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Multilevel examination of diabetes in modernising China: what elements of urbanisation are most associated with diabetes?

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Abstract

Aims/hypothesis The purpose of this study was to examine the association between urbanisation-related factors and diabetes prevalence in China.

Methods Anthropometry, fasting blood glucose (FBG) and community-level data were collected for 7,741 adults (18–90 years) across 217 communities and nine provinces in the 2009 China Health and Nutrition Survey to examine diabetes (FBG ≥ 7.0 mmol/l or doctor diagnosis). Sex-stratified multilevel models, clustered at the community and province

levels and controlling for individual-level age and household income were used to examine the association between diabetes and: (1) a multicomponent urbanisation measure reflecting overall modernisation and (2) 12 separate components of urbanisation (e.g., population density, employment, markets, infrastructure and social factors).

Results Prevalent diabetes was higher in more-urbanised (men 12%; women 9%) vs less-urbanised (men 6%; women 5%) areas. In sex-stratified multilevel models adjusting for residential community and province, age and household income, there was a twofold higher diabetes prevalence in urban vs rural areas (men OR 2.02, 95% CI 1.47, 2.78; women, OR 1.94, 95% CI 1.35, 2.79). All urbanisation components were positively associated with diabetes, with variation across components (e.g. men, economic and income diversity, OR 1.42, 95% CI 1.20, 1.66; women, transportation infrastructure, OR 1.18, 95% CI 1.06, 1.32). Community-level variation in diabetes was comparatively greater for women (intraclass correlation [ICC] 0.03–0.05) vs men (ICC ≤ 0.01); province-level variation was greater for men (men 0.03–0.04; women 0.02).

Conclusions/interpretation Diabetes prevention and treatment efforts are needed particularly in urbanised areas of China. Community economic factors, modern markets, communications and transportation infrastructure might present opportunities for such efforts.

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Keywords China · Modernisation · Multilevel analysis ·
Nutrition transition · Type 2 diabetes

Abbreviations

CHNS China Health and Nutrition Survey
FBG Fasting blood glucose
ICC Intraclass correlation coefficient

Introduction

Over the past decade, the prevalence of type 2 diabetes in China has more than doubled from approximately 3% in 1994 to 7–10% in 2008 [1, 2], which is of concern as China is home to more than 1.3 billion people and comprises one-fifth of the world's population. Rapid urbanisation, economic growth and economic and healthcare reforms have been implicated in the increased diabetes prevalence [3–6]. Urbanisation may influence diabetes prevalence through increased time engaged in sedentary lifestyle behaviours and greater consumption of animal products, high-fat foods and highly processed foods [7, 8]. Yet, little is known about which specific elements of urbanisation are most associated with diabetes.

Most of the published literature relies on dichotomous contrasts between rural and urban areas. However, in China there is complexity and heterogeneity between and within urban and rural communities [9]. For example, in China's uneven economic transition, some rural areas became larger than small cities, while other urban areas lost services and physical infrastructure with development [9]. Provincial and regional differences in services and infrastructure due to decentralised healthcare and economic systems are exacerbated because of the urbanisation process [5, 6]. Research in China suggests that a dichotomous classification of urban and rural areas inadequately captures this heterogeneity, whereas a multidimensional measure that encompasses healthcare, economic systems, services and infrastructure adequately characterises variation across the spectrum of urbanicity [9]. While it is important to understand the association between overall urbanisation and diabetes prevalence, understanding the elements of urbanisation most associated with diabetes can inform community-level diabetes prevention and treatment efforts.

In this paper, we capitalise upon unique population-based, community- and individual-level data from the China Health and Nutrition Survey (CHNS) [10] to examine the association between urbanisation and diabetes using a validated, multidimensional urbanisation index [9]. This urbanisation index includes a total score as well as 12 component measures (reflecting a diverse range of factors such as healthcare, economic systems, services and infrastructure) that improve with urbanisation. In addition, we use multilevel modelling to account for the inherent interdependence between community-level factors and individual health outcomes among members of the same community or province [11]. We hypothesise that: (1) total urbanisation is positively associated with diabetes prevalence and (2) each of the 12 components of urbanisation is separately and differentially associated with diabetes prevalence.

Methods

CHNS The CHNS collected health data in 228 communities in nine diverse provinces (Guangxi, Guizhou, Heilongjiang, Henan, Hubei, Hunan, Jiangsu, Liaoning and Shandong) throughout China from 1989 to 2009 in eight rounds of surveys. The 2009 survey included fasting blood collection for the first time. Using multistage, random-cluster sampling, counties in the nine provinces were stratified by income and weighted sampling was used to randomly select four counties in each province. Villages and small towns within counties and urban and suburban neighbourhoods within cities were selected randomly into primary sampling units defined politically and geographically based on State Statistical Office definitions [12]. The surveyed provinces represent 56% of the Chinese population. Survey procedures have been described elsewhere [10]. The study was approved by the Institutional Review Board at the University of North Carolina at Chapel Hill, the China–Japan Friendship Hospital, Ministry of Health and the Institute of Nutrition and Food Safety, China Centers for Disease Control. Individuals gave informed consent for participation.

Analysis sample The 2009 examination surveyed a total of 8,597 adult respondents aged 18 or older; 402 were not fasting before blood withdrawal and 58 were pregnant, resulting in a total of 8,137 individuals with fasting blood. Of these, 221 were missing anthropometry data, 44 were missing laboratory results and 130 were missing household income data, resulting in 7,742 with anthropometry and clinical examination information. One person was excluded from the analysis because he was the only person in his community, resulting in an analytical sample of 7,741. There were no statistically significant differences in the total 2009 sample vs the analytical sample in sex, urbanisation or BMI, although the individuals in the analytical sample had a higher age and a larger household income.

Dependent variables After an overnight fast, blood was collected by venipuncture (12 ml). Whole blood was immediately centrifuged and the serum was tested for glucose. Glucose was measured with a Hitachi 7600 analyser using a glucose oxidase phenol 4-aminoantipyrine peroxidase kit (GOD-PAP; Randox, Crumlin, UK). Self-report questionnaires were used to elicit information regarding medical history and current medication use. Diabetes was defined according to the WHO 2006 guidelines as having a fasting blood glucose (FBG) measurement ≥ 7.0 mmol/l ($n=364$) or answering 'yes' to the question 'has a doctor ever told you that you suffer from diabetes?' ($n=229$) [13].

Main exposure: urbanisation index and its components Urbanisation was measured using an urbanisation index based on

household- and community-level data representing 12 features of the community environment. Components related to healthcare, economic systems, services and infrastructure were chosen on conceptual and empirical grounds that the 12 components defined and distinguished levels of urbanicity (Table 1). The multicomponent scale is particularly relevant in areas such as China where rapid urbanisation has occurred with spatial and temporal variation in the types of changes occurring across regions. Jones-Smith et al [9]

created the multicomponent urbanisation index for China using established scaling procedures from the psychometric literature to measure urban features on a continuum. Each urbanisation component was comprised of several variables. Scoring for each component was based upon data distribution to maintain adequate spread through the range of possible values and to set each component's median value at half of the total possible points. Points were allocated on the basis of: presence (or number) of infrastructure or facilities

Table 1 Description of the urbanisation index components^a

Component	Description
Population density	Total population divided by community area; higher population density indicates greater urbanicity
Social services	Availability of medical insurance of the following types: commercial; free and women and children Availability and proximity of childcare centres for children <3 years old Greater access to medical insurance and childcare amenities indicates greater urbanicity
Health infrastructure	Availability, type, proximity, and quality of a private, city or district hospital within the community or, in lieu of those facilities, a private clinic or township hospital Presence of a pharmacy Greater access to larger and better health facilities as well as access to a pharmacy indicates greater urbanicity
Modern markets	The number of supermarkets, cafes, internet cafes, restaurants, mobile eateries, fast-food restaurants and ice-cream parlours in the community A greater number of these western-influenced businesses indicates greater urbanicity
Traditional markets	Presence and operating hours for nine types of markets (grains, oil, meat, vegetables, fish, bean curd, milk, fabric and fuel) in or in a nearby community The presence of these markets within the community and operating hours, including service more days per week, indicates greater urbanicity
Transportation infrastructure	Presence and higher number of paved vs gravel or dirt roads as well as bus and/or train stations in the community indicates greater urbanicity
Communications	Percentage of households with a television, computer or cell phone Presence of a cinema, newspaper and telephone service in the community A higher percentage of households and communities with these amenities indicates greater urbanicity
Housing infrastructure	Percentage of households within the community with indoor tap water, flushing toilets and gas stoves Average number of days per week electricity is available to the community A higher percentage of households within the community with these amenities and a greater number of days per week of electricity service indicates greater urbanicity
Sanitation	Percentage of households with treated water and with no excreta present outside the home Having treated water and less excreta present indicates greater urbanicity
Economic activity	Ordinary wage for male workers and percentage of population engaged in nonagricultural work Higher ordinary wages for male workers and smaller percentage of population in agricultural work indicates greater urbanicity
Education	Average educational attainment of adults >21 years old in the community Higher mean educational attainment indicates higher urbanicity
Education and income diversity	Variation in mean community educational attainment and household income for adults Greater variation in educational attainment and household income indicates greater urbanicity

^aAs described in Jones-Smith et al (2011) [9]. Points were assigned based on type, quality, and distance of features listed above, with more points given to more urbanised characteristics within each component. Scoring for each component was based upon data distribution, with each component's median value set to represent half of the total possible points while maintaining adequate spread through the range of possible values. All components were scaled so that each component had a possible range of 0–10. Given no a priori reason to weight one component more heavily than the others, each component was equally weighted and added together to form the urbanisation index, which ranged from 0 to 120 points. Variables described as percentages from households were derived from the CHNS household surveys in the communities. All other variables were derived from the CHNS community survey

within communities; the percentage of households within communities with specific amenities or the average proportion of individuals (or households) at specific education (or household income) levels. Once points were aggregated, each component was rescaled so that the highest possible value a community could attain was ten points. Scoring of the overall index was based on a maximum of ten points for each of 12 equally weighted components.

The selection of variables and the scoring strategy were examined for content validity [9]. In addition, Jones-Smith et al [9] assessed unidimensionality and multidimensionality using exploratory factor analysis and assessed test–retest reliability ($\alpha=0.85–0.89$ across all study years) and temporal stability ($r=0.90–0.94$ between study years). The authors assessed criterion-related validity ($\kappa=69–77\%$ across study years) using official administratively based government classification of urban/rural and construct validity using regression analysis, finding a highly statistically significant association between the urbanisation index and household per capita income. Findings from this evaluation suggest that the multicomponent measure performs better than the dichotomous administrative classification of urban and rural in models predicting health outcomes [9].

In the first stage of analysis, we were primarily interested in whether the overall multicomponent measure of urbanisation was associated with diabetes prevalence. Given that the 12 components comprising the urbanisation index interact synergistically to reflect urbanisation, a continuous measure does not adequately capture overall urbanisation. Thus, we categorised total urbanisation (range 30.4–106.6) into tertiles representing low (<59.0), medium (59.0–82.2) and high (≥ 82.3) levels. In the second stage of analysis, we examined whether diabetes prevalence varied across the 12 urbanisation components. The components were scored separately; one point represents a higher level of urbanisation for each component. To facilitate comparison across the components, we present ORs comparing high (75th percentile) vs low (25th percentile) levels of each of the 12 distinct and equally weighted urbanisation index components (Table 1).

Individual- and community-level factors Age and sex were self-reported. Income was reported at the household level. Height was measured without shoes to the nearest 0.2 cm using a portable SECA stadiometer and weight was measured without shoes and in light clothing to the nearest 0.1 kg on a calibrated beam scale. Overweight and obesity were reported according to WHO classifications (25–30 kg/m² and ≥ 30 kg/m², respectively) [14] and the Chinese classifications (24–28 kg/m² and ≥ 28 kg/m², respectively) [15]. Both sets of criteria are presented: the WHO classification for comparison across published studies and the Chinese classification for an internal reference.

Statistical analysis All analyses were conducted in Stata 12 (Stata, College Station, TX, USA). In descriptive analyses, we examined individual-level characteristics, total urbanisation level and each of the 12 urbanisation components across the nine CHNS provinces. We tested sex-specific differences in individual-level characteristics across low, medium and high urbanisation and across the nine CHNS provinces by ANOVA for continuous outcomes and χ^2 tests for categorical outcomes. In addition, we tested differences in medians (due to the skewed distributions of the urbanisation index components) using the K-sample equality of medians test. The significance level for all descriptive statistics was set at $p<0.05$.

Given the high prevalence of undiagnosed diabetes in China [2], we examined the prevalence of undiagnosed and diagnosed diabetes by province and urbanisation levels. Using the question ‘Has a doctor ever told you that you suffer from diabetes?’, a ‘yes’ response was coded as diagnosed diabetes and a ‘no’ response plus an FBG ≥ 7.0 mmol/l was coded as undiagnosed diabetes. The two mutually exclusive categories together represent total diabetes used for the central analyses.

To examine diabetes prevalence across urbanisation levels, we used multilevel logistic regression models, with individual-level age and household income as fixed effects and with the community and province of residence as random effects. Given differences across regions of China in lifestyle, climate and economic factors, we assessed whether northern (Lianoning, Heilongjiang, Shandong and Henan) and southern (Jiangsu, Jubei, Hunan, Guangxi and Guizho) region confounded the association between urbanisation and diabetes prevalence. Geographic region did not meet the 10% change-in-estimate criterion of confounding, so it was not included in the final models. However, variation due to geographic region was captured by the province random effect, which accounted for clustering in lifestyle, climate and economic factors across China.

We did not include BMI in the final models on the basis that it is a causal intermediate on the pathway from urbanisation to diabetes and, as such, could potentially bias results [16]. However, in a secondary analysis we assessed the confounding by BMI of the association between urbanisation and diabetes. Findings indicated that BMI did not meet the 10% change-in-estimate criterion of confounding, thus BMI was not included in the final models (see electronic supplementary material [ESM] Table 1).

Likelihood ratio tests showed a statistically significant sex difference in the association between the education and income diversity urbanisation component with diabetes prevalence ($p<0.05$), thus all models were stratified by sex. When age was categorised as <50 or ≥ 50 years, there was no evidence for effect modification by age in the association between urbanisation and diabetes prevalence (men

$\chi^2=5.11$, $p=0.08$; women $\chi^2=0.64$, $p=0.73$). Tests for the interaction between age and the 12 separate urbanisation index components yielded similar results ($p>0.05$), showing no empirical basis for effect modification by age, thus analyses were not age-stratified.

First, we examined prevalence of diabetes across low, medium and high urbanisation to assess whether levels of the overall multicomponent urbanisation measure was significantly associated with diabetes (statistical significance set at $p<0.05$). Second, using the individual component measures as mean-centred continuous variables, we examined whether each of the 12 components of urbanisation was separately associated with diabetes. Given collinearity between the 12 urbanisation index components (correlation ranging from 0.2 to 0.7) and interest in assessing the associations across urbanisation components, we constructed 12 separate models per sex for each of the 12 urbanisation index components, resulting in a total of 24 models. We used the Bonferroni correction for multiple comparisons with a type 1 error rate of 0.05 and 24 models (p value set at 0.002). Half of the components were linearly related to diabetes prevalence on the logit scale; thus, all components were retained as linear variables for consistency and comparability across models. The components are scored independently and vary in their distribution across provinces and in the total sample. To facilitate comparison across the components and for comparability with the total urbanisation measure, we present ORs with 95% CIs comparing the 75th (high) vs the 25th (low) percentile for each of the 12 urbanisation components. Variance and SE of the random effects are presented to characterise community- and province-level variation in diabetes prevalence. We calculated the intraclass correlation coefficients (ICCs) at the community and province levels to examine the relative contribution of province, community and individual levels to diabetes prevalence.

Sensitivity analysis First, we replicated our analyses comparing two alternative diabetes diagnostic criteria ([1] $\text{HbA}_{1c} \geq 6.5\%$ [48 mmol/mol] and [2] a combination of $\text{HbA}_{1c} \geq 6.5\%$ [48 mmol/mol] and $\text{FBG} \geq 7.0$ mmol/l) with the diabetes diagnostic criterion used in the primary analyses ($\text{FBG} \geq 7.0$ mmol/l) (see ESM Tables 2 and 3).

Second, because diabetes prevalence across levels of the urbanisation index has not been compared with a traditional urban/rural dichotomy, we examined differences in diabetes according to the administratively defined urban/rural classification from the CHNS.

Results

Diabetes prevalence differed across low, medium and high areas of urbanisation, with highest prevalence in highly

urbanised areas (Table 2). In addition, diabetes prevalence differed by sex ($p=0.001$, χ^2 test), with a higher mean prevalence across all levels of urbanisation in men (8.7% [SE 0.5]) vs women (6.7% [0.4]).

Across provinces, diabetes prevalence ranged from 4.8% to 12.1% (Table 3). Total urbanisation and the 12 urbanisation components varied across provinces, indicating a differential degree of urbanisation across the provinces.

Over half of the individuals with diabetes in this study were not previously diagnosed by a doctor (ESM Table 4). The proportion of the sample with undiagnosed and diagnosed diabetes differed across urbanisation level and province.

High vs low urbanisation was associated with an approximate twofold higher diabetes prevalence (men OR 2.02, 95% CI 1.47, 2.78; women OR 1.94, 95% CI 1.35, 2.79) after controlling for individual-level factors and accounting for clustering at the community and province levels (Table 4). In general, higher urbanisation across the 12 components was positively associated with diabetes prevalence, albeit with variation in statistical significance (ESM Figs 1 and 2). Comparatively stronger associations were observed for men.

Low variance and ICC at the community level for men suggests low variation in diabetes across communities (Table 4). In contrast, community-level variance was larger in women than men, and women had a higher ICC at the community level than at the province level. Thus, for women, variation in diabetes prevalence across communities remained even after urbanisation components were taken into account.

Although BMI was not an empirical confounder of the association between urbanisation and diabetes prevalence, findings from models with adjustment for BMI are shown in ESM Table 2. With BMI in the model, ORs were somewhat attenuated, but in general showed the same pattern of direction and statistical significance.

While models for total diabetes prevalence are presented in Table 4, we also considered associations with urbanisation for diagnosed and undiagnosed diabetes (ESM Table 5). We observed a similar pattern of association for diagnosed diabetes at high vs low urbanisation (men OR 5.86, 95% CI 3.08, 11.15; women OR 4.03, 95% CI 2.09, 7.78) after controlling for individual-level factors and accounting for clustering at the community and province levels. In contrast, the prevalence of undiagnosed diabetes did not differ by high vs low urbanisation level (men OR 1.25, 95% CI 0.84, 1.85; women OR 1.38, 95% CI 0.88, 2.17).

Sensitivity analyses related to diagnostic criteria for diabetes suggest that the use of HbA_{1c} alone (ESM Table 4), a combination of high HbA_{1c} and/or FBG (ESM Table 5), and FBG alone (Table 4) yielded findings similar in direction and magnitude.

In sensitivity analyses, we also compared an administratively defined classification of urban vs rural communities

Table 2 Individual-level characteristics by low, medium and high urbanisation level

Characteristic	Sex	Low urbanisation ^a n=2,571	Medium urbanisation ^a n=2,588	High urbanisation ^a n=2,582	p value for ANOVA/ χ^2 test within sex ^b
Age, mean (SE)	Male	50.5 (0.4)	50.3 (0.4)	52.1 (0.4)	0.005
	Female	50.4 (0.4)	50.3 (0.4)	52.4 (0.4)	<0.001
Diabetes ^{c,d} , % (SE)	Male	5.8 (0.7)	8.8 (0.8)	11.7 (0.9)	<0.001
	Female	4.5 (0.6)	6.5 (0.7)	9.1 (0.8)	0.001
Overweight (WHO criterion [BMI 25–30 kg/m ²]), % (SE) [15]	Male	21.0 (1.2)	27.8 (1.3)	28.6 (1.3)	<0.001
	Female	23.9 (1.2)	26.7 (1.2)	25.9 (1.2)	0.21
Obese (WHO criterion [BMI \geq 30 kg/m ²]), % (SE) [15]	Male	2.8 (0.5)	3.8 (0.6)	4.9 (0.6)	0.03
	Female	4.2 (0.5)	4.7 (0.6)	4.0 (0.5)	0.64
Overweight (Chinese criterion [BMI 24–28 kg/m ²]), % (SE) [14]	Male	27.1 (1.3)	32.9 (1.4)	36.0 (1.4)	<0.001
	Female	28.9 (1.2)	29.5 (1.2)	29.2 (1.2)	0.93
Obese (Chinese criterion [BMI \geq 28 kg/m ²]), % (SE) [14]	Male	5.9 (0.7)	9.5 (0.8)	10.8 (0.9)	<0.001
	Female	10.0 (0.8)	11.0 (0.8)	10.4 (0.8)	0.69

^aUrbanisation level based on tertiles of urbanisation index (range 30.4–106.6) representing low (<59.0), medium (59.0–82.2) and high (\geq 82.3) levels of urbanisation

^bStatistical testing across urbanisation level using ANOVA for continuous outcomes and χ^2 test for categorical outcomes

^cDiabetes defined as a FBG \geq 7.0 mmol/l or doctor diagnosis of diabetes

^dDiabetes prevalence differed by sex ($p=0.001$, χ^2 test), with a higher mean diabetes prevalence in men (8.7% [SE 0.5]) vs women (6.7% [0.4]) across all levels of urbanisation.

to the multidimensional urbanisation measure used in the central analysis. ORs for diabetes prevalence suggest a similar direction and magnitude for high vs low urbanisation using the traditional (men OR 1.54, 95% CI 1.21, 1.97; women OR 1.38, 95% CI 1.04, 1.84) vs multidimensional (men OR 2.02, 95% CI 1.47, 2.78; women OR 1.94; 95% CI 1.35, 2.79) measure.

Discussion

Our findings suggest that diabetes prevalence is elevated across areas of low and high urbanisation in China. However, using a detailed and multidimensional measure of urbanisation, we found that individuals living in more-urbanised areas had a twofold higher diabetes prevalence than individuals living in less-urbanised areas, even after community and province of residence, age, sex and household income were accounted for. These findings are concerning given the vast number of affected individuals. China is home to the world's largest population and approximately 80% of diabetes deaths occur in low- and middle-income countries [17]. Although we observed higher diabetes prevalence in more-urbanised areas, the relatively high level of diabetes in less-urbanised areas suggests that these areas are still in need of attention. We also observed variation in diabetes prevalence across specific components of urbanisation. Our findings suggest that community economic factors, modern markets, communications and transportation

infrastructure might be areas deserving of further research and could potentially be the focus of community-level diabetes prevention efforts.

Although others have examined disparities in diabetes [4, 18, 19] and other have examined noncommunicable diseases [2, 20, 21] across regions of China using traditional measures of urbanisation, our analysis improves on these studies in three key ways. First, the CHNS intentionally sampled the capital city and rural areas within each province to span the breadth of economic development within provinces and thus provides substantial variation across more- and less-urbanised areas. Second, our multidimensional measure of urbanisation, which has been tested for reliability and validity [9], captures heterogeneity in a variety of services and infrastructure within and between high- and low-urbanised areas. This multidimensional measure is particularly relevant in China or other countries undergoing rapid and asymmetric urbanisation, which is inadequately captured by dichotomous classification of urbanicity [22–24]. Having observed the separate associations between the 12 urbanisation index components and diabetes prevalence, we suggest that some potentially modifiable factors, namely community economic factors, modern markets, communications and transportation infrastructure, might be foci for further research and community-level policy.

Third, we used multilevel modelling to account for the clustering of individuals at the community and province levels. This technique is critical given our research question, as well as the nature of the sampling method, and the fact

Table 3 Individual-level and community-level urbanisation characteristics, by province

	Heilongjiang	Guizhou	Henan	Liaoning	Shandong	Hubei	Guangxi	Hunan	Jiangsu	Total sample
Individual-level factors										
Sample, no. (%)	847 (10.9)	657 (8.5)	856 (11.1)	723 (9.3)	879 (11.4)	808 (10.4)	968 (12.5)	956 (12.4)	1,047 (13.5)	7,741 (100.0)
Diabetes, % (SE) ^{a,b}	6.3 (0.8)	5.6 (0.9)	8.6 (1.0)	11.6 (1.2)	12.1 (1.1)	5.8 (0.8)	4.8 (0.7)	6.8 (0.8)	7.7 (0.8)	7.7 (0.3)
Household income (Yuan), mean (SE) ^b	34,017 (1,183)	32,345 (1,514)	31,284 (1,428)	37,685 (1,164)	39,420 (1,503)	38,696 (1,693)	32,842 (1,229)	39,759 (1,709)	56,511 (1,474)	38,622 (496)
Community-level total urbanisation level^c										
Low, no. (%) ^d	560 (66.1)	351 (53.4)	493 (57.6)	108 (14.9)	233 (26.5)	213 (26.4)	361 (37.3)	109 (11.4)	143 (13.7)	2,571 (33.2)
Medium, no. (%)	118 (13.9)	235 (35.8)	168 (19.6)	277 (38.3)	384 (43.7)	340 (42.1)	211 (21.8)	424 (44.4)	431 (41.2)	2,588 (33.4)
High, no. (%)	169 (20.0)	71 (10.8)	195 (22.8)	338 (46.8)	262 (29.8)	255 (31.6)	396 (40.9)	423 (44.3)	473 (45.2)	2,582 (33.4)
Median (IQR) ^b	49.7 (43.1, 75.6)	52.8 (48.0, 68.3)	54.1 (40.6, 81.2)	65.6 (56.1, 89.1)	67.5 (54.7, 82.7)	68.9 (52.8, 85.3)	71.0 (47.5, 85.8)	73.9 (61.0, 92.0)	82.1 (56.3, 90.6)	65.0 (50.9, 85.4)
Community-level 12 urbanisation index components^d										
Population density ^b , median (IQR)	5.0 (3.5, 6.5)	5.5 (5.0, 6.5)	5.5 (5.0, 6.0)	6.5 (5.5, 8.0)	6.0 (5.0, 6.5)	6.5 (5.0, 8.0)	6.0 (5.0, 6.5)	6.0 (5.0, 6.5)	6.5 (5.5, 7.0)	6.0 (5.0, 7.0)
Social services ^b , median (IQR)	1.3 (1.3, 2.5)	1.3 (1.3, 2.5)	2.5 (1.3, 5.0)	2.5 (1.3, 5.0)	1.3 (1.3, 7.5)	2.5 (1.3, 6.3)	2.5 (1.3, 5.0)	5.0 (2.5, 8.8)	1.3 (1.3, 8.8)	2.5 (1.3, 5.0)
Health infrastructure ^b , median (IQR)	6.4 (1.3, 8.8)	6.3 (2.5, 7.5)	5.0 (3.8, 7.5)	7.5 (6.3, 8.8)	7.5 (5.0, 8.8)	5.0 (3.0, 7.5)	5.0 (2.5, 8.0)	7.5 (5.0, 8.0)	7.5 (4.0, 8.8)	7.5 (3.8, 8.8)
Modern markets ^b , median (IQR)	1.5 (0.5, 5.0)	2.0 (1.0, 4.5)	3.5 (1.5, 5.0)	5.0 (3.5, 7.0)	2.5 (2.0, 6.0)	5.5 (4.0, 8.0)	6.0 (1.5, 8.5)	6.3 (4.5, 8.0)	4.5 (2.0, 6.5)	4.5 (1.5, 7.0)
Traditional markets ^b , median (IQR)	5.0 (0.0, 5.0)	5.0 (0.9, 5.1)	2.9 (0.0, 5.0)	2.9 (1.4, 5.0)	5.0 (1.4, 7.5)	5.0 (3.8, 8.8)	5.0 (4.4, 8.8)	5.0 (5.0, 10.0)	5.0 (0.0, 10.0)	5.0 (1.4, 7.9)
Transportation infrastructure ^b , median (IQR)	6.7 (5.0, 6.7)	5.0 (3.3, 6.7)	5.0 (3.3, 6.7)	6.7 (6.7, 6.7)	5.0 (3.3, 6.7)	6.7 (5.0, 6.7)	6.7 (5.0, 6.7)	6.7 (6.7, 6.7)	6.7 (3.3, 6.7)	6.7 (5.0, 6.7)
Communications ^b , median (IQR)	6.5 (5.8, 7.2)	4.9 (4.0, 6.8)	6.9 (5.3, 8.1)	6.5 (6.0, 7.2)	7.0 (6.1, 8.1)	7.2 (6.5, 7.8)	6.6 (5.5, 8.4)	7.0 (5.4, 8.6)	7.9 (6.8, 8.5)	6.8 (5.8, 8.0)
Housing infrastructure ^b , median (IQR)	7.0 (4.9, 9.3)	6.7 (6.1, 8.8)	4.6 (3.8, 8.3)	9.4 (6.9, 9.8)	7.1 (5.1, 9.3)	7.8 (6.3, 9.2)	8.5 (5.8, 9.6)	8.5 (7.6, 9.9)	9.3 (7.3, 10.0)	7.6 (5.9, 9.6)
Sanitation ^b , median (IQR)	4.5 (3.0, 6.0)	6.5 (3.2, 9.0)	4.8 (3.3, 9.5)	8.8 (5.0, 10.0)	7.3 (5.8, 9.2)	7.6 (4.3, 9.8)	7.3 (2.5, 9.5)	8.6 (6.3, 10.0)	9.8 (8.5, 10.0)	7.5 (4.3, 9.8)
Economic activity ^b , median (IQR)	5.0 (3.0, 10.0)	5.5 (3.0, 7.0)	6.0 (2.0, 10.0)	9.9 (3.1, 10.0)	8.0 (3.0, 10.0)	5.8 (3.0, 10.0)	7.0 (5.0, 9.6)	9.0 (5.5, 10.0)	9.5 (6.0, 10.0)	7.0 (4.0, 10.0)
Education ^b , median (IQR)	2.8 (2.4, 3.7)	2.2 (1.6, 3.0)	2.9 (2.4, 4.0)	4.2 (3.0, 4.8)	3.2 (2.6, 4.0)	2.9 (2.4, 3.9)	3.2 (2.6, 3.7)	3.3 (2.8, 4.0)	3.3 (2.5, 4.5)	3.0 (2.5, 4.0)
Education and income diversity ^b , median (IQR)	4.5 (4.5, 5.0)	5.3 (4.5, 6.0)	5.5 (4.5, 5.8)	5.5 (4.5, 6.0)	5.5 (5.0, 6.5)	5.5 (4.5, 6.0)	5.0 (4.5, 5.5)	5.5 (5.0, 6.0)	6.0 (5.0, 6.5)	5.5 (4.5, 6.0)

^aDiabetes defined as a FBG ≥ 7.0 mmol/l or doctor diagnosis of diabetes^bStatistically significant differences in individual-level factors and community-level 12 urbanisation index components across provinces at the $p < 0.05$ level using ANOVA for continuous outcomes, χ^2 test for categorical outcomes, and K-sample test for equivalence of medians^cUrbanisation level based on tertiles of urbanisation index (range: 30.4–106.6) representing low (< 59.0), medium (59.0–82.2) and high (≥ 82.3) levels of urbanisation^dStatistically significant differences in the proportion of low, medium and high total urbanisation level across provinces at the $p < 0.05$ level using χ^2 test

Table 4 ORs and province- and community-level variance for diabetes prevalence across total urbanisation and its components^{a,b}

Urbanisation level or component	Men				Women			
	OR (95% CI)	p value	Province	Community	OR (95% CI)	p value	Province	Community
			Variance (SE)	Variance (SE)			Variance (SE)	Variance (SE)
Total urbanisation level ^c								
Low	1		0.13 (0.08)	0.04	1		0.07 (0.05)	0.02
Medium	1.50 (1.08, 2.08)	0.02		0.00 (0.00)	1.52 (1.04, 2.21)	0.03		0.12 (0.10)
High	2.02 (1.47, 2.78)	<0.001			1.94 (1.35, 2.79)	<0.001		
12 urbanisation index components ^d								
Population density	1.27 (1.08, 1.50)	0.005	0.12 (0.07)	0.04	1.18 (0.97, 1.43)	0.09	0.07 (0.06)	0.02
Social services	1.11 (0.96, 1.28)	0.16	0.14 (0.08)	0.04	1.22 (1.05, 1.43)	0.01	0.06 (0.05)	0.02
Health infrastructure	1.39 (1.07, 1.81)	0.01	0.11 (0.07)	0.03	1.19 (0.88, 1.60)	0.25	0.06 (0.06)	0.02
Modern markets	1.49 (1.17, 1.91) ^e	0.001	0.15 (0.09)	0.04	1.46 (1.10, 1.94)	0.008	0.08 (0.06)	0.02
Traditional markets	1.30 (1.03, 1.64)	0.03	0.15 (0.09)	0.04	1.33 (1.01, 1.74)	0.04	0.09 (0.07)	0.02
Transportation infrastructure	1.11 (1.01, 1.22)	0.04	0.14 (0.08)	0.04	1.18 (1.06, 1.32)	0.004	0.08 (0.06)	0.02
Communications	1.48 (1.22, 1.79) ^e	<0.001	0.13 (0.07)	0.04	1.36 (1.10, 1.69)	0.005	0.06 (0.05)	0.02
Housing infrastructure	1.61 (1.27, 2.05) ^e	<0.001	0.15 (0.09)	0.04	1.32 (1.00, 1.73)	0.05	0.08 (0.06)	0.02
Sanitation	1.59 (1.23, 2.06) ^e	<0.001	0.12 (0.07)	0.04	1.59 (1.19, 2.12) ^e	0.002	0.07 (0.06)	0.02
Economic activity	1.51 (1.19, 1.91) ^e	0.001	0.13 (0.08)	0.04	1.25 (0.95, 1.63)	0.11	0.07 (0.06)	0.02
Education	1.29 (1.15, 1.46) ^e	<0.001	0.12 (0.07)	0.03	1.10 (0.94, 1.29)	0.22	0.06 (0.05)	0.02
Education and income diversity	1.42 (1.20, 1.66) ^e	<0.001	0.10 (0.07)	0.03	1.14 (0.94, 1.39)	0.18	0.07 (0.06)	0.02

^aTwenty-six separate sex-stratified multilevel models for total urbanisation and its components

^bDiabetes defined as a FBG ≥ 7.0 mmol/l or doctor diagnosis of diabetes

^cORs for diabetes prevalence presented for medium (59.0–82.2) and high (≥ 82.3) (range 30.4–106.6) relative to low urbanisation level (<59.0)

^dORs for diabetes prevalence presented for the 75th (high) relative to the 25th (low) percentile of each of the 12 urbanisation index components

^eOR statistically significant at Bonferroni-corrected $p < 0.002$ (type 1 error rate = 0.05 across 24 models) by two-tailed z test

that the main exposures of interest are measured at the community level [11]. Indeed, descriptive statistics suggested differences in the distribution of urbanisation components and diabetes prevalence across provinces. Accounting for these dependencies via random-effects modelling did not attenuate associations between urbanisation and diabetes. Therefore, although community and province of residence were associated with differential diabetes prevalence in China, unmeasured factors remain that might influence diabetes.

We observed variation in the level of urbanisation and prevalence of diabetes across provinces of China. The total measure of urbanisation captures synergies among its components across low-, medium- and high-urbanisation levels, and we observed higher prevalence of diabetes in high- vs low-urbanised areas of China. Our findings across the 12 urbanisation components provide additional information not captured in the overall measure. For example, we observed substantial variation in economy-related factors across provinces and increased diabetes prevalence in communities with high vs low economic activity, education, and education and income diversity. Together, these findings might be attributed to limited/regional free trade agreements, economic development that has favoured eastern China and port cities [25], and a decentralised healthcare system that determines health regulation and infrastructure at community and province levels [5, 6]. Modern markets (which in our study included supermarkets and fast-food restaurants) have also increased in number with economic development [26] and were positively associated with diabetes prevalence. Our findings, namely that multiple economy-related factors are positively associated with diabetes prevalence, suggest that the largest disease burden may be found in wealthier and more modernised areas of China.

Given inter-correlation across the 12 urbanisation components, combining them into a single model was not possible. Furthermore, we were explicitly interested in the net effect of each individual component. Looking at the associations between high vs low levels of the 12 urbanisation components with diabetes prevalence highlights modifiable factors with potential for community-level efforts to prevent and treat diabetes. While our findings are suggestive, further research on these community-level factors is needed before incorporating such factors into policy and programmes. Nonetheless, community-level factors, such as transportation and communication, which were positively associated with diabetes, have been useful for addressing chronic diseases in other developing-country contexts. For example, in Brazil, changes in transportation infrastructure have had an impact on population-level physical activity [27, 28]. In Pakistan, a communications-oriented health education campaign has positively affected nutrition-related chronic disease awareness [29]. Increased alcohol taxes in Thailand have been associated with reductions in alcohol-related mortality [30]. In China, community-level tobacco cessation

efforts have been associated with reductions in smoking [31]. Thus, community-level strategies offer promise in deterring lifestyle-related diseases such as diabetes.

In general, associations between the 12 urbanisation index components and diabetes prevalence appeared to be stronger in men. The extent to which this difference represents a differential relationship between community-level factors and diabetes by sex or whether other factors (e.g. educational and economic opportunities for men vs women) drive this difference is unknown. In the CHNS, a comparatively stronger association between urbanisation and BMI for men has been shown [32]; therefore, it is possible that lifestyle in urban vs rural areas differs by sex.

In contrast, we did not find differences in the association between urbanisation and diabetes prevalence by age. Cumulative exposure to the urban environment has been shown to be related to increased prevalence of overweight and non-communicable diseases [33]; however, because urbanisation is a relatively recent and ongoing phenomenon in China, cumulative exposure to the urban environment may not be greater in older people, so differential associations between urbanisation and diabetes by age may not be apparent.

There are limitations to the current study that should be noted: (1) the main exposure variables are based on community-level data and while our urbanisation index measure includes diverse and detailed components of urbanisation, elements such as local food prices and individual-level exposures are not captured; (2) our cross-sectional analysis does not take into account changes in urbanisation over time; (3) diabetes prevalence is influenced by a number of factors in addition to incidence; thus, differential mortality among individuals with diagnosed or undiagnosed diabetes may influence prevalence of diabetes (e.g. diabetes prevalence may be artificially lower in areas with higher mortality rates [34]). However, our findings do provide insight related to province- and community-level factors that might be targets for policy in China and other recently urbanised countries. Although we used just one FBG measure to classify individuals with diabetes, this is standard epidemiological practice for population-based studies. While FBG and HbA_{1c} capture different populations of people with diabetes [35, 36], we found consistency in results using HbA_{1c} and FBG in terms of the effect size and direction of association with overall urbanisation and its components, albeit with some differences by sex. Researchers in Korea found a greater proportion of women with diabetes when using HbA_{1c} vs FBG as the diagnosis criterion [37]; however, the reasons why HbA_{1c} would capture more women with diabetes are unknown. We are not able to distinguish type 1 from type 2 diabetes in our sample, but the incidence of type 1 diabetes in China is among the lowest in the world, with an estimate of 0.1 per 100,000 each year [38]. Due to differences between the WHO and Chinese BMI cut-off points for overweight and obesity, both

were included in Table 2. While the internationally recognised WHO cut-off points facilitate comparison across studies and nations, the lower BMI Chinese cut-off points [15] reflect the higher noncommunicable disease risk at lower BMI in Chinese populations [14].

Conclusion

Diabetes is becoming a major public health concern across China. Using a multidimensional measure of urbanisation, which captures healthcare, economic systems, services and infrastructure, we observed a twofold higher prevalence of diabetes in more- vs less-urban areas, even after community and province of residence, age, sex and household income were accounted for. We also observed that high (i.e. more urbanised) values for some components of urbanisation, particularly community economic factors, modern markets, communications and transportation infrastructure, were separately and positively associated with diabetes. As such, these factors might present opportunities for community-level diabetes prevention efforts.

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