Geospatial analysis of salmonellosis and its association with socioeconomic status in Texas

Anand Gourishankar

ABSTRACT

Objective The study's objective was to find the association between salmonellosis and socioeconomic status (SES) in hot spot areas and statewide counties.

Design A retrospective cohort study.

Setting The data were recorded regarding salmonellosis in 2017 from the Texas surveillance database. It included assessment of hot spot analysis and SES association with salmonellosis at the county level.

Participants Patients with salmonellosis of all age groups in Texas.

Results There were a total of 5113 salmonellosis from 254 counties with an unadjusted crude rate of 18 per 100,000 person-years. Seven SES risk factors in the hot spot counties were as follows: low values of the severe housing problem, unemployment, African American and high values of social association rate, fast food/full-service restaurant use, Hispanic and Hispanic senior low access-to-store (p<0.05). A 12% difference existed between local health departments in hot (25%) and cold spot (37%) counties (χ², 1, n=108)=0.5, p=0.81).

Statewide independent risk factors were severe housing problem (incidence rate ratio (IRR)=1.1; 95% CI: 1.05 to 1.14), social association rate (IRR=0.89; 95% CI: 0.87 to 0.92), college education (IRR=1.05; 95% CI: 1.04 to 1.07) and non-Hispanic senior local access-to-store (IRR=1.98; 95% CI: 1.26 to 3.11). The severe housing problem predicted zero occurrences of infection in a county (OR=0.51; 95% CI: 0.28 to 0.95).

Conclusions Disparity exists in salmonellosis and SES. Attention to unmet needs will decrease salmonellosis. Severe housing problem is a notable risk.

BACKGROUND

Salmonellosis is one of the leading foodborne illnesses in the USA. Salmonellosis (nontyphoidal salmonellosis (NTS)) is defined as a clinical illness that results from infection by serotypes of Salmonella other than typhi or paratyphi (A/B/C) that causes typhoid fever.1

In the USA, among 31 major pathogens of foodborne illness, non-typhoidal Salmonella spp was the leading cause of hospitalisation (35%) and death (28%).1 Salmonellosis accounted for 31% of all foodborne illnesses.2

Salmonellosis risk varies by serotypes, for example, Salmonella enteritidis infection rates were higher in lower socioeconomic status (SES), minority and high average number of children per family.3-4 SES was known to influence the risk of exposure to a specific pathogen.5-6 A systematic review of four foodborne-illness pathogens, including Salmonella, showed that higher SES, higher education, unemployment, poverty, urban area and lower deprivation were associated with higher incidence.5 Geographic Information System (GIS) and spatial analysis have expanded into the science of spatial epidemiology.7-9 Area-based SES such as levels at the county, census tract or block group analysis may provide information about transmission and the opportunity to prevent infections.10 The population-based ecological studies are limited in number and not generalisable, as described further here. Using forward sortation area-specific SES and disease rate, a study in Canada found that salmonellosis was associated with age <5 years, non-white race and poverty.11 A study that used block group-level found that decreasing years of education was associated with reduced Salmonella infection.12

Key points

Question

► This study explored the geographic variation of salmonellosis and its association with socioeconomic status in Texas.

Finding

► Salmonellosis clustered in the north-central part of Texas.

► High cluster counties had disparities in ethnicity, unemployment and low access to the store.

► Severe housing problem, social association rate, college education and low access to the store were associated with salmonellosis globally across Texas.

Meaning

► This study highlights the importance of location for patients with salmonellosis, emphasising location-based issues such as socioeconomic status, access to local health departments and various social structures.
A previous census tract-level study in Connecticut found an association between high SES and NTS in persons ≥5 years of age and some serotypes. A county-area level non-spatial analysis for SES found unemployment, black, Hispanic and Latino ethnicities were associated with NTS. Currently, there is no spatial analysis of area-based (county-level) SES and NTS in Texas.

A county-level assessment will provide new insights for community interventions to reduce the infection rate and community health service disparities. The objective of the current study was to analyse the relationship between non-typhoidal *Salmonella* infections and county-level SES in the hot spot areas and across all counties within the state of Texas.

### METHODS

The estimated population for each Texas county was derived from the US Bureau of Census. County-level SES was derived from The County Health Rankings & Roadmaps programme and Food Environment Atlas data (table 1). For this study, the county-level NTS (2017) across Texas was retrieved from the Texas Department of Health (www.dshs.texas.gov). The study focused on the yearly incidence, which included the cumulative outbreak strains of *Salmonella* in Texas. All the data were joined to administrative boundary shapefile of counties obtained from the TIGER/Line database (www.census.gov) using ArcGIS Pro 2.5 (ESRI Inc, Redlands, California, USA). There was no missing data. The data were derived from the following resources available in the public domain:

1. Texas Department of Health; www.dshs.texas.gov
2. RWJ Foundation Program. County Health Rankings and Roadmaps 2020; https://wwwcountyhealthrankings.org/.
3. Food environment atlas data; https://wwwersusdagov/data-products/food-environment-atlas.

#### Statistical analysis

Analysis of the total population and NTS counts included descriptive statistics. The mean (SD) and median (IQR) were calculated for continuous SES variables. Data distributions of SES variables determined the inferential statistics (Student’s t-test or Wilcoxon test). The $\chi^2$ statistics compared the proportions of local health departments of Texas. The spatial empirical Bayes (EB) smoothing method reduced the random variation of infection rates. In addition, it accounted for unstable incident rates in areas with small populations (second-order queen contiguity weights using R software V.3.6, Vienna, Austria). This study does not contain human participant involvement, and no private health information was used. A p value 0.05 or lower was considered statistically significant.

### Table 1 Explanatory variables: definitions and sources

| Variable name         | Description                                                                 |
|-----------------------|-----------------------------------------------------------------------------|
| High School Education | Percentage of the ninth-grade cohort that graduates from high school in 4 years (2014–2015) |
| College Education     | Percentage of the population ages 25–44 with some postsecondary education, such as enrollment in vocational/technical schools, junior colleges or 4-year colleges (2011–2015) |
| Unemployed            | Percentage of the county’s civilian labour force, age 16 and older, that is unemployed but seeking work (2015) |
| Median Household Income | The income ($) where half of the households in a county earn more and half of the households earn less (2015) |
| Social Association (rate) | Number of membership associations per 10 000 population (2014) |
| Severe Housing Problems | Percentage of households with at least 1 of 4 housing problems: overcrowding, high housing costs, lack of kitchen facilities or plumbing issues (2009–2013) |
| Uninsured             | Percentage of population under age 65 without health insurance (2014)        |
| Rural                 | Percentage of rural counties                                                |
| FFR-AE*               | Percentage of fast-food restaurant availability and expenditures per 1000 population (2014) |
| FSR-AE*               | Percentage of full-service restaurant availability and expenditures per 1000 population (2014) |
| Low access to store*  | Percentage of low access to store for seniors among white, black, Hispanic, non-Hispanic Asian (2015) |
| Asian                 | Percentage of population that is Asian (2015)                               |
| White                 | Percentage of the population that is non-Hispanic White (2015)               |
| Hispanic              | Percentage of the population that is Hispanic (2015)                         |
| African American      | Percentage of the population that is non-Hispanic African American (2015)     |

Source: https://www.countyhealthrankings.org

*Except https://www.ers.usda.gov/data-products/food-environment-atlas.

FFR-AE, per cent of fast-food restaurants availability and expenditures per 1000 population; FSR-AE, per cent of full-service restaurants availability and expenditures per 1000 population.
Spatial analysis
A choropleth map was created with the Texas county base map and Jenks optimisation classification method in ArcGIS Pro. Choropleth will show a large-scale variation of salmonellosis incidence rate per 100 000 population (normalised) in each county. Moran’s index represented global clustering (spatial autocorrelation) of counties with high or low salmonellosis across Texas. Moran’s index measures proximity by inverse distance weights (IDWs)—an index that combines two measures of attribute similarity and location proximity into a single index.11 With continuous data, spatial conceptualisation, IDWs, every feature is a neighbour to each other feature is appropriate. The data had 254 observations, and hence a ‘Threshold distance’ was calculated. With incremental spatial autocorrelation, the distance band of 200 000 metres had the highest z-score. Moran’s index is considered statistically significant with a high z-score and p≤0.05. Next, local clustering of the disease investigates the spatial variations and spatial associations. Therefore, optimised hot spot analysis (extension of Getis-Ord Gi* statistics) identifies hot and cold spots by comparing individual high or low values (NTS counts) with surrounding other high or low values, respectively. The optimised hot spot analysis automatically selects an appropriate scale and adjusts results for multiple testing and spatial dependence. A large positive z-score that is statistically significant (p≤0.05) detects hot spots. Similarly, cold spots have a largely negative and statistically significant z-score. The choropleth map can display the hot/cold spot z-scores with confidence intervals in red and blue colours.

Modelling approach
Multiple linear regression was associated with multicollinearity and residual autocorrelation. Also, the NTS counts were skewed. The dependent variable was the number of salmonellosis cases in Texas county in 2017. The overdispersion and 32 counties with excessive zero infections had limited Poisson regression model fit. Therefore, we applied the zero-inflation negative binomial regression model (Zinb). The county-level population at risk obtained from the census was the ‘offset’ in the model. Forward regression was carried out using p value to include variables until the final model. The χ² test on the log-likelihood difference showed that the final model fitted the data significantly better than the null model.

Vuong test showed that the final model using Zinb performed better than the negative binomial model. The Zinb model fit was further confirmed by the absence of clustering in the model residuals in ArcGIS Pro (no spatial autocorrelation of residuals). The univariable screening included the Zinb regression of the 16 variables (table 1).

RESULTS
In 2017, there were 5113 cases in 254 Texas counties. The average unadjusted county’s crude incidence rate (IR) of salmonellosis in Texas was 18 per 100 000 person-years. The median IR (per 100 000) was 20 (IQR: 11–36).

Fourteen per cent (35/254) of counties had zero salmonellosis. The annual county population for the year 2017 ranged from 80 to 4 629 700 (median=18 934; IQR: 6 975–52 020). The choropleth map illustrated the county-level comparison of crude IR and EB smoothed rate. Despite EB smoothing of unstable rates, visual inspection of both maps showed (large scale) almost similar variations in Texas (figure 1).

Geographic clustering of the IR was high in central Texas (red) compared with low rates (blue) in east Texas (figure 2A). Hot spot corresponds to public health regions 1 and 2 (figure 2B). The global spatial clustering measure, Moran’s Index was 0.1 (p<0.001), which explains that the salmonellosis cluster pattern (hot spot) was unlikely due to random chance (table 2).

When comparing hot and cold spot counties, the following SES indicators were significant (p<0.05): low access-to-store in senior Hispanic, Hispanic and African American ethnicities. Hispanics were about 28% compared with 3% of African Americans in hot spot counties. On the contrary, the proportions of white were higher than other races in both hot and cold spot counties without statistical significance. The ‘social association rate’ (high means healthy) was high in the hot spot, and the ‘severe housing problem’ (low means healthy) was high in the cold spot. The study found high proportions of fast-food and full-service restaurant availability and

Figure 1  Salmonellosis choropleth map showing crude rate and EB rate. EB, empirical Bayes.
If the percentage of ‘severe housing problem’ increased by one unit, there were 51% odds that a county would have ‘zero’ salmonellosis. Thus, the higher the housing problem, the more unlikely the county had zero infection rate (zero-inflation model, table 3).

**DISCUSSION**

Salmonellosis of Texas in 2017 showed disparities in regional clustering of cases, public health regions and SES indicators at the county area level. Optimised hot spot and global Moran’s index method identified 21 counties in the north-central part of Texas with a high IR. The higher incidence of cases in the central part of Texas can be attributed to differences in the availability of local health departments and SES indicators. The central Texas (Public Health Regions 1 and 2/3) presented high disease clusters (statistically significant hot spots, figure 2A). Although no statistical significance, a gap of 12% in the presence of local health departments between cold and hot spot regions is meaningful. Expansion of health services and expenditure may reduce Salmonella incidence.17 18 In bivariate analyses, there was a statistically significant differences between seven SES indicators in the hot and cold spot counties. Based on this study, a combination of four SES indicators of ‘severe housing problem’ (%), social association rate, college education (%) and ‘low access to store’ in non-Hispanic Asian seniors (%) could explain the variability in the occurrence of salmonellosis in Texas.

At county-level GIS analysis, this study is the first one to demonstrate hot spot and SES association regarding salmonellosis (broad, not species specific). Previous studies had attempted a similar effort to analyse SES at different area-based levels; GIS analysis at the census tract-level showed the association of age and high SES in certain Salmonella species.13 The block-level analysis had demonstrated the association of decreasing years of education with a decrease in Salmonella infection.12 Both these studies showed a general trend without hot spot analysis. Although the current study did not analyse various serotypes, it is unclear why specific serotypes affect high SES. Varga et al found S. enteritidis area-level hot spot clustering with SES indicators (income, visible minority, number of children/family).11 The hot spot analysis method differed between the current study and Vargas et al. Understanding these differences includes the interplay of physical, biological, behavioural, cultural, health services utilisation, SES and environmental factors. The current study results were consistent with some of the results from these studies. However, they could expand on analysing various county-level serotypes in the future.

A higher percentage of a college education was associated with high clusters of salmonellosis in this study. Most, but not all, studies have shown associations between high educational attainment and infection.12 14 19 It is postulated that higher education increases awareness of food safety labels.20 By contrast, the decreased incidence was reported in low SES (low education and income) due to better hand hygiene practices, less risky food and better expenditures in hot spot counties. Median income was not different between the groups, but high unemployment was in cold spot counties. There were no differences in those with high-school or college education. Furthermore, there was a disproportionate distribution of local and city health departments among Texas’ 11 public health regions. Higher values of salmonellosis (hot spot) was noted in the regions with a lower number of health departments (hot spot=25%, cold spot=37%, $\chi^2 (1, n=108)=0.5$, p=0.81, figure 2B).

In the global multivariable model, among the 16 independent variables, the final model included only four variables. These variables were ‘severe housing problem’ (%), ‘social association rate,’ college education (%) and percentage of ‘low access-to-stores in non-Hispanic Asian seniors’ (table 3). Estimates (incidence rate ratio or OR) resulted from one unit increase in the final variable, provided the other variables were held constant in the model. As such, for ‘low-access-to-store of non-Hispanic Asian seniors,’ ‘severe housing problem’ and college education, the rate of salmonellosis would be expected to increase by a factor of 1.98, 1.1 and 1.05, respectively. By contrast, if a county had an increase in ‘social association rate’ by one unit, the salmonellosis rate would be expected to decrease by a factor of 0.89. The ‘severe housing problem’ predicted zero occurrences of salmonellosis.
Other explanations for higher education and infection associations are greater access to healthcare, health-seeking behaviour, pet ownership and eating fresh produce, raw or uncooked food. Unemployment has a protective effect on salmonellosis. This factor by itself or concurrent with a lower education level can explain this observation. An Italian study by Borgnolo et al found higher non-typhi Salmonella infection rates in children whose fathers were either unemployed or working in non-blue-collar jobs. Thus, SES indicators, economic status (income) and higher educational attainment are intertwined, manifest a differential effect of SES in salmonellosis. By contrast, this study did not show a salmonellosis association with median household income consistent with previous research. Hence, the current study underscores the several explanations that interplay between education and economic status.

In a study by Lay et al African Americans had a higher incidence of salmonellosis, whereas this study found a higher percentage of African Americans in non-hot spot counties. In Younus et al they found no association with ethnicity and salmonellosis, but our study found that the Hispanics were higher in hot spot counties, consistent with Arshad et al. Ethnicity may be a function of individual risk factors and pathogen-specific (ecological effect of serotypes and SES). Although disparities exist, it is an unclear association between foodborne pathogens and ethnicities. The disparity in salmonellosis among ethnicities can arise from gentrification and housing segregation. Thus, exposing the

| Variable statistics | Hot spot county* (n=46) | Cold spot county* (n=37) | P value |
|---------------------|-------------------------|--------------------------|---------|
| Population          | 7403 (3470–15 300)      | 36640 (14670, 65845)     | <0.001  |
| Incidence rate      | 26 (17–49)              | 12 (9–19)                | <0.001  |
| Low access to store—Hispanic senior (%) | 6.8 (3–13) | 1.7 (0.7–3.5) | <0.001 |
| FFR-AE (%)          | 0.7 (0.5–1)             | 0.5 (0.4–0.6)            | 0.008   |
| FSR-AE (%)          | 0.8 (0.5–0.9)           | 0.5 (0.4–0.6)            | <0.001  |
| College education (%) | 49.5 (41–59)          | 46 (43–54)               | 0.38    |
| Social association rate (per 10,000) | 16.7 (13–22) | 11.8 (10–14) | <0.001 |
| Severe housing problem (%) | 11.5 (10–14) | 14 (13–16) | <0.001 |
| Unemployed (%)      | 4.3 (3.7–4.7)           | 5.3 (5–6.5)              | <0.001  |
| African American (%) | 3 (1–5)                | 14 (8–18)                | <0.001  |
| Household medium income ($) | 43290 (40500–48750) | 43000 (40770–46700) | 0.4 |

*Estimate and statistics—median (IQR) and p value from Wilcoxon test. FFR-AE, per cent of fast-food restaurants availability and expenditures per 1000 population; FSR-AE, per cent of full-service restaurants availability and expenditures per 1000 population.

| Explanatory variable | Estimate | SE | IRR | 95% CI | P value* |
|----------------------|----------|----|-----|--------|---------|
| Low access to store—non-Hispanic Asian seniors (%) | 0.6839 | 0.2313 | 1.98 | 1.26 to 3.11 | 0.003 |
| Social association (per 10 000) | −0.1122 | 0.0151 | 0.89 | 0.87 to 0.92 | <0.001 |
| Severe housing problem (%) | 0.091 | 0.0216 | 1.1 | 1.05 to 1.14 | <0.001 |
| College education (%) | 0.052 | 0.0081 | 1.05 | 1.04 to 1.07 | <0.001 |
| Severe housing problem (%) | −0.6661 | 0.3114 | OR=0.51 | 0.28 to 0.95 | 0.03 |

IRR, incidence rate ratio; SES, socioeconomic status.
population segment to any number of high-risk SES indicators noted in this study. Although the Hispanic ethnicity emerged significant, demographic factors, behavioural or cultural, and other individual risk factors may affect these associations.

The other highlight of this study was that the full-service restaurant utilisation among seniors in the county was associated with salmonellosis (seniors are a known, high-risk age group). Darcey and Quinlan and Signs et al have found differences in SES with health code violations and food safety in retail, respectively. Appling et al reported the risk of Salmonella infection and violations in the restaurant. Although the differences between hot-cold counties on restaurant use in this study were small but statistically significant, further studies can strengthen this study’s findings. Furthermore, low SES communities are more likely to visit fast-food restaurants. Fast-food and full-service restaurant availability and expenditures can be associated with economic disadvantages such as poverty, unemployment or low educational attainment. Thus, despite food safety measures by agencies and food education, SES indicators are the significant determinants for salmonellosis. The ‘low access to stores for seniors’ had divergent results for Hispanic compared with non-Hispanic. There may be a bias due to a higher percentage of Hispanics in the hotspot counties. The ‘social association rate’ is a powerful predictor of health status (positive perception and health behaviour). Although the ‘social association’ is a ‘rate’, limited by self-reporting of local entities, it measures vital health-related memberships, such as fitness centres, sports organisations, religious organisations, civic and business organisations. In addition, social networking and community improvement (social capital) support the belief that if individuals are not isolated and have strong social networks, they make healthy choices.

Based on this study, ‘severe housing problem’ (a measurement of the percentage of lack of kitchen or plumbing facilities, overcrowding and severely cost-burdened) demonstrated substantial influence in salmonellosis. It is the only indicator that predicted the ‘zero’ occurrence of salmonellosis infection. Although it appears protective, it underscores the magnitude and strength of ‘adverse societal problems’. Severe housing problem also reflects reduced food access, poverty, positive food storage and likely under-reporting and less access to healthcare. Adequate housing (a proxy of high SES) prevents harmful exposures and provides a sense of safety, contributing to health. The current study supports allocating resources and services to home environment assessments, indoor pest management, grants for community development, housing and inclusionary zoning and housing policies. With growing awareness of geomedicine in primary practice, physicians can expand on the knowledge of where the patients live. Therefore, one may explore their location-based access to local health resources and social, economic and environmental conditions.

SES is challenging as it lacks a single metric and SES index. Also, measurements vary among studies for a single variable. However, Jouve et al had described the inherent issue of a complex interaction between SES and the outcome of interest, a function of differential exposure and differential vulnerability. The analysis of Salmonella as a homogenous group may underestimate the association between various serotypes and SES. Under-reporting (decreased case ascertainment) due to passive surveillance of salmonellosis reduces the true incidence. Hence, the established associations will need cautious interpretation. Finally, ecological analyses do not assess confounding, and ‘ecological fallacy’ is inevitable. Although multiple regression addresses confounders, the final model is susceptible to mis-specification. The study’s strength includes group-level analysis accounting for both individual-level and community-level SES, rigorous hot spot analysis, no missing data and modelling at the local and global levels of SES. Data included in the county-level analysis can miss individual-level variation, whereas it provide information for directing policy and resources to the community.

CONCLUSIONS
The regional disparity in salmonellosis demands better research, improved capacity and an effort for surveillance to identify actual infection rates and the allocation of resources. However, the weight of the evidence suggests improving SES indicators and access to health services can reduce salmonellosis in Texas and across the USA.

Twitter Anand Gourishankar @a_gourishankar
Acknowledgements I thank the following individuals, Swati Jain, PhD, and Himanshu Batra, PhD, for their expertise and assistance in writing, technical editing and proofreading the manuscript.
Contributors AG conceptualised, analysed and wrote the entire manuscript.
Funding The author has not declared a specific grant for this research from any funding agency in the public, commercial or not-for-profit sectors.
Map disclaimer The depiction of boundaries on this map does not imply the expression of any opinion whatsoever on the part of BMJ (or any member of its group) concerning the legal status of any country, territory, jurisdiction or area or of its authorities. This map is provided without any warranty of any kind, either express or implied.
Competing interests None declared.
Patient consent for publication Not applicable.
Provenance and peer review Not commissioned; externally peer reviewed.
Data availability statement No data are available.
Open access This is an open access article distributed in accordance with the Creative Commons Attribution Non Commercial (CC BY-NC 4.0) license, which permits others to distribute, remix, adapt, build upon this work non-commercially, and license their derivative works on different terms, provided the original work is properly cited, appropriate credit is given, any changes made indicated, and the use is non-commercial. See: http://creativecommons.org/licenses/by-nc/4.0/.

REFERENCES
1 Scallan E, Hoekstra RM, Angulo FJ, et al. Foodborne illness acquired in the United States—major pathogens. Emerg Infect Dis 2011;17:7–15.
2 Mead PS, Slutsker L, Dietz V, et al. Food-related illness and death in the United States. Emerg Infect Dis 1999;5:607–25.
3 Graziani C, Luzzi I, Owczarek S, et al. Salmonella enterica serovar Napoli infection in Italy from 2000 to 2013: spatial and spatio-temporal analysis of cases distribution and the effect of human and animal density on the risk of infection. PLoS One 2015;10:e014219.

4 Varga C, PearI DL, McEwen SA, et al. Spatial-temporal epidemiology of human Salmonella enteritidis infections with major phage types (PTS 1, 4, 5B, 8, 13, and 13A) in Ontario, Canada, 2008-2009. BMC Public Health 2015;15:1247.

5 Newman KL, Leon JS, Reichsbaa PA, et al. The impact of socioeconomic status on foodborne illness in high-income countries: a systematic review. Epidemiol Infect 2015;143:2473–85.

6 Madler JL, Clophger P, Libby T, et al. Relationship between census Tract-Level poverty and domestically acquired Salmonella incidence: analysis of foodborne disease active surveillance network data, 2010-2016. J Infect Dis 2020;222:1405–12.

7 Tatem AJ. Innovation to impact in spatial epidemiology. BMC Med 2018;16:209.

8 Kirby RS, Delmelle E, Eberth JM. Advances in spatial epidemiology and geographic information systems. Ann Epidemiol 2017;27:1–9.

9 Lovett DA, Poots AJ, Clements JTC, et al. Using geographical information systems and cartograms as a health service quality improvement tool. Spat Spatotemporal Epidemiol 2014;10:67–74.

10 Krieger J, Higgins DL. Housing and health: time again for public health action. Am J Public Health 2002;92:758–68.

11 Varga C, PearI DL, McEwen SA, et al. Evaluating area-level spatial clustering of Salmonella Enteritidis infections and their socioeconomic determinants in the greater Toronto area, Ontario, Canada (2008-2009): a retrospective population-based ecological study. BMC Public Health 2013;13:1078.

12 Younus M, Hartwick E, Siddiqi AA, et al. The role of neighborhood level socioeconomic characteristics in Salmonella infections in Michigan (1997-2007): assessment using geographic information system data. Geospat Health 2016;11:428.

13 Whitney BM, Mainero C, Humes E, et al. Socioeconomic Status and Foodborne Pathogens in Connecticut, USA, 2000-2011(1). Emerg Infect Dis 2015;21:1617–24.

14 Chang M, Groseclose SL, Zaidi AA, et al. An ecological analysis of sociodemographic factors associated with the incidence of salmonellosis, shigellosis, and E. coli 0157:H7 infections in US counties. Epidemiol Infect 2009;137:81–20.

15 Kang SY, Cramb SM, White NM, et al. Making the most of spatial information in health: a tutorial in Bayesian disease mapping for areal data. Geospat Health 2016;11:428.

16 Marshall RJ. Mapping disease and mortality rates using empirical Bayes estimators. J R Stat Soc Ser C Appl Stat 1991;40:283–94.

17 Dunn JR. Housing and health inequalities: review and prospects for research. House Stud 2000;15:341–66.

18 Erwin PC, Greene SB, Mays GP, et al. The association of changes in local health department resources with changes in state-level health outcomes. Am J Public Health 2011;101:609–15.

19 Simonsen J, Frisch M, Ethelberg S. Socioeconomic risk factors for bacterial gastrointestinal infections. Epidemiology 2008;19:282–90.

20 Yang S, Angulo FJ, Attekurse SF. Evaluation of safe food-handling instructions on raw meat and poultry products. J Food Prot 2000;63:1321–5.

21 Patil SR, Cates S, Morales R. Consumer food safety knowledge, practices, and demographic differences: findings from a meta-analysis. J Food Prot 2005;68:1884–94.

22 Li-Cohen AE, Bruhn CM. Safety of consumer handling of fresh produce from the time of purchase to the plate: a comprehensive consumer survey. J Food Prot 2002;65:1287–96.

23 Nesbitt A, Majowicz S, Finley R, et al. High-risk food consumption and food safety practices in a Canadian community. J Food Prot 2009;72:2575–86.

24 Painter JA, Hoekstra RM, Ayers T, et al. Attribution of foodborne illnesses, hospitalizations, and deaths to food commodities by using outbreak data, United States, 1998-2008. Emerg Infect Dis 2013;19:157–63.

25 Lynch MF, Tauxe RV, Hedberg CW. The growing burden of foodborne outbreaks due to contaminated fresh produce: risks and opportunities. Epidemiol Infect 2009;137:307–15.

26 Borgnolo G, Barbone F, Scomparinca G, et al. A case-control study of Salmonella gastrointestinal infection in Italian children. Acta Paediatr 1996;85:804–8.

27 Lay J, Varma J, Zansky S, et al. Racial and ethnic Disparities in foodborne illness, 2000. Chicago, IL: Infectious Disease Society of America, 2000.

28 Anshad MM, Wilkins MJ, Downes FP, et al. A registry-based study on the association between human salmonellosis and routinely collected parameters in Michigan, 1995-2001. Foodborne Pathog Dis 2004;7:16–25.

29 Quinlan JJ. Foodborne illness incidence rates and food safety risks for populations of low socioeconomic status and minority race/ethnicity: a review of the literature. Int J Environ Res Public Health 2013;10:3634–52.

30 Darcey VL, Quinlan JJ. Use of geographic information systems technology to track critical health code violations in retail facilities available to populations of different socioeconomic status and demographics. J Food Prot 2011;74:1524–30.

31 Signs RJ, Darcey VL, Carney TA, et al. Retail food safety risks for populations of different races, ethnicities, and income levels. J Food Prot 2011;74:1577–83.

32 Appling KS, Lee P, Hedberg CW. Understanding the relation between establishment food safety management and risk factor violations cited during routine Inspections. J Food Prot 2018;81:1936–40.

33 Powell LM, Slater S, Mirtcheva D, et al. Food store availability and neighborhood characteristics in the United States. Prev Med 2007;44:189–95.

34 Fielding JE, Simon PA. Food deserts or food swamps?: comment on “Fast food restaurants and food stores”. Arch Intern Med 2011;171:1171–2.

35 House JS, Landis KR, Umberson D. Social relationships and health. Science 1988;241:540–5.

36 RWJ Foundation Program. County health rankings and Roadmaps, 2020. Available: https://www.countyhealthrankings.org/}

37 Shaw M. Housing and public health. Annu Rev Public Health 2004;25:397–418.

38 Andersen MJ, Skinner A, Williamson AB, et al. Housing conditions associated with recurrent gastrointestinal infection in urban Aboriginal children in NSW, Australia: findings from search. Aust N Z J Public Health 2018;42:247–53.

39 Brave man PDM, Egerter S, Sadegh-Novari T. Housing and health: exploring the social determinants of health. Robert wood Johnson Foundation, 2011. Available: www.buildhealthylplaces.org

40 Jouve J, Aagaard-Hansen J, Aidara-Kane A. Food safety: equity and social determinants. In: Equity, social determinants and public health programmes, 2010: 95–114.