Contents lists available at ScienceDirect

## Heliyon



journal homepage: www.cell.com/heliyon

#### Research article

CelPress

# Socioeconomic inequalities in depressive symptoms in China: The role of social capital

### Xixi Fu, Shunzhuang Peng, Xing Lin Feng

Department of Health Policy and Management, School of Public Health, Peking University, Beijing, 100191, China

#### ARTICLE INFO

Keywords: Social determinants of health Depressive symptoms Social capital China

#### ABSTRACT

This cohort study used the China Health and Retirement Longitudinal Study (CHARLS, 2015-2018) to investigate the effects of socioeconomic status and social capital to the incidence of depressive symptoms among middle-aged and older individuals in China, incorporating a sample size of 9949 participants. Socioeconomic status, social capital and other explanatory variables were collected in 2015, while depressive symptoms were assessed in 2018. Basic characteristics and social capital measures were compared between urban and rural residents using the chi-square test. Logistic regression was used to explore the relationship between socioeconomic status, social capital and depressive symptoms, and the Karlson, Holm, and Breen (KHB) method was employed to verify the mediating role of social capital. We reported persistent socioeconomic inequalities in depressive symptoms, with rural residents and the illiterate having 1.45 times and 1.34 times higher odds of depression. We ascertained social capital from both the cognitive and structural constructs, where we enriched the measurement of structural social capital from three specific dimensions, i.e., informal interaction, altruism, and formal social participation. We found that both cognitive and structural social capital were associated with lower incidence of depressive symptoms, where informal interaction had the largest effect. The mediation analysis further illustrated that informal interaction contributed most to explain 6 %-12 % of the socioeconomic inequalities in depressive symptoms. These results highlighted the unsatisfied mental wellbeing of the vulnerable older people living in rural areas. The finding suggested that older people may benefit more from personal interactions than formal participations. To fulfill the Health in All vision, government and social organizations should consider how to create opportunities to better integrate the older people into the community.

#### 1. Introduction

Depression has become a global epidemic and one of the main causes of disability worldwide [1], affecting about 280 million people [2] and claiming nearly 800,000 lives each year [3]. Especially after the outbreak of the COVID-19 epidemic, the prevalence of depression, anxiety and other common mental diseases increased by 25 % in the first year alone [4]. Depression disproportionately affects older populations [5,6]. In China, for example, an average of 20 % of the population aged 60 and above were encountering depressive symptoms during 1992–2018 [7]. Prevalence of depression follows the social gradients, where people with lower socio-economic status (SES), whether measured by educational achievement, wealth or occupation, are more susceptible to experiencing

\* Corresponding author. School of Public Health, Peking University, Xueyuan Road 38#, Beijing, 100191, China. *E-mail address:* fxl@bjmu.edu.cn (X.L. Feng).

https://doi.org/10.1016/j.heliyon.2024.e24918

Received 9 June 2023; Received in revised form 12 January 2024; Accepted 17 January 2024

Available online 18 January 2024

<sup>2405-8440/© 2024</sup> The Authors. Published by Elsevier Ltd. This is an open access article under the CC BY-NC-ND license (http://creativecommons.org/licenses/by-nc-nd/4.0/).

depression [8–10]. Except for these individual-level factors, urban-rural residency is also considered as an indicator of SES and an important determinant of depression [11]. However, previous studies reported conflicting findings in different settings for the associations between depression and urban-rural residency from various settings. In many developed countries, the prevalence of depression was nearly 50 % higher amongst urban elderly than the rural ones [12], whilst in most developing countries, including China, older people living in the rural areas had higher prevalence [13,14].

As an upstream determinant of health, social capital is thought to be a predicator of better population health status [15,16], and mental wellbeing [17]. Social capital is an ecological variable and a contextual feature of a community, typically manifested as the degree of citizen participation in the community and the level of trust among community members. Although various taxonomies were suggested, public health literatures normally measured the cognitive and structural dimensions when examining the associations between social capital and populations' mental health [18–20]. Previous studies have found that socioeconomic status was a powerful predictor of social capital [21], and there was a positive correlation between socioeconomic status (household income, educational achievement) and social capital such as social participation and social trust [22].

Prior studies consistently reported negative associations between cognitive social capital and depression, but findings on the relationship between structural social capital and depression were not confirmative [23]. The discrepancies observed in the results could potentially be attributed to the variations in the study design and measurement methods employed for assessing structural social capital [23]. For example, a cross-sectional study conducted among Korean adults aged 65 and older found that structural social capital, measured as social connection, was not associated with depression [24], while another longitudinal study on Mexican older people [25] found that structural social capital, measured as collective action and cooperation, information and communication, empowerment and political action, was negatively associated with depression. In China, findings on the relationship between structural social capital and depression were inconsistent as well [26,27].

Social capital was proposed as a critical factor to improve health equity as manifested in the Health in All initiative [28]. In China, for example, constructing social capital and reducing the socioeconomic inequalities in depression have been endorsed in the Healthy China 2030 strategy under the theme of "Co-construction, Sharing, and Health for All". Previous research suggested that social capital may explain the socioeconomic inequality of depression in theory [29–31]. However, there is a lack of empirical investigations that report on the mediating effects of social capital for the socioeconomic inequalities in depression, for which our systematic search of the literature identified only two papers from Asian countries [27,32]. Both studies were based on cross-sectional designs, which are limited for causal inferences. In addition, the measurement of social capital varied between the studies, leading to inconsistent findings.

Adopting a cohort study design, we explored in this study the socioeconomic inequalities in the incidence of depressive symptoms among middle-aged and older people in China. We assessed participants' exposure to socioeconomic differentials and social capital in 2015 and follow up the occurrence of depressive symptoms in 2018. We enriched the measurement of structural social capital from three dimensions: informal interaction, altruism and formal social participation. By comparing the role of varied dimensions of social capital that mediated the socioeconomic inequalities in depression which were across educational achievement, wealth, and urban-rural settings, respectively, we stressed the importance of informal interaction, and discussed the importance of creating opportunities to better integrate the older people into the community.

#### 2. Conceptual framework

As already known, higher socioeconomic status may predict lower risks of depression and higher stock of social capital. Higher stock of social capital, either in the cognitive or the structural dimension, were associated with better mental wellbeing. Prior studies



Fig. 1. A conceptual framework that models the mediating effect of social capital in explaining socioeconomic inequalities in depression.

also suggested that social capital may explain the socioeconomic gradient of depression in theory [29–31]. Building upon these theories, we have developed a conceptual framework to provide guidance for this research. As shown in Fig. 1, both socioeconomic status and social capital may have a direct impact on depression. In addition, socioeconomic status may also have an indirect impact on depression through the pathways involving social capital. We proposed three specific research questions: 1) Whether better socioeconomic status, either measured in terms of urban-rural residency, wealth, or educational achievement, were associated with lower incidence of depressive symptoms; 2) Whether higher stock of social capital, either measured in cognitive and structural constructs, were associated with lower incidence of depressive symptoms; And 3) How much could each dimension of social capital explain the socioeconomic inequalities of depressive symptoms across each SES indictor.

#### 3. Methods

#### 3.1. Study design

We designed a cohort study to explore the three questions above. The participants' socioeconomic status in 2015, including urbanrural residency, household wealth, and educational achievement, were measured as exposures. In addition, their social capital was assessed, comprising three dimensions of structural social capital and one dimension of cognitive social capital, also in 2015. These exposures were then associated with the outcomes, that is, being identified as having depressive symptoms after follow-up in 2018 using the CES-D 10 scale.

#### 3.2. Data

We adopted a cohort study design, using two waves of survey from the China Health and Retirement Longitudinal Study (CHARLS), where we recruited all eligible participants from the 2015 survey and followed up their occurrence of depressive symptoms in 2018. The CHARLS is a longitudinal study of Chinese people aged 45 and above, by use of a multistage stratified probability proportion [33].

The CHARLS surveyed 21,098 participants in 2015 from 28 provinces, 150 counties/districts and 450 villages/communities across mainland China. Of the 21,098 participants from 2015, 18,136 participants (excluding 889 died and 2073 lost to follow-up) were successfully followed up in 2018, with a success rate of 85.96 %. We then excluded 814 participants who were less than 45 years of age, and 4974 participants who had already recorded depressive symptoms in 2015. Further excluding 2399 participants who did not respond to the depression scales in the 2018 survey, we finally included 9949 adults aged 45 years and older in this study (Fig. 2).

#### 3.3. Measures

**Outcome.** The CHARLS assessed depressive symptoms using the ten-item short form of the Center for Epidemiologic Studies Depression Scale (CES-D 10). The CES-D 10 had demonstrated satisfactory psychometric property [34]. Researchers evaluated the reliability and validity of CES-D 10 in a large sample population and found that the Cronbach coefficient of the simplified scale was greater than 0.8 [35]. It recorded a total score range from 0 to 30. In line with the literature, we categorized a participant as having depressive symptoms should he/she reported a total score of no less than 10 [34].

**Socioeconomic status.** We used three indicators to measure each participant's socioeconomic status separately: urban-rural residency, educational achievement, and household wealth. We defined urban-rural residency based on the observation and verification of observers and the responses of respondents through a question in CHARLS: Was your address in the village or city/town? We categorized participant's educational achievement into three groups: illiterate, primary, and secondary and above. Based on the questionnaire of CHARLS, we included each household's expenses for healthcare, food, education, tourism, transportation, postal and telecommunications, daily necessities, fuel, clothing, heating, furniture, and durable consumer goods. After summing up, the total household expenditure was obtained and distributed to each person based on the number of family members, which was the annual per capita household expenditure. We categorized the participants into quartiles, with Q1 representing the poorest group and Q4 the



Fig. 2. Flowchart of the participant selection process.

richest. To report on comparative results, we further grouped educational achievement into illiterate and non-illiterate, and wealth into poor (Q1 and Q2) and rich (Q3 and Q4) in the mediation analysis.

**Social capital.** The cognitive dimension, that is, norms, values, and reciprocity between individuals, was usually measured in the term of social trust; whilst the structural dimension, that is, relationships and networks among individuals, was mostly measured in the term of organizational membership, social participation, social connection and interactions [23]. We defined four dichotomous variables to measure social capital from two distinct constructs: structural social capital and cognitive social capital.

The CHARLS asked whether the participants engaged in any of the eight social activities during the past month before the survey (Table 1). Building on Engbers [36] and Rotenberg [23], we used participation of any of the eight social activities to define three dimensions of structural social capital: informal interaction, altruism and formal social participation. Informal interaction referred to weather the participants be engaged in at least one of these four activities: (1) interacting with friends (2) playing Ma-jong, playing chess, playing cards, or going to community club (3) going to a sport, social, or other kind of club (4) taking part in a community-related organization. Altruism referred to weather the participants be engaged in at least one of these with you (2) doing voluntary or charity work (3) caring for a sick or disabled adult who does not live with you. And formal social participation referred to weather the participants be engaged in attending an educational or training course.

We measured cognitive social capital based on trust [23]. The CHALRS asked each participant whether he/she having relatives or friends (besides spouse/partner) who would be willing and able to help over a long period of time, should the participant need help for basic daily activities such as eating or dressing. For an answer of yes, we defined the participant to have cognitive social capital, that is, perceived trust to receive others support when he/she needs help.

**Other covariates.** Building on the literature [37,38], we included the following covariates that might be predictors of depression, in order to control potential covariates: Age (grouped as 45–54 years, 55–64 years, 65–74 years, 75 years or older), sex (males, females), ethnicity (Han majority, other minorities), marital status (married/partnered, single), health insurance coverage (Urban Employee Basic Medical Insurance, Urban-rural Residency Basic Medical Insurance or New Rural Cooperative Medical Scheme, other insurances, without insurances), smoking (yes, no), drinking alcohol (yes, no), and number of chronic conditions (0, 1, 2, 3 and above).

#### 3.4. Statistical analysis

We described the cohorts' basic characteristics, and reported on the proportion of participants having each construct or dimension of social capital. We stratified participants by urban-rural residency and perform Chi square test to compare the differences. We performed logistic regression to examine the association between each SES determinants (urban-rural residency, education and wealth), and each construct/dimension of social capital (informal interaction, altruism, formal social participation, and trust), with the incidence of depressive symptoms in the follow-up period, adjusting for all other potential covariates. We employed the Karlson, Holm, and Breen (KHB) method to estimate the mediating effects of each social capital construct/dimension on the socioeconomic inequalities in depression. The KHB [39] method decomposed the total effects of each SES indicator into two components: direct effects and indirect effects, and then reported on, for each of the SES indicators, the proportion of socioeconomic inequalities in the outcome that were explained by each mediating variable. Since depressive symptom was measured as a dichotomous variable, we specified a logit link function in the KHB model. All analysis was performed using Stata 13.1.

#### 3.5. Subgroup analysis

The dual structure between urban-rural settings in China is prominent, with significant infrastructural, policy, economic and environmental differences [40]. To test whether urban-rural typology would affect the finding, we tested the interaction effects between the urban-rural dummy variable and educational achievement, wealth, and the four dimensions of social capital by including the interaction terms of urban-rural dummy variable with each of the other covariate and testing the joint significance of the interaction terms using Wald test. We further stratified the mediation analysis by urban and rural subgroups, to check whether urban-rural residency would affect the mediating effects of the various social capital construct/dimension on education and wealth related

Tal	ole	1		

Measure	of	social	capital.
---------	----	--------	----------

Construct	Dimension	Questions
Structural social capital	Informal interaction	Activity 1, interacted with friends Activity 2, played Ma-jong, played chess, played cards, or went to community club Activity 3, went to a sport, social, or other kind of club Activity 4 took part in a community-related organization
	Altruism	Activity 5, provided help to family, friends, or neighbors who do not live with you Activity 6, done voluntary or charity work Activity 7, cared for a sick or disabled adult who does not live with you
	Formal social participation	Activity 8, attended an educational or training course
Cognitive social capital	Trust	Having relatives or friends (besides spouse/partner) who would be willing and able to help over a long period of time, should the participant need help for basic daily activities such as eating or dressing?

inequalities in depression.

#### 4. Results

#### 4.1. Cohort characteristics

There were 9949 individuals included in the cohorts, among which 43 (0.4 %) participants lacked information for household wealth. Table 2 presented the cohorts' basic characteristics in 2015 and the incidence of depressive symptoms in 2018, stratified by participants' urban-rural residency. There were 57.6 % (5737/9949) participants living in rural areas. Fig. 3 showed that 25.9 % (2579/9948) participants reported depressive symptoms in 2018, including 1697 living in rural areas and 882 living in urban areas, with an incidence rate of 29.6 % and 20.9 %, respectively (p < 0.001).

The cohort presented varied basic characteristics across urban-rural settings. More rural residents (26.7 %) were in the lowest wealth quartile, comparing to a proportion of 13.7 % for their urban counterparts (p < 0.001). And 17.8 % of the participants were illiterate, with the proportion being 11.3 % for the urban residents and 22.6 % for the rural one (p < 0.001). Despite the differences in

#### Table 2

Cohorts' basic characteristics in 2015 and incidence of depressive symptoms in 2018, by urban and rural residency.

n $(n = 2949)$ $(n = 4212)$ $(n = 5737)$ Educational achievement         n $\vartheta$ n $\vartheta$ Educational achievement         1774         17.8         47.8         11.3         29.0         22.6           Primary school and above         4027         40.5         1433         34.0         2594         45.2           Household wealth         1         2010         54.6         187.0         1531         2.6.7           Q1 poorest         2108         2.1.2         57.7         13.7         1531         2.6.7           Q3         2040         2.6.2         1534         36.4         1208         2.1.1           Age groups         -         -         -         -         -         4.5.4         2.03         2.0.4         2.5.2         2.6.0         1.534         36.4         1208         2.1.1           Age groups         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -         -	Basic Characteristics	Overall		Urban		Rural					
n%n%n%Educational achievement177417.87.811.31.2962.2.6Primary school402740.51.43334.0259445.2Secondary school and above14841.720154.6184732.2Housechold wealth*771.3.71.5312.6.7Q1 poorest20821.22.715.31.6002.7.9Q3260426.21.2212.9.01.88324.1Age groups7422.7.61.5433.6.41.0282.1.1Age groups7422.7.61.6033.5.11.9723.4.4Af-5-74 years394339.61.6083.8.22.3354.0.755-64 years394339.61.6083.8.22.3354.0.755-64 years394339.61.6083.8.22.3354.0.755-64 years3.9423.5.32.696.42.554.452years3.030.2.43.5.50.5.01.155.5.0Male5.245.32.0434.8.52.5824.5.0Male9.259.2.739469.3.75.2799.2.0Maried or partnerd90909.1.43.8239.0.85.2679.1.8Single or others8598.63.899.24708.2UBMir Or NCMS*7.968.63.813.32.674.7 <th></th> <th>(n = 9949)</th> <th></th> <th>(n = 4212)</th> <th></th> <th>(n = 5737)</th> <th></th>		(n = 9949)		(n = 4212)		(n = 5737)					
set of the		n	%	n	%	n	%				
Illicatic17,417,847811,3129622.6Primary school and above414841.7230154.6184732.2Bouschol weath'1230154.6184732.2Houseath'1230154.6184732.2Q1 poorsi20842.1.257713.715312.6.7Q2245224.685220.216002.7.9Q326042.6.215.3436.412082.1.1Q4 richest274227.615.3436.412082.1.1Age groups	Educational achievement										
Primary school402740.5143334.0259445.2Secondary school and above414841.72054.6154.732.2Household weath* </td <td>Illiterate</td> <td>1774</td> <td>17.8</td> <td>478</td> <td>11.3</td> <td>1296</td> <td>22.6</td>	Illiterate	1774	17.8	478	11.3	1296	22.6				
secondary school and above1481,72015,61,671,322,22Household weath''71,371,5312,6,7Q2245224,685220,216002,79Q326042,6215343,6,412082,11Q4 reference26042,6215343,6,412082,11Q4 reference26042,6215343,6,412082,11Q4 reference200320,4155420,31,9723,4465-74 years34523,1714803,5119723,4465-74 years304320,22,048,522,031,172,0575 years and above3042,048,522,031,172,0575 years and above3042,048,522,031,172,0575 years and above3023,133,155,033,155,0075 reference752,192,133,132,172,1375 years and above32,243,133,132,172,1375 years and above32,243,133,132,173,138,161,193,132,173,133,132,173,139,169,101,143,132,173,133,132,173,133,133,133,133,133,133,133,133,133,133,133,133,133,133,133	Primary school	4027	40.5	1433	34.0	2594	45.2				
Honoreshi di ventiniQ1 pororest210821225771.3.71.5.312.6.7Q2245224.685220.216002.7.9Q320402.6.21.212.9.01.3.832.4.1Q4 richest27422.6.21.212.9.01.3.832.4.1Age groups1.5.33.6.41.2082.1.1Age structure1.6083.5.11.9723.4.155-64 years3.4523.4.71.4803.5.11.9723.4.165-74 years20302.0.48.5.52.0.31.1752.0.575 years and above5.245.3.52.0.31.5.53.1555.0.175 years and above5.245.3.52.1695.1.53.1555.0.1Male5.247.32.666.34.588.0.15.0.1Marie di partnerdt7247.32.666.34.569.0.85.16Maried or partnered9.0909.1.43.8239.0.85.2679.1.8Single or others8.502.513.549.0.85.679.1.8Maried or partnered9.9909.1.43.8239.0.85.2679.1.8Maried or partnered9.9909.1.43.8239.0.85.2679.1.8Maried or partnered9.9909.1.43.8239.0.85.3129.6.1Maried fuer1.14.81.19.71	Secondary school and above	4148	41.7	2301	54.6	1847	32.2				
Q1 porest210821.27713.715.3126.7Q2245224.685220.2160027.9Q3260426.2122129.0138324.1Q4 richtest27.227.6153436.4120821.1Age group160838.2233540.735-64 years345234.7148035.1197234.465-74 years203020.485520.3117520.575 years and above525.32696.42554.465-74 years203020.485520.3115520.575 years and above525.3216951.525824.5.0Male35245.5204348.525824.5.0Male35245.32666.34588.0Han major2259.7394637.32666.34588.0Single or others8598.039.99.24708.29.1Marited or partnered90991.4382390.852.6791.8Single or others8598.0326326.553.99.09.1URMIn NRCMS'79868.0326326.253.12.679.1URMIn NRCMS'79868.0326326.253.12.679.1No52053.12.489.09.1 </td <td>Household wealth<sup>a</sup></td> <td></td> <td></td> <td></td> <td></td> <td></td> <td></td>	Household wealth <sup>a</sup>										
Q2245224.68520.2160027.9Q3260426.2122129.0138324.1Q4 richest274227.6153436.4120821.1Age groups </td <td>Q1 poorest</td> <td>2108</td> <td>21.2</td> <td>577</td> <td>13.7</td> <td>1531</td> <td>26.7</td>	Q1 poorest	2108	21.2	577	13.7	1531	26.7				
Q326426.212129.0138324.1Q4 richs27.6153.436.4120821.1Age groups45-54 years345239.616838.2233540.755-64 years345234.7148035.1197234.465-74 years203020.485520.3117520.575 years and above52.45.320485520.3117520.575 years and above53245.3204348.5258245.0SterEmale462546.5204348.5258245.0Maie532453.5204348.5258245.0SterierEmale462546.5204348.5258245.0Ban major92292.7394693.7257992.7934693.792.791.0Ban major92292.7394693.792.791.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.091.0	Q2	2452	24.6	852	20.2	1600	27.9				
Q4 richest274227.6153436.4120821.1Age groups45-54 years394339.6160835.1197234.465-74 years345234.7148035.1197234.565-74 years303020.485520.317531531575 years and above5245.32696.42554.4SecFemale462546.5214951.5315555.0Handa53253.5216951.5315555.0EthnicityOther minorities747.326636.345.88.0Hamajor922592.7394693.7527992.0Married or partnered90991.4382390.8526791.3Single or others8598.639.282.635.113.32674.7UEBMI or NRCMS <sup>6</sup> 92.693.725.315.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.0 <th< td=""><td>Q3</td><td>2604</td><td>26.2</td><td>1221</td><td>29.0</td><td>1383</td><td>24.1</td></th<>	Q3	2604	26.2	1221	29.0	1383	24.1				
<table-container>Age croups45-54 years394339.6108038.2233547.755-64 years342034.7148035.1197234.865-74 years203020.485520.3117520.575 years and above5320.313525.74.8Series and above5320.34.85325.84.8Series and above53.521.695.321.6931.555.50Bernale46.524.6321.693.1525.208.0Marine of partnered partnered partnered92.232.66.345.88.0Marine of partnered partnered91.438.2392.03.679.29.2UBMM or NRCMS<sup>6</sup>85.98.638.99.247.08.2UBM or NRCMS<sup>6</sup>79868.0326326.25535.49.39.3Other insurances10.114481.19.11.01.01.01.0No92.492.693.49.39.27.41.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.01.0&lt;</table-container>	Q4 richest	2742	27.6	1534	36.4	1208	21.1				
45-54 years         3943         39.6         108         38.2         2335         40.7           55-64 years         3452         34.7         1480         35.1         1972         34.4           65-74 years         2030         20.4         855         20.3         1175         20.5           75 years and above         524         5.3         269         6.4         255         4.4           Sex           86.5         218.2         25.5         3.5           Male         5324         53.5         2043         48.5         2582         45.0           Mark         5324         53.5         2169         51.5         3155         55.0           Han major         9225         92.7         3946         93.7         2527         92.0           Marited or partnered         9090         91.4         3823         90.8         5267         91.8           Single or others         859         8.6         389         9.2         470         82.3           UEBM <sup>TA</sup> 786         80.3         2632         62.5         5354         93.6           UEBM <sup>TA</sup> 1586         15.9         1	Age groups										
55-64 years345234.7148035.1197234.465-74 years203020.485020.3117520.575 years and above5245.32696.42554.4Sex </td <td>45–54 years</td> <td>3943</td> <td>39.6</td> <td>1608</td> <td>38.2</td> <td>2335</td> <td>40.7</td>	45–54 years	3943	39.6	1608	38.2	2335	40.7				
65-74 years203020.485520.3117520.575 years and above5245.32.696.42554.4SexFemale462546.52.04348.52.58245.0Male53245.352.16951.53.1555.26745.0Dther minorities7247.32.666.345.88.0Han major9259.273.9469.3.752799.25Maritel or partnered90909.1.43.8239.0.85.2679.1.8Maritel or partnered8.98.63.899.24708.2Health insurance overage7.22.72.135.15.2679.1.8URBM or NRCMS7.9868.02.6326.2.55.3549.3.1Other insurances1.051.14.81.1571.0Without insurances1.051.14.81.1571.0Vithout insurances9.0949.1.437828.9.85.3129.2.6Yes8.63.312.2445.3.330.365.29Yes4.6694.6.91.9684.6.72.7014.71No5.2673.3.12.2445.3.330.365.29Yes8.63.312.2445.3.330.365.29Yes4.693.512.463.3.330.365.29Yes6.603.312.2445	55-64 years	3452	34.7	1480	35.1	1972	34.4				
75 years and above5245.32696.42554.4sexFemale462546.5204348.5258245.0Male532453.5216951.5315555.0EthnicityOther minorities7447.32666.335.7527992.0Married or partnered922592.7394693.7527992.0Married or partnered8598.63899.24708.2Bigle or others8598.63899.24708.2UEBMI*15.931.32674.7URMI or NRCMS*798680.326326.25535493.3Other insurances2722.72135.1591.0Without insurances2722.72135.1591.0Without insurances2722.72135.129.29.2Other insurances90949.14378289.85.129.29.2No5.805.312445.3330365.20Yeas4.691.001.332.701.632.72Insurance or insurance1.11.337.93.512.2Ao5.805.312.445.3330365.20Yeas2.625.355.402.3185.503.0575.33Insurance2.635.402.3185.503.057 <td>65–74 years</td> <td>2030</td> <td>20.4</td> <td>855</td> <td>20.3</td> <td>1175</td> <td>20.5</td>	65–74 years	2030	20.4	855	20.3	1175	20.5				
SecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSecSec <th colspan="4" sec<="" t<="" td=""><td>75 years and above</td><td>524</td><td>5.3</td><td>269</td><td>6.4</td><td>255</td><td>4.4</td></th>	<td>75 years and above</td> <td>524</td> <td>5.3</td> <td>269</td> <td>6.4</td> <td>255</td> <td>4.4</td>				75 years and above	524	5.3	269	6.4	255	4.4
Fenale462546.5204348.5258245.0Male532453.5216951.551.551.551.5Etnicity747.32666.345.88.0Han najor92592.7394693.7526791.8Marital statusuMaried or partnered909091.4382.390.8526791.8Single or others8598.638992.247082.3UEBMI°158615.9131.931.32674.7URBMI or NRCMS°798680.3263262.5535493.3Other insurances1051.1481.1571.0Without insurances1051.1378289.8531292.6Vers8528.3124453.330.652.7No528053.124453.330.652.7Vers28053.124453.330.652.7No528053.124453.330.652.7Vers528053.124.453.330.652.7Vers128353.124.453.330.652.7No537554.023.855.030.5753.31128312813.3721281128312.95113.37212.91128312.951 <t< td=""><td>Sex</td><td></td><td></td><td></td><td></td><td></td><td></td></t<>	Sex										
Male532453.5216951.5315555.0EthnicityOther minorities7247.32666.34588.0Han major92592.739469.3.7527992.0Marited statusMarried or partnered909091.4382.390.8526791.8Single or others8598.29.247082Health insurance coverage015.9131913.32674.7URBMI or NRCMS78680.3263262.5536493.3Other insurances7222.72135.1591.0Without insurances1051.1481.1571.0Smote01.4378289.853122.72.7No52805.312445.3330362.55.1Pres4694.9196845.330362.53.1No52805.312445.3330362.53.1Parting alcohol2.71.002.3185.503.0375.333.032.21.2I05.552.562.562.563.512.53.512.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.53.5 <t< td=""><td>Female</td><td>4625</td><td>46.5</td><td>2043</td><td>48.5</td><td>2582</td><td>45.0</td></t<>	Female	4625	46.5	2043	48.5	2582	45.0				
EthnicityOther minorities7247.32666.34588.0Itan major92592.739606.34588.0Martiel status382390.8526791.8Single or others909091.4382390.8526791.8Burlied coverage8.63899.24708.2UEBMI°158615.9131931.326747.1URBMI or NRCMS°798680.3263262.5535493.3Other insurances2722.72135.1591.0Without insurances1051.1378289.8531292.6Yes909491.4378289.8531292.6Yes46991.4378289.8531292.6Yes20053.1224453.3303652.9Yes46946.9196846.7270147.1O537554.0231855.035.12.22.2222.554.010.02.31.52.2222.554.013.372.015.62.23122.654.02.63.33.05.03.1UBMI°4.62.02.12.63.12.22.6Yes5.05.03.12.33.12.22.6<	Male	5324	53.5	2169	51.5	3155	55.0				
Other minorities7247.32666.34588.0Han major922592.7934693.7527992.0Marital status82390.8526791.8Maried or partnered909091.4823.390.8526791.8Single or others8598.63899.24708.2Health insurance coverage $V$ $V$ $V$ $V$ UBBMI®158615.9131931.3267 $V$ $V$ URBMI or NRCMS®798680.3263262.5535493.3 $O$ $O$ $V$ <	Ethnicity										
Han major92592.7394693.7527992.0Marriel statusMarriel or partnered90091.4382390.8526791.8Single or others85986.03899.247091.8Health insurance coverage15.9131.931.32674.7UEBMT <sup>b</sup> 796.080.3263262.5535493.3Other insurances7222.721351571.0Without insurances1051.1481.1571.0SmokeNo90491.4378289.8531292.6Vers85353.124453.3303652.9Yes46953.124453.3303652.9Vers526355.8100023.7156327.211256325.8100023.7156327.221<28.1	Other minorities	724	7.3	266	6.3	458	8.0				
Married or partnered         9090         91.4         3823         90.8         5267         91.8           Single or others         859         8.6         389         9.2         470         8.2           Healti insurance coverage           1319         31.3         267         4.7           UEBMI <sup>0</sup> 1586         15.9         1319         31.3         267         4.7           URBMI or NRCMS <sup>6</sup> 7986         80.3         2632         62.5         5354         93.3           Other insurances         272         2.7         213         5.1         59         1.0           Without insurances         9094         1.1         48         1.1         57         1.0           Smok           512         92         7.4         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.0         1.	Han major	9225	92.7	3946	93.7	5279	92.0				
Married or partnered909091.4382390.8526791.8Single or others8598.63899.24708.2Health insurance coverage8.28.2UEBMI®158615.931.32679.3Other insurances798680.326326.2.553549.3Other insurances2722.72135.1591.0Without insurances1051.1481.1571.0SmokNo909491.4378289.853129.29.2Yes8558.643010.24257.4No528053.1224453.3303652.9Yes46940.919.845.92.72.7No528053.1224453.3303652.9Yes64.92182.72.72.72.72.7No528053.12.24453.3303652.9Yes54.02.1855.0305753.32.7I2.531.0002.3.71.5632.72.63 and above7.841.956.11.3.37.21.6Jepseverynoms (year 2018)7.47.433307.91.0407.4	Marital status										
Single or others         859         8.6         389         9.2         470         8.2           Health insurance coverage                 8.2           UEBM <sup>1</sup> 1586         15.9         1319         31.3         267         4.7           UBBMI or NRCMS <sup>c</sup> 786         8.3         2632         62.5         5354         93.3           Other insurances         7986         8.3         2632         62.5         5354         93.3           Other insurances         7986         8.3         2632         62.5         5354         93.3           Without insurances         105         1.1         88         1.1         57         1.0           Smoke           9094         91.4         3782         89.8         5312         92.6           Yes         855         8.6         360         10.2         425         7.4           No         9280         5311         2444         53.3         3036         53.1           Yes         4669         21.8         1000         23.7         156.3         27.2         27.2	Married or partnered	9090	91.4	3823	90.8	5267	91.8				
Headfinisurance coverage         Ites insurance coverage         Ites insurance coverage         Ites insurance         Ites insura	Single or others	859	8.6	389	9.2	470	8.2				
UEBMID158615.9131931.32674.7URBMI or NRCMSC798680.3263262.5535493.3Other insurances2722.72135.1591.0Without insurances1051.1481.1571.0SomeNo909491.4378289.8531292.6Yes8558.64.3010.289.8531292.6Drinking alcohol90491.4378289.8531292.6No528053.1224453.3303652.9Yes466946.9196846.7270147.1Number of chronic conditions937554.0231855.0305753.311256325.8100023.7156327.2227.87.956113.372.212.63 and above787.83337.939554.9No737074.1330079.1404070.4	Health insurance coverage										
URBMI or NRCMS <sup>c</sup> 798680.3263262.5535493.3Other insurances2722.72135.1591.0Without insurances1051.1481.1571.0SmokeNo909491.4378289.8531292.6Yes85586.040.010.245.052.0Drinking alcohol528053.1224453.3303652.9Yes466946.9196846.7270147.1O537554.0231855.0305753.31256325.8100023.7156327.221212.812.956113.372.212.63 and above78.078.079.1404070.4	UEBMI <sup>b</sup>	1586	15.9	1319	31.3	267	4.7				
Other insurances2722.72135.1591.0Without insurances1051.1481.1571.0SmokeNo909491.4378289.8531292.6Yes8558.630010.2 $2512$ 92.6Drinking alcoholVNo528053.1224453.3303652.9Yes466946.9196846.7270147.1Number of chronic conditions0537554.0231855.0305753.31256325.8100023.7156327.2212.8312.956113.372.212.63 and above72.073.074.1330079.1404070.4	URBMI or NRCMS <sup>c</sup>	7986	80.3	2632	62.5	5354	93.3				
Without insurances1051.1481.1571.0SmokeNo09491.4378289.8531292.6Yes85586.030010.292.692.6Drinking alcoholNo528053.124453.3303652.9Yes46946.9196846.7201147.1Number of chronic conditions0537554.0231855.0305753.31256325.8100023.7156327.22128312.956113.372.012.63 and above783079.1404070.4	Other insurances	272	2.7	213	5.1	59	1.0				
Smoke         No         9094         91.4         3782         89.8         5312         92.6           Yes         855         8.6         430         10.2         425         7.4           Drinking alcohol              7.4           No         5280         53.1         2244         53.3         3036         52.9           Yes         4669         46.9         1968         46.7         2701         471           Number of chronic conditions           1968         46.7         2701         53.3           1         2563         54.0         2318         55.0         3057         53.3           1         2563         25.8         1000         23.7         1563         27.2           2         1283         12.9         561         13.3         722         12.6           3 and above         728         7.3         333         7.91         4040         7.4	Without insurances	105	1.1	48	1.1	57	1.0				
No909491.4378289.8531292.6Yes8558.643010.24257.4Drinking alcoholNo528053.1224453.3303652.9Yes46046.9196846.72701451Number of chronic conditions31855.0305753.31256325.8100023.7156327.22128312.956113.372212.63 and above128312.956113.372.212.6Depressive symptoms (year 2018)No737074.1333079.1404070.4	Smoke										
Yes         855         8.6         430         10.2         425         7.4           Drinking alcohol         5280         53.1         2244         53.3         3036         52.9           Yes         4669         46.9         1968         46.7         2701         47.1           Number of chronic conditions         540         54.0         2318         55.0         3057         53.3           1         2653         54.0         2318         55.0         3057         53.3           2         2         1283         12.9         561         13.3         722         12.6           3 and above         728         7370         74.1         3330         79.1         4040         70.4	No	9094	91.4	3782	89.8	5312	92.6				
Drinking alcohol         State	Yes	855	8.6	430	10.2	425	7.4				
No         5280         53.1         2244         53.3         3036         52.9           Yes         4669         46.9         1968         46.7         2701         47.1           Number of chronic conditions         5375         54.0         2318         55.0         3057         53.3           1         2563         25.8         1000         23.7         1563         27.2           2         1283         12.9         561         13.3         722         12.6           3 and above         728         7.3         333         7.9         9.5         12.5           No         7370         74.1         3330         79.1         4040         70.4	Drinking alcohol										
Yes         4669         46.9         1968         46.7         2701         47.1           Number of chronic conditions         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5         5 <th5< td=""><td>No</td><td>5280</td><td>53.1</td><td>2244</td><td>53.3</td><td>3036</td><td>52.9</td></th5<>	No	5280	53.1	2244	53.3	3036	52.9				
Number of chronic conditions         Sinther of chronic conditions         Sin	Yes	4669	46.9	1968	46.7	2701	47.1				
0         5375         54.0         2318         55.0         3057         53.3           1         2563         25.8         1000         23.7         1563         27.2           2         1283         12.9         561         13.3         722         12.6           3 and above         728         7.3         333         7.9         395         6.9           Depressive symptoms (year 2018)           No         7370         74.1         3330         79.1         4040         70.4	Number of chronic conditions										
1256325.8100023.7156327.22128312.956113.372212.63 and above7287.33337.93956.9Depressive symptoms (year 2018)No737074.1333079.1404070.4	0	5375	54.0	2318	55.0	3057	53.3				
2       1283       12.9       561       13.3       722       12.6         3 and above       728       7.3       333       7.9       395       6.9         Depressive symptoms (year 2018)         No       7370       74.1       3330       79.1       4040       70.4	1	2563	25.8	1000	23.7	1563	27.2				
3 and above     728     7.3     333     7.9     395     6.9       Depressive symptoms (year 2018)       No     7370     74.1     3330     79.1     4040     70.4	2	1283	12.9	561	13.3	722	12.6				
Depressive symptoms (year 2018)         7370         74.1         3330         79.1         4040         70.4	3 and above	728	7.3	333	7.9	395	6.9				
No 7370 74.1 3330 79.1 4040 70.4	Depressive symptoms (year 2018)										
	No	7370	74.1	3330	79.1	4040	70.4				
Yes 2579 25.9 882 20.9 1697 29.6	Yes	2579	25.9	882	20.9	1697	29.6				

Note.

p-Value reports Chi-square tests for all categorical variables.

<sup>a</sup> There were 43 (0.4 %) participants missing information for household wealth.

<sup>b</sup> UEBMI: Urban Employee Basic Medical Insurance.

<sup>c</sup> URBMI or NRCMS: Urban Resident Basic Medical Insurance or New Rural Cooperative Medical Scheme.



Fig. 3. Prevalence of depressive symptoms by urban-rural residency, educational achievement and wealth.

socioeconomic differentials, urban-rural disparities were also found for age, sex, ethnicity, health insurance coverage, smoking, and number of chronic conditions (p < 0.001).

#### 4.2. Stock of social capital

Table 3 described the stock of social capital in four distinct dimensions, stratified by urban-rural settings. The descriptive results by wealth and educational achievement showed similar patterns (Supplementary Tables 1 and 2). A total of 5516 (55.4 %) participants possessed structural social capital, i.e., participating in at least one listed social activity, whilst 6729 (67.6 %) participants possessed social trust, i.e., cognitive social capital. Within the construct of structural social capital, informal interaction (51.0 %) was more common than formal social participation (1.3 %). Comparing to the urban residents (60.5 %), rural residents (51.8 %) reported lower stock of structural social capital (Fig. 4a), with informal interaction activity taking the largest urban to rural difference (10.5 %, p < 0.001). However, no evidence was found for urban-rural differences in the construct of cognitive social capital (Fig. 4b).

#### 4.3. The association between SES, social capital and depressive symptoms

Table 4 showed the factors associated with depressive symptoms, focusing on the role of SES and social capital. All SES indicators (urban-rural residency, educational achievement and household wealth) were negatively associated with the incidence of depressive symptom in the follow-up period, taking similar scales across differentials. For example, urban residents had 37 % lower odds (OR = 0.63, 95%CI: 0.57 to 0.69) of experiencing depressive symptoms than the rural residents; people with higher education had lower odds than the illiterate (OR = 0.45, 95%CI: 0.40 to 0.51), and the richest group had 35 % lower odds than the poorest group (OR = 0.65, 95%CI: 0.57 to 0.73). Socioeconomic inequalities in the incidence of depressive symptoms persisted in the adjusted model. For example, the urban to rural aOR was 0.69 (95%CI: 0.62 to 0.76) after adjusting for age, marital status, health insurance coverage, smoking, drinking alcohol and number of chronic conditions.

Both structural and cognitive social capital in 2015 were negatively associated with the incidence of depressive symptoms in the follow-up period in the adjusted analysis (Table 4). The effects of cognitive social capital were larger (aOR = 0.71, 95%CI: 0.64 to 0.79) than structural social capital (aOR = 0.87, 95%CI: 0.79 to 0.95). Within the structural social capital construct, only the effects of

Table	3
-------	---

Social capital	Overall (n = 9949)		Urban (n =	Urban (n = 4212)		Rural (n = 5737)	
	n	%	n	%	n	%	
Structural social capital	5516	55.4	2546	60.5	2970	51.8	< 0.001
Informal interaction Altruism	5070 2006	51.0 20.2	2401 885	57.0 21.0	2669 1121	46.5 19.5	<0.001 0.071
Formal social participation	127	1.3	86	2.0	41	0.7	< 0.001
Cognitive social capital (Trust)	6729	67.6	2815	66.8	3914	68.2	0.898

Note.

We defined a participant having each construct/dimension of social capital should he/she engaged in at least one activity within this construct/ dimension.

p-Value reports Chi-square tests for all categorical variables.



a Stock of structural social capital by urban-rural residency, educational achievement and wealth



b Stock of cognitive social capital by urban-rural residency, educational achievement and

#### wealth

Fig. 4. (a) Stock of structural social capital by urban-rural residency, educational achievement and wealth (b) Stock of cognitive social capital by urban-rural residency, educational achievement and wealth.

informal interaction persisted in the adjusted analysis (aOR = 0.84, 95%CI: 0.76 to 0.92).

#### 4.4. Mediation analysis

Table 5 showed the results of the mediation analysis, reporting the total, direct and indirect effects, and the proportion of total effects mediated, for each social capital dimension, along the pathway how each SES differential takes an impact. Only informal interaction within the construct of structural social capital showed evidence of mediating effects in explaining the socioeconomic inequalities in the incidence of depressive symptoms. Informal interaction could explain 6.17 % of the urban-rural differences (with a direct effect of 0.425 and an indirect effect of 0.028), 11.24 % of education related inequalities and 11.78 % of wealth related inequalities in depressive symptoms. However, cognitive social capital showed no evidence in mediating the socioeconomic inequalities in depressive symptoms either for urban-rural residency, educational achievement or household wealth related inequalities (p = 0.455, 0.362, and 0.321).

#### 4.5. Subgroup analysis

Results of Wald test showed that all the interaction effects between the urban-rural dummy variable and educational achievement, wealth, and the four dimensions of social capital were not significant (Supplementary Table 4). The subgroup analysis that compared the mediating effects of social capital to education and wealth related inequalities in depressive symptoms by urban-rural settings reported similar findings (Supplementary Tables 5 and 6).

#### Table 4

Associations between socioeconomic status, social capital and depressive symptoms.

Independent variables	Unadjusted mo	del		Adjusted model			
	OR	95 % CI	<i>p</i> -Value	aOR	95 % CI	<i>p</i> -Value	
Socioeconomic status							
Urban-rural residency							
Rural	ref.			ref.			
Urban	0.63	(0.57, 0.69)	< 0.001	0.69	(0.62, 0.76)	< 0.001	
Educational achievement							
Illiterate	ref.			ref.			
Primary school	0.74	(0.66, 0.84)	< 0.001	0.85	(0.75, 0.97)	0.013	
Secondary school and above	0.45	(0.40, 0.51)	< 0.001	0.58	(0.51, 0.67)	< 0.001	
Household wealth							
Q1 poorest	ref.			ref.			
Q2	0.86	(0.75, 0.97)	0.018	0.89	(0.78, 1.01)	0.081	
Q3	0.66	(0.58, 0.75)	< 0.001	0.72	(0.63, 0.83)	< 0.001	
Q4 richest	0.65	(0.57, 0.73)	< 0.001	0.75	(0.65, 0.85)	< 0.001	
Social capital							
Structural social capital							
Yes	0.83	(0.76, 0.91)	<0.001	0.87	(0.79, 0.95)	0.002	
No	ref.			ref.			
Informal interaction							
Yes	0.81	(0.74,0.89)	< 0.001	0.84	(0.76,0.92)	< 0.001	
No	ref.			ref.			
Altruism							
Yes	0.88	(0.78,0.98)	0.023	0.96	(0.85,1.08)	0.475	
No	ref.			ref.			
Formal participation							
Yes	0.50	(0.31,0.81)	0.005	0.64	(0.39,1.04)	0.072	
No	ref.			ref.	. , ,		
Cognitive social capital (Trust)							
Yes	0.73	(0.66,0.80)	<0.001	0.71	(0.64, 0.79)	< 0.001	
No	ref.			ref.			

Note.

Logistic regressions were performed to report on the odds ratios (ORs). The adjusted model reported aORs that additionally control for age, marital status, health insurance, smoking, drinking alcohol and number of chronic conditions.

#### 5. Discussion

Using the data from CHARLS, we explored the longitudinal effects of socioeconomic status and social capital on the incidence of depressive symptoms. We reported persistent socioeconomic inequalities in depression, irrespective of how socioeconomic status were measured. We measured social capital from two constructs and four distinct dimensions: informal interaction, altruism, formal social participation, and trust. We found that both cognitive social capital and structural social capital were associated with lower incidence of depressive symptoms, where informal interaction took the largest effect within the construct of structural social capital. The mediation analysis further illustrated that informal interaction took the single most important medicating effect that explains 6%–12% of the inequalities in depressive symptoms related to each socioeconomic determinant.

The findings of persistent education and wealth related inequalities in depression corroborated data from other countries, and China as well [32][41–45]. Based on a cohort design, our data also confirmed the protective role of cognitive social capital for depression. We contributed to the literature by specify three distinct dimensions of structural social capital, where we found that informal interaction took larger effects than altruism and formal social participation. Such findings were in line with prior data on the role of social participation activities such as mahjong and physical exercise in protecting mental wellbeing [46]. We only found two previous research that quantified the mediating effects of social capital in explaining socioeconomic inequalities in depression. Xin and Ren found that cognitive and structural social capital could explain 14.3 % and 3.8 % of the educational related inequalities in depression amongst Chinese older people [27]. Whilst data from Korean older people showed that cognitive social capital, measured as reciprocity, could explain about 10.2 % of the wealth related inequalities in depression [32]. These two studies were all based on cross-sectional data. We therefore enriched the literature by providing evidence from cohort design and ascertaining more dimensions of structural social capital.

Windsor's study indicated that low socioeconomic status, such as poor education or income, may undermine community cohesion, social trust and reciprocity, leading to reduced resistance to stress and eventually affecting mental wellbeing [31]. However, we did not find evidence of socioeconomic differences in cognitive social capital, which helped to explain why trust offers no mediating effects in explaining the socioeconomic inequalities in depression in our case. Importantly, our data showed strong protecting effects of structural social capital for depression, with the dimension of informal interaction taking the single most important role. Whenever using urban-rural residency, household wealth or educational achievement as the SES indicator, we found that informal interaction mediates about 6%–12 % of the socioeconomic inequalities in depressive symptoms. We found that only 1.3 % participants were engaged in formal social participation, in line with the literature that formal social participations and institutions played a weak role in

#### Table 5

Mediation analysis for the role of each dimension of social capital on socioeconomic inequalities in the incidence of depressive symptoms.

Social participation	ation Total effect Direct effect		Indirect effect				Mediated			
	β1	95 % CI	<i>p</i> - Value	β <sub>2</sub>	95 % CI	<i>p</i> - Value	$\beta_3$	95 % CI	<i>p</i> - Value	(%)
Urban-rural residency										
Structural Social capital	0.452	(0.355,0.550)	< 0.001	0.430	(0.333,0.528)	< 0.001	0.022	(0.011,0.033)	< 0.001	4.90
Informal interaction	0.453	(0.356,0.550)	< 0.001	0.425	(0.327,0.523)	< 0.001	0.028	(0.015,0.041)	< 0.001	6.17
Altruism	0.451	(0.354,0.549)	< 0.001	0.450	(0.353,0.547)	< 0.001	0.001	(-0.001,0.004)	0.303	0.33
Formal social	0.452	(0.355,0.549)	< 0.001	0.447	(0.350,0.544)	< 0.001	0.005	(-0.001,0.012)	0.092	1.20
participation	0.455	(0.254.0.556)	<0.001	0.454	(0.252.0.555)	<0.001	0.001	(0.005.0.007)	0.755	0.22
(Trust)	0.455	(0.334,0.330)	<0.001	0.454	(0.333,0.333)	<0.001	0.001	(-0.003,0.007)	0.755	0.22
Educational achievement										
Structural Social capital	0.359	(0.240,0.479)	< 0.001	0.326	(0.206,0.447)	< 0.001	0.033	(0.017,0.049)	< 0.001	9.17
Informal interaction	0.360	(0.240,0.479)	< 0.001	0.319	(0.199,0.440)	< 0.001	0.040	(0.023,0.058)	< 0.001	11.24
Altruism	0.359	(0.239,0.477)	< 0.001	0.355	(0.235,0.474)	< 0.001	0.003	(-0.002,0.009)	0.274	0.88
Formal social	0.358	(0.239,0.478)	< 0.001	0.355	(0.235,0.474)	< 0.001	0.003	(-0.000,0.007)	0.073	0.92
participation										
Cognitive social capital	0.362	(0.239,0.485)	< 0.001	0.358	(0.235,0.481)	< 0.001	0.004	(-0.004,0.012)	0.362	1.07
(Trust)										
Household wealth										
Structural Social capital	0.300	(0.190,0.408)	< 0.001	0.267	(0.158,0.377)	< 0.001	0.031	(0.017,0.046)	< 0.001	10.53
Informal interaction	0.299	(0.190,0.374)	< 0.001	0.264	(0.155,0.374)	< 0.001	0.035	(0.020,0.050)	< 0.001	11.78
Altruism	0.297	(0.189,0.406)	< 0.001	0.294	(0.185,0.403)	< 0.001	0.004	(-0.004,0.012)	0.341	1.27
Formal social	0.298	(0.186,0.407)	< 0.001	0.294	(0.186,0.403)	< 0.001	0.003	(-0.000,0.007)	0.079	1.13
participation										
Cognitive social capital	0.321	(0.225,0.418)	< 0.001	0.317	(0.220,0.414)	< 0.001	0.004	(-0.002,0.011)	0.164	1.36
(Trust)										

Note: KHB methods were performed to decompose the total effects of each SES on incidence of depressive symptoms into direct effects and indirect effects. Logistic regressions were performed which additionally adjusted for age, marital status, health insurance, smoking, drinking alcohol and number of chronic conditions. Indirect effects reflect the contribution of each social capita dimension to the socioeconomic inequalities related to urban-rural residence, educational achievement, and wealth, respectively. We define each SES differential as a dichotomous variable, i.e. rural vs urban, illiterate vs non-illiterate, and poor vs rich (Q1-Q2 vs Q3-Q4).

China's social structure [47]. In fact, informal social participation, such as playing Ma-jong or square dance was observed to be more common among older Chinese [46]. These informal interactions provided additional opportunities for social support, social networking, and social participation. Social support from peers might buffer the negative effects of stress [48,49]. Social networking positively influenced older people [50] by giving them opportunity to fulfill their personal expectations, increase self-esteem and sense of purpose, and dampen stress-related neuroendocrine responses [30]. While social participation might help strengthen the social ties between older people to obtain more economic or cultural resources [51]. To improve equity of mental health, our finding therefore suggested that the government should create opportunities for residents to organize or engage in social activities on their own by strengthening activity planning, venue support, and other supportive measures. Social networking positively influenced older people to obtain more economic or cultural resources [51]. To improve equity of mental health, our finding therefore suggested neuroendocrine responses [30]. While social participation might help strengthen the opportunity to fulfill their personal expectations, increased self-esteem and sense of purpose, and dampened stress-related neuroendocrine responses [30]. While social participation might help strengthen the social ties between older people to obtain more economic or cultural resources [51]. To improve equity of mental health, our finding therefore suggested that the government should create opportunities for residents to organize or engage in social activities on their own by strengthening activity planning, venue support, and other supportive measures. For example, strengthening public transportation construction can facilitate residents' travel and help them establish close connections with family and friends.

A recent systematic review on urban-rural disparities for depression reported conflicting findings in different settings across countries among the older people [12]. In the developed world, depressive disorders were more prevailed in the urban areas, especially in large cities; whilst the developing world presented reverse epidemic patterns in general [14]. The higher incidence of depressive symptoms in rural China, as we reported, corroborate data from the other developing countries, and China as well [40,46]. In developed countries, rural residents also had access to mental health care services. However, urban residents were more susceptible to sleep disruption due to excessive exposure to artificial light at night, thereby increasing the risk of depression [12]. However, Enormous urban-rural disparities existed in China's health systems, policies and environment [52–54]. Weaker infrastructure and less convenient traffic disproportionately affected the rural areas [12,55]. Mental health services concentrated in urban areas, health insurance is more generous to the urban residents, the rural vulnerable are therefore less likely to receive adequate care when in needs [56,57]. In addition, evidence was emerging that many rural older Chinese were suffering empty-nests as urbanization rapidly processes, which also undermined rural elder's opportunities for social connection. As our data showed, 49 % rural older people were less likely to participate in four informal social activities. Indeed, the stronger association between urban-rural residency and incidence of depressive symptoms, as compare to the effects of household wealth and educational achievement, suggested that urban-rural division might be a more fundamental social determinant of health in China's specific case.

#### 5.1. Strength and limitations

This study was based on a cohort design, providing solid evidence and also informing policy interventions. We enriched the measurement of structural social capital and highlight the role of informal interaction that explains the socioeconomic gradients in depression. However, there were several important limitations. First, 2073 (9.8 %) participants lost to follow up for the CES-D. Compared to the 9949 cohorts included, more of these participants lived in urban areas (19.2 % higher, p < 0.001) and are poorer (7.1 % higher, p < 0.001) (Supplementary Table 3), which may lead to the underestimation of the incidence of depressive symptoms in our case. Therefore, it needed to be cautious when interpreting our results across the country. Second, we excluded 4974 (23.6 %) participants who had already recorded depressive symptoms in 2015, which would also lower the rates of depression and may lead to an overestimation of the social capital effect on depressive symptoms. However, we reported a incidence of 25.9 %, higher than the results reported by Tang et al. in their 2021 review [7]. These findings were consistent with previous data that the CHARLS tend to report higher prevalence of depressive symptoms using CES-D [7]. Third, questions to define cognitive social capital were limited in CHARLS. There were previous studies using trust in parents, friends, and strangers to define trust, as well as a common definition of cognitive social capital based on feelings of belonging, trust, and willingness to provided help to others. However, due to the limitation in CHARLS, our research can only select trust measured by one question to define cognitive social capital. Future work is warranted to better explore the definition of trust and cognitive social capital. Besides, the CES-D 10 scale we used had limitations in defining depression, as it is only a screening scale and has no diagnostic value.

#### 6. Conclusion

Our data highlighted the unsatisfied mental wellbeing of the vulnerable older people with low income, low educational achievement, and who lives in the rural areas. The single important protecting effect of informal interaction suggested a deliberate thinking on how to make change. Older people seemed rely more on personal interactions instead of formal participations, collective activities such as training, course, clubs and sports are more likely to track attention and helped the seniors to increase their interaction with partners and reduce loneliness, especially for those with relatively low socioeconomic status. To promote the Health in All Policies, government and social organizations should consider more on how to intervene on the environment and create more opportunities that encourage the older people to connect and better able to integrate into the community [58].

#### Data availability statement

CHARLS is a survey project hosted by the National Development Research Institute of Peking University and executed by the China Social Science Survey Center. It is a major project funded by the National Natural Science Foundation of China. CHARLS is publicly available data that can be applied for by anyone in need, with website link attached (http://charls.pku.edu.cn/). However, CHARLS requires the applicant to sign an agreement and not directly share data with the outside world. Any user with a need should directly apply to CHARLS. If the publication requires the provision of raw data, it is available to apply to CHARLS, and meanwhile, we will also provide all code of Stata involved in the data analysis and processing process.

#### Funding

National Natural Science Foundation of China (71761130083).

Ethical approval for all the CHARLS waves was granted by the Institutional Review Board at Peking University (IRB number: IRB00001052-11015). All participants signed a paper informed consent form.

#### CRediT authorship contribution statement

Xixi Fu: Writing – review & editing, Writing – original draft, Visualization, Validation, Software, Project administration, Methodology, Investigation, Formal analysis, Data curation, Conceptualization. Shunzhuang Peng: Methodology, Data curation, Conceptualization. Xing Lin Feng: Writing – review & editing, Supervision, Methodology, Funding acquisition, Conceptualization.

#### Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

#### Acknowledgements

Xixi Fu is Xing Lin Feng's master student, she did this research under Xing Lin Feng's supervision.

#### Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.heliyon.2024.e24918.

#### References

- [1] D. Moreno-Agostino, et al., Global trends in the prevalence and incidence of depression: a systematic review and meta-analysis, Journal of affective disorders 281 (2021) 235–243
- [2] Vos, T., et al., Global burden of 369 diseases and injuries in 204 countries and territories, 1990–2019: a systematic analysis for the Global Burden of Disease Study 2019. The Lancet, 2020. 396(10258): p. 1204-1222.
- O. Liu, et al., Changes in the global burden of depression from 1990 to 2017; Findings from the Global Burden of Disease study, J Psychiatr Res 126 (2020) 134-140. [3] https://doi.org/10.1016/j.jpsychires.2019.08.002.
- [4] W.H. Organization. World mental health report: Transforming mental health for all executive summary, World Health Organization, 2022.
- [5] Y. Huang, et al., Prevalence of mental disorders in China: a cross-sectional epidemiological study, The Lancet Psychiatry 6 (3) (2019) 211-224.
- [6] G. Thornicroft, et al., Undertreatment of people with major depressive disorder in 21 countries, Br J Psychiatry 210 (2) (2017) 119–124, https://doi.org/10.1192/ bip.bp.116.188078.
- [7] T. Tang, J. Jiang, X. Tang, Prevalence of depressive symptoms among older adults in mainland China: A systematic review and meta-analysis, Journal of affective disorders 293 (2021) 379-390.
- [8] X. Lei, et al., Depressive symptoms and SES among the mid-aged and elderly in China: evidence from the China Health and Retirement Longitudinal Study national baseline, Soc Sci Med 120 (2014) 224-232, https://doi.org/10.1016/j.socscimed.2014.09.028
- C. Murata, et al., Association between depression and socio-economic status among community-dwelling elderly in Japan: the Aichi Gerontological Evaluation Study [0] (AGES), Health Place 14 (3) (2008) 406-414, https://doi.org/10.1016/j.healthplace.2007.08.007
- [10] V. Lorant, et al., Socioeconomic inequalities in depression: a meta-analysis, Am J Epidemiol 157 (2) (2003) 98-112, https://doi.org/10.1093/aje/kwf182.
- [11] F. Wang, et al., Association of socioeconomic status and health-related behavior with elderly health in China, PLoS One 13 (9) (2018) e0204237, https://doi.org/ 10.1371/journal.pone 02042
- [12] J. Purtle, et al., Urban-Rural Differences in Older Adult Depression: A Systematic Review and Meta-analysis of Comparative Studies, Am J Prev Med 56 (4) (2019) 603-613, https://doi.org/10.1016/j.amepre.2018.11.008
- [13] X. Dong, M.A. Simon, Health and aging in a Chinese population: urban and rural disparities, Geriatr Gerontol Int 10 (1) (2010) 85–93, https://doi.org/10.1111/ i 1447-0594 2009 00563 x
- [14] L. Zhang, et al., The prevalence of depressive symptoms among the older in China: a meta-analysis, Int J Geriatr Psychiatry 27 (9) (2012) 900–906, https://doi.org/ 10.1002/gps.2821.
- [15] A. Ehsan, et al., Social capital and health: A systematic review of systematic reviews, SSM Popul Health 8 (2019) 100425, https://doi.org/10.1016/j. ssmph 2019 100425
- [16] S. Choi, et al., Association of community level social trust and reciprocity with mortality: a retrospective cohort study, BMC Public Health 20 (1) (2020) 1793, https:// doi org/10.1186/s12889-020-09944-3
- [17] A.M. Ehsan, M.J. De Silva, Social capital and common mental disorder: a systematic review, J Epidemiol Community Health 69 (10) (2015) 1021–1028, https://doi. org/10.1136/jech-2015-205868
- [18] Bourdieu, P., The forms of capital. (1986). Cultural theory: An anthology, 2011. 1: p. 81-93.
- [19] A.K. Forsman, et al., Structural and cognitive social capital and depression among older adults in two Nordic regions, Aging Ment Health 16 (6) (2012) 771–779, https://doi.org/10.1080/13607863.2012.667784.
- [20] C.H. Hsieh, A concept analysis of social capital within a health context, Nurs Forum 43 (3) (2008) 151-159, https://doi.org/10.1111/j.1744-6198.2008.00107.x.
- [21] Nutakor, J.A., et al. Socioeconomic Status and Quality of Life: An Assessment of the Mediating Effect of Social Capital. in Healthcare. 2023. MDPI.
- [22] J. Han, et al., Social capital, socioeconomic status and self-efficacy, Applied Economics and Finance 2 (1) (2015) 1-10.
- [23] M. Rotenberg, K.K. Anderson, K. McKenzie, Social capital and psychosis: A scoping review, Social psychiatry and psychiatric epidemiology 55 (2020) 659-671. [24] H.J. Lee, D.K. Lee, W. Song, Relationships between Social Capital, Social Capital Satisfaction, Self-Esteem, and Depression among Elderly Urban Residents: Analysis of Secondary Survey Data, Int J Environ Res Public Health 16 (8) (2019), https://doi.org/10.3390/ijerph16081445.
- [25] I. Bojorquez-Chapela, et al., Effect of social capital and personal autonomy on the incidence of depressive symptoms in the elderly: evidence from a longitudinal study in Mexico, Aging Ment Health 16 (4) (2012) 462-471, https://doi.org/10.1080/13607863.2011.651432.
- [26] W. Cao, et al., Social capital and depression: evidence from urban elderly in China, Aging Ment Health 19 (5) (2015) 418-429, https://doi.org/10.1080/ 13607863.2014.948805
- [27] Y. Xin, X. Ren, Social Capital as a Mediator through the Effect of Education on Depression and Obesity among the Elderly in China, Int J Environ Res Public Health 17 (11) (2020), https://doi.org/10.3390/ijerph17113977
- S. Van den Broucke, Implementing health in all policies post Helsinki 2013: why, what, who and how. Health Promot Int 28 (3) (2013) 281-284, https://doi. [28] rg/10.1093/heapro/dat050.
- [29] E.P. Uphoff, et al., A systematic review of the relationships between social capital and socioeconomic inequalities in health: a contribution to understanding the psychosocial pathway of health inequalities, Int J Equity Health 12 (2013) 54, https://doi.org/10.1186/1475-9276-12
- P.A. Thoits, Mechanisms linking social ties and support to physical and mental health, J Health Soc Behav 52 (2) (2011) 145–161, https://doi.org/10.1177/ [30] 0022146510395592
- [31] T.D. Windsor, et al., Structural and functional social network attributes moderate the association of self-rated health with mental health in midlife and older adults, Int Psychogeriatr 28 (1) (2016) 49-61, https://doi.org/10.1017/s1041610215001143.
- [32] K.M. Han, et al., Social capital, socioeconomic status, and depression in community-living elderly, J Psychiatr Res 98 (2018) 133-140, https://doi.org/10.1016/j. insychires 2018 01 002
- [33] Y. Zhao, et al., Cohort profile: the China Health and Retirement Longitudinal Study (CHARLS), Int J Epidemiol 43 (1) (2014) 61-68, https://doi.org/10.1093/ije/ dvs203
- [34] E.M. Andresen, et al., Screening for depression in well older adults: evaluation of a short form of the CES-D (Center for Epidemiologic Studies Depression Scale), Am J Prev Med 10 (2) (1994) 77-84.
- [35] Q. Huang, X. Wang, G. Chen, Reliability and validity of 10-item CES-D among middle aged and older adults in China, Chin J Health Psychol 7 (2015) 1036–1041. [36] T.A. Engbers, M.F. Thompson, T.F. Slaper, Theory and Measurement in Social Capital Research, Social Indicators Research 132 (2) (2016) 537-558, https://doi. org/10.1007/s11205-016-1299-0.
- [37] A. Maier, et al., Risk factors and protective factors of depression in older people 65+. A systematic review, PLoS One 16 (5) (2021) e0251326, https://doi.org/ 10.1371/journal.pone.0251326.
- [38] X.L. Feng, Undiagnosed and Uncontrolled Chronic Conditions in China: Could Social Health Insurance Consolidation Make a Change? Med Care Res Rev 75 (4) (2018) 479–515, https://doi.org/10.1177/1077558717690303.
- [39] U. Kohler, K.B. Karlson, A. Holm, Comparing coefficients of nested nonlinear probability models, The Stata Journal 11 (3) (2011) 420-438.
- [40] L.W. Li, et al., Understanding Rural-Urban Differences in Depressive Symptoms Among Older Adults in China, J Aging Health 28 (2) (2016) 341–362, https://doi.org/ 10 1177/0898264315591003
- [41] S. Bauldry, Variation in the protective effect of higher education against depression. Society and mental health 5 (2) (2015) 145–161.
- [42] J. Lee, Pathways from education to depression, Journal of cross-cultural gerontology 26 (2) (2011) 121–135.
- [43] H. Chang-Quan, et al., Education and risk for late life depression: a meta-analysis of published literature, The International Journal of Psychiatry in Medicine 40 (1) (2010) 109-124.
- M. Lotfaliany, et al., Variation in the prevalence of depression and patterns of association, sociodemographic and lifestyle factors in community-dwelling older adults in six [44] low-and middle-income countries, Journal of Affective Disorders 251 (2019) 218-226.
- [45] I. Kawachi, et al., Social capital, income inequality, and mortality, Am J Public Health 87 (9) (1997) 1491–1498, https://doi.org/10.2105/ajph.87.9.1491.

- [46] R. Wang, et al., Melancholy or mahjong? Diversity, frequency, type, and rural-urban divide of social participation and depression in middle- and old-aged Chinese: A fixed-effects analysis, Soc Sci Med 238 (2019) 112518, https://doi.org/10.1016/j.socscimed.2019.112518.
- [47] J. Liu, et al., Social engagement and elderly health in China: evidence from the China health and retirement longitudinal survey (CHARLS), International journal of environmental research and public health 16 (2) (2019) 278.
- [48] S. Cohen, T.A. Wills, Stress, social support, and the buffering hypothesis, Psychological bulletin 98 (2) (1985) 310.
- [49] I. Kawachi, L.F. Berkman, Social ties and mental health, Journal of Urban health 78 (3) (2001) 458-467.
- [50] S.H. Lee, H. Lee, S. Yu, Effectiveness of Social Support for Community-Dwelling Elderly with Depression: A Systematic Review and Meta-Analysis, Healthcare (Basel) 10 (9) (2022), https://doi.org/10.3390/healthcare10091598.
- [51] R. Wang, et al., Relationship between neighbourhood social participation and depression among older adults: A longitudinal study in China, Health & social care in the community 28 (1) (2020) 247–259.
- [52] D. Su, et al., Depression and social support between China' rural and urban empty-nest elderly, Arch Gerontol Geriatr 55 (3) (2012) 564–569, https://doi.org/ 10.1016/j.archger.2012.06.006.
- [53] X. Ma, et al., Prevalence and sociodemographic correlates of depression in an elderly population living with family members in Beijing, China, Psychol Med 38 (12) (2008) 1723–1730, https://doi.org/10.1017/s0033291708003164.
- [54] Feng, L., et al., Burden and correlates of geriatric depression in the Uyghur elderly population, observation from Xinjiang, China. PLoS One, 2014. 9(12): p. e114139. doi.10.1371/journal.pone.0114139.
- [55] Choi, N.G. and D.M. DiNitto, Depressive Symptoms Among Older Adults Who Do Not Drive: Association With Mobility Resources and Perceived Transportation Barriers. Gerontologist, 2016. 56(3): p. 432-43. doi.org/10.1093/geront/gnu116.
- [56] F. Cai, et al.. The elderly and old age support in rural China, World Bank Publications, 2012.
- [57] J. Li, X.L. Feng, Health care-seeking behaviours and health expenditures in adults aged 45 years and older in China, 2011-2013, Trop Med Int Health 22 (5) (2017) 638–654, https://doi.org/10.1111/tmi.12865.
- [58] D.E. Rodrigues, et al., The Influence of Neighborhood Social Capital on Leisure-Time Physical Activity: a Population-Based Study in Brazil, J Urban Health 95 (5) (2018) 727–738, https://doi.org/10.1007/s11524-018-0293-z.