Neighbourhood Environment Features and Active Commuting in Chennai, India

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Abstract

Background

Few studies have assessed built environment correlates of active commuting in low-and-middle-income countries, but the different context could yield distinct findings. This cross-sectional study investigated associations between home neighbourhood environment characteristics and active commuting in Chennai, India.

Methods

Adults (N = 370, 47.2% female, mean age = 37.9 years) were recruited from 155 wards in the metropolitan area of Chennai in southern India between January and June 2015. Participants self-reported their usual mode of commute to work, with responses recoded into three categories: (1) multi-modal or active commuting (walking and bicycling); (2) public transit; and (3) private transport. Environmental attributes around participants' homes were assessed using the Neighborhood Environment Walkability Scale-India (NEWS-India). Associations between environmental characteristics and likelihood of active commuting and public transit use were modelled using logistic regression with private transport (driving alone or carpool) as the reference category, adjusting for age, gender, and household car ownership.

Results

Consistent with other international studies, participants living in neighbourhoods with a mix of land-uses and a transit stop within a 10-minute walk from home were more likely to use active commuting (both p < 0.01). Land-use mix was significantly associated with the use of public transit compared to private transport (aOR = 5.2, p = 0.002). Contrary to findings in high-income countries, the odds of active commuting were reduced with improved safety from crime (aOR = 0.2, p = 0.003), aesthetics (aOR = 0.2, p = 0.005) and street connectivity (aOR = 0.2, p = 0.003).

Conclusions

Different environmental attributes were associated with active commuting, suggesting that these relationships are complex and may distinctly differ from those in high-income countries. Unexpected inverse associations of perceived safety from crime and aesthetics with active commuting emphasize the need for high quality epidemiologic studies with greater context-specificity in the study of physical activity in LMICs. Findings have public health implications for India and suggest that caution should be taken when translating evidence across countries.

Key words

India; active commuting; public transit; physical activity; built environment

Background

Physical inactivity is the fourth leading risk factor for global mortality and a major contributor to noncommunicable diseases (NCDs) [1, 2]. Worldwide, an estimated 5.3 million deaths can be attributed to insufficient physical activity [3]. This is particularly important in the context of low-and-middle-income countries (LMICs), where a bulk of the NCD burden falls—LMICs account for 80% of deaths from NCDs and physical inactivity [4]. In India, NCDs account for 53% of the disease burden, with incidence rates for heart diseases and cancer—two of the leading causes of mortality in the country—on a sharp uptick [5].

Walking and cycling are recommended forms of moderate-to-vigorous physical activity that can serve as means of travel to substitute for short car trips and are feasible ways for people to incorporate regular physical activity into their daily lives [6]. The use of public transit usually involves walking or cycling to and from bus stops or train stations and has the potential to contribute to overall physical activity. The promotion of walking, bicycling, use of public transport and other non-motorized means of travel, collectively referred to as active commuting or active travel is as a key strategy to increase physical activity [7]. Active commuting has the potential to be incorporated into people's daily routines and might therefore be more easily adopted and maintained than other forms of physical activity (especially recreational activity) [8-10]. Active commuters tend to achieve greater levels of physical activity than those who use automobiles [11].

Active commuting has been specifically associated with reduced cardiovascular risk, lower obesity, higher physical fitness, and weight control in adults [12, 13]. Despite these benefits, mass adoption of private motorised transport and the design of cities to favour automobile use has likely resulted in declining levels of active commuting and a rise in population prevalence of overweight, obesity, and related NCDs in India [14]. Our previous research from India showed that urban living was associated with lower leisure-time and transport physical activity and increasingly sedentary lifestyles [15, 16]. Active commuting may be influenced by characteristics of the neighbourhood built environments, yet few studies have quantified the associations between these factors in India.

The World Health Organization's Global Action Plan for the Prevention and Control of NCDs 2013-2020 has stressed the need for, "urban planning and transport policies to improve the accessibility, acceptability and safety of, and supportive infrastructure for, walking and cycling," and urged member states to ensure, "the creation and preservation of built environments with a particular focus on providing infrastructure to support active commuting" [1]. In a number of high-income countries, most notably the Netherlands, Denmark, Sweden, Finland, and Germany, active commuting to work, school, or to run errands is commonplace [17]. These countries have implemented effective policies and made investments in urban environments to increase active commuting. However, such efforts remain largely under-developed in LMICs like India, which suffers from fundamental concerns such as overcrowded street conditions, lack of activity-supportive infrastructure, and poor enforcement of traffic rules and regulations, all of which create barriers to active commuting [16, 18].

Reversing the decline in walking and bicycling for travel related purposes, especially for short trips, presents a major opportunity for improving physical activity worldwide. To develop effective interventions to promote active modes of transport as alternatives to car driving, an understanding of the factors associated with this particular behavior is required [7]. Public health experts contend that substantial changes in the built environment are needed if active commuting is to become a widely accepted option [7]. A few environmental and psychological factors and policies have been associated with active commuting or the use of public transit, with commuting distance being the strongest and most consistent factor [19-21]. Additionally, the provision of sidewalks, crosswalks, and dedicated bicycle facilities on major roads, introduction of traffic signals for pedestrians, bicyclists, and the use of traffic calming devices is known to increase active travel [22]. However, nearly all of the evidence is from high-income countries. With the exception of some settings in South America [23-25] and Africa [26, 27], data on active commuting in LMICs are scarce.

To our knowledge, there are no previous studies examining the associations between home neighbourhood characteristics and active commuting in India. The objective of the current study was to examine built environments correlates of usual commuting mode in an adult population in Chennai, India.

Methods

Sampling and Recruitment

Participants (N=370; female=47.2%) were recruited from 155 wards in the metropolitan area of Chennai in southern India (5th largest city; population as per 2011 Census = 8,653,521) between December 2014-June 2015. Wards were stratified to maximize variance in neighborhood walkability and socio-economic status (SES) to enhance the representativeness of the sample because low-SES populations tend to be underrepresented in studies of this nature [28, 29].

Inclusion and exclusion criteria for participants were based on studies conducted by by the IPEN (International Physical Activity and Environment Network) study protocol [30] in LMICs such as Africa, Brazil, and China [27, 30, 31]. Eligibility criteria included: (i) current residents of the Chennai metropolitan area; (ii) residents for at least 6 months; (iii) 18–65 years of age; (iv) being able and willing to answer questions in English or Tamil (official language in the study region); (v) not having any disability that prevented independent walking; and (vi) no visible signs of cognitive impairment. One individual per household was recruited to ensure independence of observations. Sample size was determined using a moderate-to-large effect size (effect size statistic [d=] 0.75), which is greater than what has been used in previous IPEN studies in LMIC contexts [32, 33]. To reduce the risk of bias, this study adopted a sampling strategy to represent diverse environments was used. In addition, the use of reliable measures and standardized protocols for data collection, including training of study personnel, minimized inter-observer variability when multiple field workers were gathering and entering data. Details of neighborhood stratification, sampling, recruitment, and survey properties have been described in detail elsewhere [34].

Measures

Main outcome

The main outcome of interest was commuting mode. Participants self-reported their usual mode of commuting to work. The dependent variable response options included: walk, bicycle, public transport (bus, train, auto rickshaw), private transport (car, motorcycle) and combination of walk/bicycle with public and private transport modes. Participants could select multiple modes. To account for health benefits, unless a walking trip was at least 10 minutes (e.g., to or from a public transit stop or parking lot), participants were asked not to report it as a 'walk' [6]. Commuting mode was recoded into three categories: (1) multi-modal or active commuting (walking and bicycling); (2) public transit; and (3) private transport. Participants who reported multi-modal travel (using both active and non-active modes) were grouped together with active commuting to capture the active components of multi-modal commuters. Participants who reported using multi-modal travel involving only non-active modes were grouped together with private transport.

Independent variables and covariates

Socio-demographic and individual characteristics

Self-reported data on age, gender, race, marital status, education, household income, number of vehicles in the household, number of children younger than 18 years old in the household, and chronic conditions including heart disease, diabetes, and cancer were elicited from participants. The sociodemographic characteristics were categorized into 2–4 categories where appropriate (Table 2).

Home neighborhood built environment features

Built environment features of participants' home neighborhoods were assessed using the Neighborhood Environment Walkability Scale-India (NEWS-India) consisting of 91 items grouped into eight subscales as listed in Table 1 [34]. NEWS item scoring and subscale score calculations followed the NEWS-Adult scoring scheme recommended by the IPEN study protocol [30]. Four-point Likert-type scale response options for all NEWS-India items ranging from 1 (strongly agree) to 4 (strongly disagree) were combined as "agree" (strongly agree, agree) and "disagree" (disagree, strongly disagree). All NEWS-India items were positively scored to ensure that a higher score denoted a more activity-supportive neighbourhood. Based on scoring procedures and results from confirmatory factor analyses of NEWS conducted previously [17, 30], two aggregate scores (mean of subscales) were computed: (i) A conceptual NEWS-India score including variables that are known to be related to active commuting (Table 1, subscales a-e, excluding aesthetics, safety from crime and safety from traffic since they are more likely to affect leisure physical activity); and (ii) A composite score specific to Chennai consisting of only the built environment variables that were significantly positive or not significant (Table 1,

subscales a, b, e). Test-retest reliability of NEWS-India items have been previously established, with reliability coefficients that were generally high (ICC = 0.48 to 0.99) with almost perfect strength of agreement, indicating that the items were generally reliable [34].

Data Analysis

Data were analysed in two multiple logistic regression models with private transport (i.e., driving alone or carpool) as the reference commuting mode to examine: (1) the correlates associated with using multi-modal or active commuting (Table 3), and (2) the correlates associated with using public transit (Table 4). A summary of variables used in multiple logistic regression models is presented in Table 1. Descriptive statistics of the sample population are presented in Table 2. Unadjusted and adjusted odds ratios are presented in Tables 3 and 4. All models were adjusted for age, gender, and household car ownership. These covariates were selected to control for the confounding effects of these variables as shown in similar studies conducted in LMICs [17, 30, 35]. Data were analyzed using the Statistical Package for the Social Sciences (SPSS) version 21. By default, missing values were removed in SPSS in a list wise manner, leaving only cases with all variables in the final regression model.

	Model 1	Model 2
Dependent Variables	Active Travel	Public Transit
	vs.	vs.
	Private Transport (reference category) Private Transport (reference category)
Independent Variables	1. Public transit access (Transit	stop within a 10 min walk from home)
	2. Commute distance to work (l-5km, 5.1-10km, 10.1-15km, >15km)
	 NEWS-India Subscale Scores Residential density (7 ite Land use mix–diversity (4 Land use mix–access (7 i Street connectivity (5 ite Infrastructure for walkin Safety from traffic (6 ite Safety from crime (4 iter Aesthetics (6 items) 	(Mean of items): ms) 43 items) tems) ms) g and cycling (13 items) ns)
	4. NEWS-India Aggregate score	(Mean of subscales a-e) ¹
Covariate/Controls	1. Age (continuous)	
-	2. Gender (dichotomous)	
	3. Household car/motor vehicle	ownership (dichotomous)

Table 1. Summary of variables used in multiple logistic regression models.

¹NEWS-India aggregate score excluding aesthetics, safety from crime and safety from traffic.

Descriptive Characteristics	
Age (in years) Mean (SD)	37.9 (15.3)
Gender (n, %)	
Female	199 (54.2)
Male	166 (45.2)
Marital Status (n, %)	
Married	226 (61.2)
Not Married	143 (38.8)
Religion (n, %)	
Hindu	304 (82.2)
non-Hindu	65 (17.6)
Educational Level (n, %)	
Uneducated	48 (13.0)
Primary-Middle School	57 (15.5)
High School or Diploma	79 (21.5)
Graduate or Professional	184 (49.7)
Monthly Family Income in US Dollars (n, %)	
≤ 80	74 (25.3)
81-200	43 (14.7)
201-549	24 (8.2)
≥ 550	152 (51.9)
Marsh Chattan (m. 94)	
	404 (07 F)
Unemployed	134 (37.5)
Blue collar	112 (31.4)
White collar	111 (31.1)
Note: 1 US Dollar=approx. 65.69 Indian Rupe	es (average

Table 2. Descriptive statistics of the sample population (n = 367)

Note: 1 US Dollar=approx. 65.69 Indian Rupees (average currency exchange rate, January-April 2015); cut-off values in table based on SES classification for India by Gururaj and Maheshwaran (2014) [36].

	Multi modal or active commuting (n=56)						
	Unadjusted logistic				Adjusted logistic		
	regressions			regressions			
	OR	95%CI	р	aOR	95%CI	р	
Public transit access							
Transit stop within a 10 min walk from home (ref:							
Disagree)							
Agree	4.5	2.1-9.6	<.001	5.0	1.7-14.4	0.003	
Commuting distance (ref: >15 km)							
1-5 km	22.5	2.8-179.4	0.003	6.4	0.7-55.4	0.09	
5.1-10 km	2.1	0.2-25.3	0.6	0.0	0.0-0.0	1.0	
10.1-15 km	1.9	0.2-22.7	0.6	1.6	0.1-20.4	0.7	
Residential density (ref: Low)							
High	2.2	1.1-4.2	0.02	1.1	0.4-2.7	0.9	
Land use mix-diversity (ref: Low)							
High	7.7	3.5-17.0	<.001	6.8	2.3-20.6	.001	
Land use mix-access (ref: Disagree)							
Agree	0.5	0.2-1.0	0.05	0.4	0.1-1.2	0.1	
Street connectivity (ref: Disagree)							
Agree	0.2	0.1-0.4	<.001	0.2	0.1-0.6	0.003	
Infrastructure for walking, bicycling (ref: Disagree)							
Agree	2.1	1.0-4.5	0.05	2.6	0.9-7.1	0.07	
Safety from traffic (ref: Disagree)							
Agree	0.8	0.4-1.7	0.6	0.8	0.3-2.3	0.7	
Safety from crime (ref: Disagree)							
Agree	0.2	0.1-0.4	<.001	0.2	0.1-0.6	0.003	
Aesthetics							
Agree	0.3	0.1-0.7	0.01	0.2	0.0-1.0	0.05	
NEWS-India Conceptual Score (ref: Low)	2.1	1.1-4.1	0.03	1.1	0.4-2.6	0.9	
High							
NEWS-Chennai Composite Score (ref: Low)							
High	2.1	1.1-4.1	0.03	1.1	0.4-2.6	0.9	

Table 3. Crude and adjusted odds ratios examining associations between multi modal or active commuting vs. private transport and home neighbourhood supports in Chennai, India (Model 1).

CI: Confidence Interval

Unadjusted logistic regressionsAdjusted logistic regressionsAdjusted logistic regressionsAdjusted logistic regressionsAdjusted logistic regressions p aOR $95\%CI$ p p aOI AGI AG	Reference: Private Transport (n=111)	Public transit (n=52)						
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High 5.2 2.3-11.7 <.001	Land use mix-diversity (ref: Low)							
Land use mix-access (ref: Disagree) 0.4 0.2-0.9 0.02 0.6 0.2-1.7 0.4 Street connectivity (ref: Disagree) 0.2 0.1-0.4 <.001	High	5.2	2.3-11.7	<.001	5.2	1.8-14.9	.002	
Agree 0.4 0.2-0.9 0.02 0.6 0.2-1.7 0.4 Street connectivity (ref: Disagree) 0.2 0.1-0.4 <.001 0.2 0.1-0.6 0.004 Infrastructure for walking, bicycling (ref: Disagree) 0.9 0.4-2.2 0.8 1.3 0.4-3.7 0.7 Safety from traffic (ref: Disagree) 0.9 0.4-2.2 0.8 1.3 0.4-3.7 0.7 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Mesthetics 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Land use mix-access (ref: Disagree)							
Street connectivity (ref: Disagree) 0.2 0.1-0.4 <.001 0.2 0.1-0.6 0.004 Infrastructure for walking, bicycling (ref: Disagree) 0.9 0.4-2.2 0.8 1.3 0.4-3.7 0.7 Safety from traffic (ref: Disagree) 1.7 0.9-3.4 0.1 1.9 0.8-4.5 0.2 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Mesthetics 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Agree	0.4	0.2-0.9	0.02	0.6	0.2-1.7	0.4	
Agree 0.2 0.1-0.4 <.001	Street connectivity (ref: Disagree)							
Infrastructure for walking, bicycling (ref: Disagree) 0.9 0.4-2.2 0.8 1.3 0.4-3.7 0.7 Safety from traffic (ref: Disagree) 1.7 0.9-3.4 0.1 1.9 0.8-4.5 0.2 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.0 0.7 Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.0 0.7 Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.0 0.7 Messhetics 0.7 0.3-1.6 0.4 1.0 0.4-2.0 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Agree	0.2	0.1-0.4	<.001	0.2	0.1-0.6	0.004	
Agree 0.9 0.4-2.2 0.8 1.3 0.4-3.7 0.7 Safety from traffic (ref: Disagree) 1.7 0.9-3.4 0.1 1.9 0.8-4.5 0.2 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Infrastructure for walking, bicycling (ref: Disagree)							
Safety from traffic (ref: Disagree) 1.7 0.9-3.4 0.1 1.9 0.8-4.5 0.2 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Aesthetics Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Agree	0.9	0.4-2.2	0.8	1.3	0.4-3.7	0.7	
Agree 1.7 0.9-3.4 0.1 1.9 0.8-4.5 0.2 Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Aesthetics 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Safety from traffic (ref: Disagree)							
Safety from crime (ref: Disagree) 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Aesthetics 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Agree	1.7	0.9-3.4	0.1	1.9	0.8-4.5	0.2	
Agree 0.6 0.3-1.1 0.09 0.8 0.4-2.0 0.7 Aesthetics Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) High 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Safety from crime (ref: Disagree)							
Aesthetics Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) High 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Agree	0.6	0.3-1.1	0.09	0.8	0.4-2.0	0.7	
Agree 0.7 0.3-1.6 0.4 1.0 0.4-2.6 0.9 NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7 High 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	Aesthetics							
NEWS-India Conceptual Score (ref: Low) 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7 High	Agree	0.7	0.3-1.6	0.4	1.0	0.4-2.6	0.9	
High	NEWS-India Conceptual Score (ref: Low)	1.3	0.7-2.4	0.5	0.8	0.4-2.0	0.7	
	High							
NEWS-Chennai Composite Score (ref: Low)	NEWS-Chennai Composite Score (ref: Low)							
High 1.3 0.7-2.4 0.5 0.8 0.4-2.0 0.7	High	1.3	0.7-2.4	0.5	0.8	0.4-2.0	0.7	
Note: aOR: Adjusted Odds Ratios; adjusted for age (continuous), gender (categorical), and household car	Note: aOR: Adjusted Odds Ratios; adjusted for age (cont	inuous), g	ender (cate	gorical), a	nd hou	sehold car		

Table 4. Crude and adjusted odds ratios examining the associations between public vs. private transport and
home neighbourhood supports in Chennai, India (Model 2).Reference: Private Transport (n=111)Public transit (n=52)

CI: Confidence Interval

Results

The availability of a transit stop within a 10-minute walk around homes was strongly associated with an increase in the likelihood of active commuting (aOR=5.0, p=0.003) and a trend toward higher likelihood of public transit usage, although this relationship was not significant (aOR=2.5, p=0.08). Among the built environment characteristics, a mix of land-uses was related to higher active commuting (aOR=6.8, p=0.001) and with use of public transit compared to private transport (aOR=5.2, p=0.002). The availability of supportive infrastructure for walking and cycling tended to improve the odds of active commuting (aOR=2.6, p=0.07), but this relationship was not significant. Aesthetics (aOR=0.2, p=0.003) and safety from crime (aOR=0.2, p=0.05) were significantly associated with reduced likelihood of active commuting. Street connectivity also was associated with a reduced likelihood of active commuting (aOR=0.2, p=0.004). Shorter commuting distances (1-5 km) explained uptake of active commuting (aOR=6.4, p=0.09) but this relationship was not significant in the adjusted model. Commuting distance was not associated with use of public transit.

In unadjusted models, the NEWS-India conceptual score and the NEWS-Chennai composite score significantly predicted an increase in odds of active commuting by approximately two times (OR = 2.1, p=0.03), but they were not associated with use of public transit (OR = 1.3, p=0.5). In adjusted models, both scores were neither associated with active commuting (aOR=1.1, p=0.9) nor public transit use (aOR=0.8, p=0.7).

Discussion

Principal findings

This study provides new evidence on neighbourhood environment attributes associated with active commuting in a rapidly urbanizing LMIC context. A large majority of the prior evidence on built environment correlates of active commuting was limited to high-income countries, and more recently, some LMICs in South America and Africa [37, 38]. Findings from this initial study in India highlight some contrasting results in comparison with those reported in high-income countries, emphasizing the need for more high quality epidemiologic studies from LMICs. In general, we found that supportive environments were associated with more active commuting and public transit use. Residents living in neighbourhoods with a diversity of destinations and availability of walking and bicycling facilities were more likely to choose active modes of commuting. Access to public transport in the neighbourhood can act as a facilitator for a more active lifestyle among its residents. However, some surprising results were that street connectivity, safety from crime, and aesthetics were inversely related to active commuting. These results raise the possibility that built environment attributes found to facilitate active commuting in higher-income countries do not fully generalize to LMICs. Given that active commuting is decreasing in LMICs while NCDs are increasing, there is a need for more research in LMICs to inform context-specific built environment strategies to increase physical activity.

Comparison with existing literature

Several land-use and transportation factors are known to be associated with active commuting [9, 37, 39, 40]. Consistent with other international studies from both high-income countries and LMICs, the strongest correlate of active commuting and use of public transit was land use mix [24, 41-43]. Results from the present study confirmed previously reported associations that the availability of a mix of destinations around homes— shops and stores, recreation facilities, such as parks, walking trails, bike paths, and recreational centers—was associated with higher likelihood of active commuting or multi-modal commuting. Lower commuting distances and the availability of walking and cycling infrastructure were also associated with active commuting, which aligns with previous research. A recent study on longitudinal associations on built environment characteristics found that supportive environments predicted uptake of active commuting—participants living in neighbourhoods with a greater density of employment locations were more likely to engage in and maintain their active commuting [10]. In the US, two studies of university students indicated that the further distance or longer the commuting times between home and university, the lower the likelihood of choosing to use an active commuting mode [44, 45].

In contrast with findings from high-income countries, improved safety from crime and aesthetics reduced the odds of active commuting. Previous research has yielded similar inconclusive results acknowledging that the impact of perceived safety from crime and aesthetics on physical activity behaviors in residential neighborhoods needs careful examination [46]. Some studies suggest that higher fear of crime was associated with lower levels of walking, and higher perceived danger for pedestrians and cyclists was associated with

increases in car use [47], although some studies found no associations between changes in perceptions of safety and walking [48]. These mixed findings may relate to the complexity of measuring crime, which likely to be sensitive to time of occurrence, location, people's perceptions and social context [49].

In our analyses, short commuting distance from home to work demonstrated a strong association with active commuting in adjusted models, which confirms previous research [39], but did not remain significant in adjusted models. Studies have shown distance from home to work was a strong predictor of uptake and maintenance of active commuting, and local planners may be able to co-locate new residential developments and workplaces, thus reducing the distances required to travel to work [10]. In unadjusted models with NEWS-India aggregate scores, the active commuting model was significant, while the public transit model was not. An explanation for this may be that the built environment variables are more relevant to walking and bicycling rather than motorized transport [50]. It was surprising that the Chennai-specific model did not perform better than the full conceptual model that included some variables with inverse associations. The weak performance of the conceptual and composite variable could be due to the inclusion of several variables with inverse associations with active commuting (e.g., street connectivity, land use mix-access), suggesting that a different combination of environmental variables may be optimal in some LMIC contexts. Furthermore, the loss of significance in adjusted models could be due to the inclusion of car ownership that may be acting as a mediating variable between the built environment and commute mode choices. After car ownership, age and gender may be the next most influential factors as studies have found that women exercise less compared to men and people are less active as they get older [51, 52]. In studies from LMICs like Mexico, Colombia and Nigeria, car ownership was negatively associated with transport-based activity levels-higher levels of transport-based activity levels tended to occur in the non-vehicle owners and may be strongly driven by necessity [38, 40, 53]. Thus, evidence is growing to support an interpretation that car owners in LMICs will drive regardless of whether the neighbourhood environment supports active transport. Researchers have recommended that this inverse relationship calls for a need-based framework for understanding active commuting in LMICs, versus the more common choice-based framework [31]. More research is needed to understand and assess additional factors (e.g., socio-economic status, car ownership, commuting patterns, and travel distances) and their relationship to active commuting across larger geographical areas and over extended periods of time in LMIC contexts.

Limitations

The cross-sectional study design and a relatively small sample from a single city in India preclude causal inference and generalizability of results [54]. Demographic differences between the neighbourhoods sampled, residual confounding, and self-selection of individuals into walkable neighbourhoods may further limit generalizability [55-58]. Self-reported measures are subject to bias (e.g., overestimation; social desirability of physical activity; physically active people may notice more built environment features and commuting destinations) [59]. Recent studies have used accelerometers and GPS devices to objectively measure physical activity [40, 60]. However, given the early stages of this research in India, the present study provides initial evidence of active commuting and built environment attributes. A lack of consensus on measuring domain-specific activity (e.g., inadequate details on types of PA and their components, lack of reliable measurement tools) is another limitation of this study and physical activity literature in LMICs [61, 62]. Despite these limitations and a relatively small sample, this study has notable strengths. To the best of our knowledge, this study is among the first to use the validated NEWS-India tool to document built environment features and active commuting behaviours in India.

Implications

The opportunity for encouraging active lifestyles through supportive infrastructure in neighbourhoods—as such, promoting the incorporation of transport-related physical activity into a daily routine—is a key takehome message of this study. Even though some findings were different from those in higher-income countries, mixed land use, pedestrian and bicycling infrastructure, and proximity to transit were all positively related to active transport in Chennai, India. These findings suggest that improved urban design and transport options could improve active transport and physical activity in an Indian context. Within India, this study complements the objectives of on-going national government schemes in India such as the Smart Cities Mission (<u>http://smartcities.gov.in/content/</u>), Swachh Bharat Mission (<u>http://www.swachhbharaturban.in/sbm/home/</u>) and the Atal Mission for Rejuvenation and Urban Transformation (AMRUT, <u>https://amrut.gov.in/</u>) in aspects of urban land use and health promotion. Present results simultaneously inform interlinked agendas including spatial growth management, effective land use utilization, formulation of transport policies to ease traffic

congestion, vision zero initiatives and clean air policies with the potential to establish and extend the field of active living research in India [63].

Conclusion and Next Steps

A significant portion of the NCD burden is concentrated in LMICs, but research on environmental correlates of active commuting is lacking that represents the diverse contexts of LMICs. The costs to the wider Indian economy from a heavy reliance on the private car as a mode of transportation run into tens of billions of dollars every year [64, 65]. The evidence base for increasing the uptake of active commuting is extensive [66], and active commuting presents an untapped opportunity to achieve regular physical activity. Achieving increased active commuting requires coordinated work among various stakeholders, but the benefits are numerous, and go beyond physical inactivity and sedentary lifestyles, including positive impacts on traffic congestion, air pollution, and carbon emissions [66, 67]. However, the present study adds to other evidence from LMICs [35, 37, 48] that car ownership appears to negate any effect of walkable neighborhood design on active commuting. Thus, special efforts may be needed to encourage car owners in LMICs to walk for transport, in addition to activity supportive built environments.

The science of measuring and improving the built environment related to physical activity is inherently multidisciplinary. It is important to recognize that land use mix is not just a research instrument for examining the impact of environments on physical activity, but is a valid planning tool that practitioners and policy makers can use to make neighbourhoods more conducive to active lifestyles [41]. Accurate built environment measures and development of more precise benchmarks through further research may be needed to assist informed decisions about the planning and design of activity-friendly neighbourhoods in India, which may be helpful in improving the long-term health of residents. In India, much remains to be done to measure, adapt and design built environments to promote active commuting. It is therefore essential to promote more collaboration among public health investigators as well as those from non-public health disciplines (e.g., urban design/planning, transportation engineering, sociology) to identify overlapping research priorities. There is a clear need for public health professionals to work closely with other disciplines on research, policy, and practice that will lead to joint efforts to meet societal needs.

Future progress depends on forging effective collaborations across disciplines, improving training and education, increasing the resources provided by funding agencies, and capacity building efforts that cross traditional disciplinary boundaries to facilitate knowledge exchange between developed countries and LMICs. Local evidence can guide regional or national policies to improve built environment factors that directly or indirectly affect transport mode choices have the potential to increase physically active transport. An integrated effort towards encouraging walking, cycling, public transport use and reducing dependency on cars is a promising strategy to significantly increase active commuting, and therefore, physical activity in LMICs.

Abbreviations

NCDs: non-communicable diseases LMICs: low-and-middle-income countries

Declarations

Ethics approval and consent to participate

All study procedures were approved by the Institutional Review Board, Human Research Protection Office, Washington University in St. Louis, USA (study ID 201410052). This study presented no more than minimal risk to participants and was given an exempt status by the Human Research Protection Office at WUSTL. It involved no procedures for which written consent is normally required outside of the research context. No names or identifying information about participants were collected on the survey. All participants were provided with a written statement regarding the research study, in the form of an information sheet. WUSTL IRB approved this study without a requirement for written consent.

Consent for publication

Not applicable

Availability of data and material

The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.

Competing interests

The authors declare that they have no competing interests.

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Authors' Contributions

DA conceived and designed the study, created study protocols; JAH and RCB reviewed study methods and protocols; JFS reviewed study methods, provided guidance on measures, survey items and sampling procedures; DA conducted the study and supervised research assistants for data collection; DA conducted statistical data analysis, interpreted the results, and wrote the manuscript; JAH, JFS and RCB supervised statistical analysis. All authors reviewed and edited the manuscript, and approved the final manuscript as submitted.

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