveness of GIS technology and spatial data within social science and related research (partly reflecting the ‘spatial turn’ in the social sciences) and also within the policy and civil society spheres. As GIS, and in particular mapping technology, becomes more prevalent, social policy researchers and the groups they research may become more spatially literate and start to use maps and spatial analysis more in their research. A good example of this is the gradually increasing uptake of open source GIS software and the development of qualitative GIS which is slowly being adopted into new academic and policy areas that have previously viewed GIS as inaccessible, unfamiliar or even inappropriate. However, qualitative GIS is still emerging and it lacks the suite of analytical tools and processes available for traditional GIS applications. There remain challenges with geo-referencing qualitative data, particularly textual data where spatial references may be encoded using descriptions such as ‘close to where I live’ or ‘in the neighbouring town’ or ‘far away from here’ or use vernacular place names that may not exist in official gazetteers. Recent developments in natural language processing, fuzzy matching and querying, and crowd-sourced mapping products such as OSM are helping here though and GIS software packages are now better at handling qualitative data records.

To conclude, GIS and the socioeconomic data landscape are evolving, and increasingly the field has a lot to offer to social policy.

1 Innovations in qualitative GIS have allowed a more mixed-method
2 approach, better suited to social policy research with its emphasis on
3 understanding lived experiences and also evaluating the impact of
4 policy interventions. New mapping techniques that emphasise flows
5 of people, trade, ideas, and so on, allow maps to move away from the
6 limited view of containerised Cartesian space whilst dynamic and
7 interactive mapping challenges the static view of GIS and opens up
8 temporal and well as spatial analysis of policy outcomes. Increasing
9 access to micro-level and population-level data sources through open
10 data and Big Data initiatives can start to address some of the concerns
11 of sample size, data currency, MAUP and EF in relation to policy
12 analysis whilst the increasing pervasiveness of maps and mapping in the
13 social sciences and beyond means that geographical methods and ways
14 of thinking are becoming more salient in the policy arena.

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