## Effect of neighbourhood deprivation and social cohesion on mental health inequality: a multilevel population-based longitudinal study

## D. Fone<sup>1</sup>\*, J. White<sup>2</sup>, D. Farewell<sup>1</sup>, M. Kelly<sup>3</sup>, G. John<sup>4</sup>, K. Lloyd<sup>5</sup>, G. Williams<sup>6</sup> and F. Dunstan<sup>1</sup>

<sup>1</sup>Institute of Primary Care and Public Health, School of Medicine, Cardiff University, Cardiff, Wales, UK

<sup>2</sup>Centre for the Development and Evaluation of Complex Public Health Interventions, School of Medicine, Cardiff University, Cardiff, Wales, UK

<sup>3</sup>Institute for Translation, Innovation, Methodology and Engagement, School of Medicine, Cardiff University, Cardiff, Wales, UK

<sup>4</sup>NHS Wales Informatics Service, Cardiff, Wales, UK

<sup>5</sup>College of Medicine, Swansea University, Swansea, Wales, UK

<sup>6</sup> School of Social Sciences, Cardiff University, Cardiff, Wales, UK

**Background.** The common mental disorders (CMDs) of anxiety and depression are the most common form of poor mental health in the general population. Evidence from the small number of previous cohort studies on the role of neighbourhood factors in mental health is inconclusive. We tested the hypothesis that high levels of neighbourhood social cohesion modify an adverse association between change in individual mental health and neighbourhood deprivation.

**Method.** We carried out a longitudinal multilevel analysis using data from the Caerphilly Health and Social Needs Cohort Study with a 7-year follow-up (n=4426; age range 18–74 years at baseline). Neighbourhood deprivation and neighbourhood social cohesion were assessed at baseline and change in mental health between follow-up and baseline was assessed using the five-item Mental Health Inventory (MHI-5).

**Results.** Residence in the most deprived neighbourhoods was negatively associated with change in mental health, after adjusting for baseline individual socio-economic risk factors and transitions in life events. This negative effect was significantly reduced in high social cohesion neighbourhoods. The predicted change in mental health score was calculated for the 10th and 90th centiles of the household low-income distribution. The difference between them was –2.8 in the low social cohesion group and 1.1 in the high cohesion group. The difference between the groups was 3.9 [95% confidence interval (CI) 0.2–7.6].

**Conclusions.** The public health burden of poor mental health and mental health inequality could potentially be reduced by strengthening social cohesion in deprived neighbourhoods. This offers a mechanism to address the adverse effect of neighbourhood deprivation on population mental health.

Received 25 February 2013; Revised 3 December 2013; Accepted 13 December 2013; First published online 22 January 2014

Key words: Epidemiology, longitudinal studies, mental health, social environment, social medicine.

#### Introduction

The common mental disorders (CMDs) of anxiety and depression are the most common form of psychological distress in the general population (McManus *et al.* 2009). The prevalence of CMD is reported to vary between 15% and 30%, depending on the case definition used (Weich, 1997), and these high levels of psychological distress constitute a major public health burden of reduced physical and social functioning

(Lloyd et al. 1996), disability (Rai et al. 2012), higher mortality (Russ et al. 2012), lost productivity from sickness absence from work (Croft-Jefferys & Wilkinson, 1989; Dewa et al. 2011) and high levels of welfare benefits (Ford et al. 2010). CMDs are significantly more common in socially disadvantaged populations (Fryers et al. 2003). The social determinants of CMD are well described and include, for example, age, sex and socio-economic status (Weich, 1997; Sacker & Wiggins, 2002; Fryers et al. 2003; Skapinakis, 2007), and transitions in life events such as employment history (Pevalin & Goldberg, 2003; Thomas et al. 2005, 2007), marital status, bereavement (Wade & Pevalin, 2004; LaPierre, 2012) and physical health (Pevalin & Goldberg, 2003). Whether the neighbourhood social environment plays an important role of

<sup>\*</sup> Address for correspondence: D. Fone, Professor of Health Sciences Research, Institute of Primary Care and Public Health, School of Medicine, Cardiff University, Heath Park, Cardiff CF14 4YS, Wales, UK.

<sup>(</sup>Email: foned@cf.ac.uk)

in the aetiology of CMD remains an important research question. Evidence from multilevel analyses of crosssectional datasets has suggested that people living in socio-economically deprived neighbourhoods have higher levels of CMDs than expected from their individual socio-economic risk factors (Fone et al. 2007a,b; Mair et al. 2008). However, significant neighbourhood effects have not been confirmed in longitudinal studies (Propper et al. 2005; Weich et al. 2005; Stafford et al. 2008c). A second neighbourhood factor of interest is the concept of social capital, defined as 'features of social organization, such as trust, norms, and networks that can improve the efficiency of society by facilitating coordinated actions' (Putnam, 1993). The idea of social capital is much debated, with particular difficulties arising for empirical studies from the complexities of meaning, definition and measurement of a multidimensional concept (Muntaner et al. 2000; Fine, 2001). However, a useful model of social capital recognizes two components, structural and cognitive (Harpham et al. 2002). Structural refers to associational links and activity within society, such as civic engagement or 'what people do'. Cognitive refers to people's perceptions of their social relationships, such as interpersonal trust, sharing and reciprocity, or 'what people feel', and is a measurable indicator of social cohesion. In the present study our interest was in the concept of social cohesion, which is conceptualized as a collective community-level characteristic measured by the levels of trust, norms of reciprocity and the formation of strong social bonds within the local social structure (Kawachi & Berkman, 2000; Stafford et al. 2003; Subramanian et al. 2003). There is evidence to suggest that this concept of community-level social cohesion is useful in investigating the determinants of general health status (Kawachi et al. 1999; Subramanian et al. 2002; Stafford et al. 2003).

Similarly to neighbourhood deprivation, there is cross-sectional evidence for an association between higher levels of neighbourhood social cohesion and lower levels of CMD (De Silva *et al.* 2005; Almedom, 2005; Fone *et al.* 2007*b*; Stafford *et al.* 2008*a*), but the longitudinal effect of neighbourhood social cohesion on change in mental health is unknown (Mair *et al.* 2009; Murayama *et al.* 2012).

The joint role of these potentially important contextual social determinants of the CMDs is unknown. It has been hypothesized that high levels of neighbourhood social cohesion could modify an association between neighbourhood deprivation and the mental health of individuals (Cattell, 2001; McKenzie *et al.* 2002). In the present study we tested the hypothesis that high levels of neighbourhood social cohesion modify an adverse association between change in individual mental health and neighbourhood deprivation. We carried out a longitudinal analysis using data from the Caerphilly Health and Social Needs Cohort Study, a community study of health inequality set in Caerphilly county borough, South Wales, UK (Fone et al. 2012). We examined the effect of baseline neighbourhood deprivation and social cohesion on change in mental health, adjusting for baseline socioeconomic status, the occurrence of important life event transitions, and moving between neighbourhoods within the study area. Our specific objectives were to investigate whether change in mental health was associated with (1) baseline neighbourhood deprivation and (2) baseline neighbourhood social cohesion, and (3) whether any association between neighbourhood deprivation and change in mental health was modified by high levels of neighbourhood social cohesion.

## Method

## Participants

We have previously described the Caerphilly Health and Social Needs Cohort Study in detail (Fone et al. 2012). In brief, we carried out a baseline postal questionnaire population survey in 2001, obtaining responses from 10892 (60.6%) participants aged 18-74 years. Sampling was carried out stratified by the 36 electoral wards of local government in the study area (mean population 3600 adults), aiming to achieve an equal number of participants in each ward. Individual records were linked to the 1991 UK Census enumeration district of residence using the address postcode. The enumeration district is a smaller geographically defined area than the electoral ward with an average population of 400 adults. There are 325 enumeration districts in the study area and these were used as the best available representation of neighbourhood in the study. We carried out a followup postal questionnaire survey of the 9551 baseline respondents still alive and resident in the study area 7 years later in 2008, obtaining 4558 completed questionnaires with validated age and sex, representing an adjusted response of 50.2% (Fone et al. 2012). Full ethical approval was obtained for both waves of the study from the South-East Wales Research Ethics Committee.

## Outcome measure: change in mental health

Both waves of the survey included the 36-item Short-Form Health Survey (SF-36) version 2 and we used the 5-item Mental Health Inventory (MHI-5) subscale scores as the measure of CMD (Ware *et al.* 1993, 2000). The MHI-5 has been shown to be a robust measure of mental health in the general adult

population (Berwick *et al.* 1991; McCabe *et al.* 1996; Ware & Gandek, 1998; Kelly *et al.* 2008) and is effective at screening for mood and anxiety disorders assessed using Diagnostic Interview Schedules, but less effective at identifying other disorders such as somatoform disorders, phobias, substance misuse and some anxiety disorders (Berwick *et al.* 1991; Rumpf *et al.* 2001; Cuijpers *et al.* 2009).

The MHI-5 scale comprises five questions asking about symptoms in the past 4 weeks: (1) Have you been very nervous? (2) Have you felt so down in the dumps that nothing could cheer you up? (3) Have you felt calm and peaceful? (4) Have you felt downhearted and depressed? and (5) Have you been happy? Each of the questions has five response categories that score 1=all of the time, 2=most of the time, 3=some of the time, 4=a little of the time, and 5=none of the time. Where a question was not answered, missing data were imputed using the scaledevelopers method as the mean score of question responses, given that at least three of the five questions were answered (Ware et al. 2000). The scores for each question were summed to give a range of scale scores from 5 to 25, with the scoring of questions (3) and (5) reversed so that higher scores indicated better mental health for each question. Following the standard method, the scores were then range transformed to a final discrete scale ranging from zero to 100, where 100 represents the best level of mental health (Ware et al. 2000). The outcome measure modelled was the change in mental health score between follow-up and baseline score, so that a positive change indicated better mental health.

## Socio-economic variables and transitions in life events

We selected variables that were associated with change in the mental health score. From the baseline data we included age in 10-year bands, sex, Registrar General occupational social class (Macintyre et al. 2003), and the mental health score. We used information on employment status to create a three-level categorical variable of (1) full- or part-time employed; (2) available for and seeking work; or (3) not available for work and so economically inactive (comprising respondents who were looking after home or children full-time, longterm carer, full-time student, on a government training scheme, retired from paid work, or permanently unable to work due to illness or disability). Marital status was coded as (1) married or living as married, (2) single, divorced or separated, or (3) widowed. We assessed physical health status using the Physical Component Summary score (PCS) of the SF-36 (range 0 to 100) as it is constructed to be independent of mental health status (Ware et al. 2000). We divided responses into poor (scores 0 to <50) or good (scores 50 to 100) physical health. Gross household income was categorized into two groups: 'low' income (< £10000 p.a.) and 'high' income ( $\geq$  £10000 p.a.). These categories were chosen because £10000 p.a. in 2001 approximately equated to 60% of median income after housing costs, the UK definition of household poverty (DWP, 2012). Housing tenure was coded as owner-occupier or not owner-occupier, and property wealth was assessed using the national system of council tax valuation bands, in which local government authorities value each domestic residence and assign to one of eight ordinal categories, A to H in Wales. We obtained the council tax valuation band for each participant by matching the sample frame address to the local authority council tax register (Fone et al. 2006*a*). We then created a two-category variable with roughly equal numbers of participants in each: the lowest value council tax bands A and B, and the remaining six higher value bands C to H. Finally, we derived categorical variables for between-wave transitions in employment status, marital status, physical health status, gross household income, housing tenure, and council tax valuation band.

#### Neighbourhood deprivation

We assessed neighbourhood deprivation using an enumeration district dataset of gross household income estimates for 2001 in successive £5000 income bands (Fone et al. 2007b). We defined two variables to distinguish between two levels of neighbourhood deprivation: The first variable, labelled 'high neighbourhood deprivation', was defined as the percentage of households in each neighbourhood with a gross annual household income of less than £5000, approximately equivalent to income from State welfare benefits. The second less extreme variable, labelled 'neighbourhood deprivation', was defined as the percentage of households with an income less than £10000 per annum, based on the UK definition of poverty in 2001 (DWP, 2012). Both variables were transformed to a z score, with zero mean and unit variance, for the analysis.

We assessed neighbourhood migration by deriving a four-level categorical variable labelled as (1) not moving, (2) moving within the borough to a more deprived neighbourhood, (3) moving to a less deprived neighbourhood or (4) moving to the same level of deprivation.

## Individual and neighbourhood social cohesion

We derived neighbourhood social cohesion scores using Buckner's Neighbourhood Cohesion scale

(Buckner, 1988). We first carried out an exploratory factor analysis. Extraction was by principal components analysis and a varimax rotation and two components were extracted with eigenvalues>1. We used the criterion of a factor loading of  $\ge 0.5$  for assigning each of the scale question items to a component. One component, the social cohesion subscale, clearly loaded items on trust and reciprocity whereas the second component loaded items on 'neighbourhood belonging', which is a different construct and was not used in the present study. An ecometric analysis of individual responses to the social cohesion subscale suggested it was an appropriate measure of social cohesion at neighbourhood level (Fone et al. 2006b). We then estimated mean neighbourhood social cohesion scores and divided the distribution of scores into thirds to create a new categorical variable of low, medium and high social cohesion with equal numbers of neighbourhoods in each category (Fone et al. 2007b). To adjust for between-wave change in individual social cohesion scores, we also derived a three-category variable with the middle category representing minimal change of up to  $\pm 2$  points on the scale (range 8–40), the low category representing a decrease in cohesion score of>2 points, and the high category an increase in reported cohesion score of>2 points.

#### Statistical analysis

#### Multilevel modelling strategy

We modelled the change in mental health scores as a continuously distributed variable in a linear multilevel regression model of individuals nested within enumeration districts. We used the iterative generalized least squares estimation procedure in MLwiN software (Rasbash *et al.* 2009), and report parameter estimates for the neighbourhood fixed effects and the intraclass correlation coefficients (ICCs) with accompanying 95% confidence intervals (CIs).

The modelling strategy started with the 'null' twolevel variance components model, in which the variation in the change in mental health score was modelled by a random intercept term for neighbourhoods and a random error term for individuals. In model 1 we fitted the socio-economic and life event transition covariates. Missing data were modelled for each categorical variable as a separate category to minimize data loss and allow direct comparison between models using the deviance (-2 log likelihood). The data were weighted to allow for the unequal electoral ward sampling probability and survey non-response as a function of age, sex and neighbourhood deprivation score that were available for the whole sampling frame, using a standard method (Crockett *et al.* 2011; Fone *et al.* 2012).

We investigated the first objective of our study by adding the two neighbourhood deprivation variables separately to model 1 to form two independent models, 2.1 and 2.2. We investigated the second objective of our study in model 3 by adding the neighbourhood and individual-level social cohesion variables to model 1. Model 3 did not include neighbourhood deprivation. Finally, we investigated the third objective by adding the neighbourhood deprivation and social cohesion variables to model 1. We derived a separate model for both measures of neighbourhood deprivation. Model 4.1 included the high deprivation variable and model 4.2 included the neighbourhood deprivation variable. Each model included terms for the main effects and the interaction between them (neighbourhood deprivation × neighbourhood social cohesion).

To further assess the influence of missing data, we repeated the final analysis of model 4.2 accounting for missing data with multiple imputation using chained equations (White *et al.* 2010). We estimated a pooled estimate from five models with all individual and neighbourhood variables included as predictors.

## Results

The change in mental health score was available for 4426 (97.1%) respondents and the distribution was approximately normal. Over the study period, there was a small decrease in mental health overall: mean change in mental health score -0.83 [standard deviation (s.D.) 20.3, interquartile range (IQR) -10 to 10].

The mean percentage of households per neighbourhood earning less than £5000 p.a. was 12.9% (s.D.=7.6%, 10th–90th centile 4.4–23.1%). For less than £10000 p.a., the mean was 31.3% (s.D.=13.0%, 10th–90th centile 15.5–49.2%). Fig. 1 shows the relationship between neighbourhood deprivation and neighbourhood social cohesion. Although the low social cohesion category had the highest proportion of deprived neighbourhoods, there was substantial variability in neighbourhood deprivation scores within each social cohesion category, such that neighbourhoods could be categorized as both deprived and of high social cohesion.

The distribution of the numbers (%) of respondents and the mean change in mental health score within thirds of neighbourhood deprivation and neighbourhood social cohesion categories are shown in Table 1. The general pattern of the mean change in mental health score associated with the neighbourhood variables was as expected, with a negative change in the high neighbourhood deprivation and low cohesion categories.



Fig. 1. Distributions of neighbourhood deprivation scores (percentage low-income households <£10000 p.a.) by the categories of social cohesion.

# Associations with socio-economic and life event transition variables

Online Supplementary Table S1 shows the numbers (%) of respondents, univariable associations with change in mental health score and the results from fitting the socio-economic and life event transition covariates in model 1. A positive change in mental health score was most strongly associated with age groups 45-74 years at baseline. Compared to remaining in employment, a decrease in mental health score was most strongly associated with the transition to economic inactivity or remaining economically inactive. Transitions from both employment and economic inactivity to being unemployed and seeking work were also associated with a negative change in mental health, although only a few subjects made these transitions. The most striking marital status transition was the decline in mental health associated with bereavement. A decline in mental health was also associated with remaining in low value housing, remaining or becoming a low-income or non-owneroccupier household, and remaining or becoming in poor physical health. Moving to a higher, lower or same level of neighbourhood deprivation compared to not moving showed little association with change in mental health.

## Effect of neighbourhood deprivation and social cohesion

In models 2.1 and 2.2, a decline in mental health was significantly associated with residence in the high deprived neighbourhoods, after adjusting for socio-economic and transition in life event covariates (Table 2). The coefficient for the high neighbourhood deprivation variable was -0.79 (95% CI -1.45 to -0.13), which represents the negative change in mental health score for an increase of 1 s.D. or 7.6% in low-income households. This is equivalent to a change of -1.9 points (95% CI -3.6 to -0.3) on the mental health scale between the 10th and 90th centiles of the high neighbourhood deprivation distribution over the period of the study.

Compared to low social cohesion neighbourhoods, both the medium and high cohesion categories were significantly associated in model 3 with an improvement in mental health (Table 2). The parameter estimates, of +2.15 (95% CI 0.78–3.52) and +1.68 (95% CI 0.27–3.09) respectively, are of greater magnitude than the decline in mental health associated with the neighbourhood deprivation variables in models 2.1 and 2.2.

Both of the neighbourhood deprivation variables were significantly associated with a negative change in mental health in their respective final models

Table 1.	Change in	mean mental	health score	for sub	jects within	neighbourhood	ls
----------	-----------	-------------	--------------	---------	--------------	---------------	----

Variable	Category	Survey respondents ( <i>n</i> =4426) <i>n</i> (%)	Change in mean mental health score (95% CI)
Neighbourhood social cohesion <sup>a</sup>	Low	1270 (28.7)	-1.94 (-3.09 to -0.79)
	Medium	1588 (35.9)	0.26 (-0.81 to 1.33)
	High	1568 (35.4)	-0.89 (-1.96 to 0.18)
High neighbourhood deprivation <sup>b</sup>	Low	1708 (38.6)	-0.43 (-1.41 to 0.55)
	Medium	1414 (31.9)	-0.16 (-1.24 to 0.92)
	High	1304 (29.5)	-2.01 (-3.26 to -0.76)
Neighbourhood deprivation <sup>c</sup>	Low	1708 (38.6)	-0.54 (-1.52 to 0.44)
	Medium	1414 (31.9)	-0.05 ( $-1.14$ to $1.04$ )
	High	1304 (29.5)	-1.99 (-3.23 to -0.75)

CI, Confidence interval.

<sup>a</sup> The distribution of neighbourhood mean social cohesion scores was divided into thirds with equal numbers of neighbourhoods in each third.

<sup>b</sup> The distribution of neighbourhood percentage low-income households<£5000 p.a. was divided into thirds with equal numbers of neighbourhoods in each third.

 $^{c}$  The distribution of neighbourhood percentage low-income households <£10000 p.a. was divided into thirds with equal numbers of neighbourhoods in each third.

4.1 and 4.2 (Table 2). These effects were larger in magnitude than in models 2.1 and 2.2. The medium and high categories of social cohesion also remained significantly associated with an improvement in mental health. The positive coefficient for the interaction between neighbourhood deprivation and social cohesion (+1.49, 95% CI 0.08-2.90) suggested that the adverse association of neighbourhood deprivation with a negative change in mental health was significantly attenuated in high social cohesion neighbourhoods. Fig. 2 shows the predicted change in individual mental health score from model 4.2 associated with neighbourhood deprivation, together with the difference in slopes between the social cohesion groups representing the interaction term. The predicted change in mental health score was calculated for the 10th and 90th centiles of the household low-income distribution. The difference between them was -2.8 in the low social cohesion group and 1.1 in the high cohesion group. The difference between the groups was 3.9 (95% CI 0.2-7.6).

#### Neighbourhood-level variation

In the null model, 2.8% (95% CI 1.2–4.5) of the variance was at the neighbourhood level, and this reduced minimally to 2.7% (95% CI 0.7–4.6) in model 1 after fitting the covariates (Table S1). The ICCs (%) for the proportion of unexplained random variance in the change in mental health score at neighbourhood level are shown in Table 2 for the neighbourhood models. In the final model 4.2, the ICC was 2.1%

(95% CI 0.2–4.0). The results from the multiple imputation models showed no substantive differences in the magnitude of the parameter estimates, nor in their interpretation.

#### Discussion

Our results show significant associations between neighbourhood deprivation and increasingly poor mental health over a 7-year period. We found a widening of mental health inequalities between neighbourhoods and the effect was strongest in the most deprived neighbourhoods. Living in a medium or high social cohesion neighbourhood was strongly associated with an improvement in mental health compared to living in a low social cohesion neighbourhood. The significant positive interaction between neighbourhood deprivation and social cohesion suggested that living in a high social cohesion neighbourhood modified the adverse effect of neighbourhood deprivation on poor mental health. After adjustment for socio-economic and life event risk factors associated with change in mental health, and possible health selection effects from migration, we can be confident of a temporal relationship in this longitudinal analysis because both the neighbourhood deprivation and social cohesion variables were measured 7 years before the change in mental health score.

The positive impact of high social cohesion offers potential mechanisms through which the adverse impact of residence in deprived neighbourhoods on mental health could be addressed. The relationships

		Parameter estimate (95% CI)					
Variable	Parameter	Model 2.1 <sup>ª</sup> High neighbourhood deprivation <sup>b</sup>	Model 2.2 <sup>ª</sup> Neighbourhood deprivation <sup>c</sup>	Model 3 <sup>a,e</sup> Social cohesion <sup>d</sup>	Model 4.1 <sup>a,b,e</sup> High neighbourhood deprivation <sup>b</sup> and social cohesion <sup>d</sup>	Model 4.2 <sup>a,c,e</sup> Neighbourhood deprivation <sup>c</sup> and social cohesion <sup>d</sup>	
Neighbourhood deprivation	Percentage low-income households (z score)	-0.79 (-1.45 to -0.13)	-0.51 (-1.17 to 0.15)		-1.34 (-2.23 to -0.45)	-1.08 (-2.01 to -0.14)	
Neighbourhood social cohesion	Low cohesion Medium cohesion High cohesion			Reference 2.15 (0.78 to 3.52) 1.68 (0.27 to 3.09)	Reference 1.90 (0.54 to 3.26) 1.62 (0.16 to 3.09)	Reference 1.98 (0.62 to 3.34) 1.70 (0.24 to 3.17)	
Interaction: neighbourhood deprivation×social cohesion	Low cohesion Medium cohesion High cohesion				Reference 1.43 (-0.02 to 2.85) 1.72 (0.26 to 3.18)	Reference 1.23 (-0.17 to 2.64) 1.49 (0.08 to 2.90	
ICC (%)	0	2.60 (0.6 to 4.5)	2.66 (0.7 to 4.6)	2.28 (0.4 to 4.1)	2.02 (0.12 to 3.9)	2.12 (0.20 to 4.0)	

CI, Confidence interval; ICC, intraclass correlation coefficient.

<sup>a</sup> Adjusted for baseline age group, sex, social class and mental health score; transitions in employment, marital status, housing tenure, council tax band of residence, household income, physical health status; within-study area migration.

<sup>b</sup> Percentage low-income households<£5000 p.a.

<sup>c</sup> Percentage low-income households<£10000 p.a.

<sup>d</sup> The distribution of neighbourhood mean social cohesion scores was divided into thirds with equal numbers of neighbourhoods in each third.

<sup>e</sup> Also adjusted for change in individual social cohesion.



**Fig. 2.** Relationship between model-predicted change in mental health scores, social cohesion (SC) and neighbourhood deprivation (percentage low-income households<£10000 p.a.).

between neighbourhood deprivation and social cohesion are complex; people living in deprived neighbourhoods tend to have more locally based family ties and friendships, but with lower degrees of trust and practical help (Stafford et al. 2003). A possible explanation for our findings is that high levels of neighbourhood social cohesion based on friendships, visiting and borrowing and exchange of favours with neighbours may offset the detrimental effects of social disadvantage by facilitating access to networks and services that influence health, and social and emotional support (Kawachi & Berkman, 2000). These mechanisms suggest that interventions that help to facilitate social interaction and exchange may increase levels of social cohesion in deprived neighbourhoods by enhancing trust and reciprocity.

There have been three previous studies, all set in the UK, that have investigated longitudinal multilevel associations between neighbourhood deprivation and mental health in a general population sample. Two of the studies, using the British Household Panel Survey, found that over periods of 1 year (Weich et al. 2005) and 5 years of follow-up (Propper et al. 2005), neighbourhood deprivation had little effect on change in mental health. The third study, analysing data from the Whitehall II study of British civil servants, found that longer residence of more than 10 years in a deprived neighbourhood, compared to less deprived, was associated with a widening of mental health inequality (Stafford et al. 2008c). Our study findings of a widening in mental health inequality is further evidence that there are potentially important effects of neighbourhood deprivation on mental health. One previous longitudinal study set in the USA investigated associations between neighbourhood social cohesion and change in mental health, in 1919 subjects aged 45–84 years with 4–5 years of follow-up (Mair *et al.* 2009). The onset of depression, measured using the Center for Epidemiologic Studies Depression scale, was not associated with baseline social cohesion measured at the census tract level from responses to four questions on trust and reciprocity. To our knowledge, our present study is the first to suggest that neighbourhood social cohesion is associated with an improvement in mental health over time and is a potentially important factor in reducing the adverse impact of neighbourhood deprivation on mental health.

In common with these four previous longitudinal multilevel analyses of neighbourhoods and mental health, we found that most of the unexplained variability in change in mental health scores was between individuals. However, although the between-neighbourhood variance was of the order of 2%, there was sufficient contrast of exposure to detect an important association between change in mental health and the neighbourhood variables (Sampson *et al.* 2002; Merlo, 2003).

The main strength of the study arises from the in-depth sampling of a geographically defined population of socio-economic contrast, with detailed longitudinal data on exposures from respondents linked to small neighbourhood areas of residence (Fone *et al.* 2012). We were able to examine the joint temporal effect of robust neighbourhood measures of deprivation and social cohesion on change in mental health over a 7-year period. Measuring social cohesion at neighbourhood level is not straightforward, but is important because social cohesion is conceptualized as a collective neighbourhood-level characteristic measured by the levels of trust, norms of reciprocity and the formation of strong social bonds within the local social structure (Kawachi & Berkman, 2000; Stafford *et al.* 2003; Subramanian *et al.* 2003). In an ecometric analysis we found that Buckner's Neighbourhood Cohesion scale gave an acceptable neighbourhood measure of social cohesion, derived from the aggregation of individual responses within neighbourhoods (Fone *et al.* 2006*b*). We found no evidence that samesource bias, in which subjects with poorer mental health may be more likely to report negative responses to the social cohesion scale, was operating in the survey scale responses (Fone *et al.* 2007*b*).

Significant neighbourhood-level effects can result from health selection bias, where people of higher socio-economic status are able to move out of deprived neighbourhoods (Norman *et al.* 2005; Connolly *et al.* 2007). To determine whether this bias was operating in our study, we collected accurate data on time and place of within-borough migration. We assessed whether the subject moved to a same, more, or less deprived neighbourhood and found no evidence of a health selection effect.

We carried out a comprehensive adjustment for potentially confounding compositional effects by including baseline socio-economic status and transitions in life events associated with change in mental health, including employment status (Pevalin & Goldberg, 2003; Thomas et al. 2005, 2007), marital status (Pevalin & Goldberg, 2003; Wade & Pevalin, 2004) and physical health (Pevalin & Goldberg, 2003), and we also adjusted for other transitions in housing tenure and other factors that were significantly associated with change in mental health in our dataset. Our findings are in line with these previous studies. However, we had no information on when the individual transitions occurred during the 7-year follow-up, or whether more than one transition occurred within any category. As the data were only recorded at two time-points, individuals who underwent several transitions may not have been correctly classified. The same problem arises in the measurement of mental health at two time-points without information on intervening variability in scores. Two-wave longitudinal studies can confound true change with measurement error (Singer & Willett, 2003), but the MHI-5 scale has been shown to be a valid and reliable measure of change (Hemingway et al. 1997).

A further limitation of the study is the loss to followup of 6334 (58%) subjects from the baseline 10892 respondents aged 18–74 years (Fone *et al.* 2012). Of these, 1341 (12%) had died or moved out of the study area and were removed from the follow-up sampling frame. Of the non-responders, 358 (3.7%) of the 9551 subjects could not be traced, 108 (1.1%) were too unwell to complete a questionnaire and 10 (0.1%) died after the wave 2 sample was taken. A total of 4277 refused to participate. We could not validate the age and/or sex of 240 responders, leaving a total of 4558 for analysis, representing 27.4% of the original sampling frame eligible in the study. Bias may have resulted from non-participants at follow-up reporting lower baseline mental health scores and socio-economic status, and we had no information on change in mental health in non-responders between different neighbourhood deprivation and social cohesion groups. We were able to account for attrition as far as possible by weighting the data for the unequal sampling probability of the sampling design and survey non-response (Fone et al. 2012), using a standard published method for population surveys (Crockett et al. 2011).

Although we measured a wide range of socioeconomic and life event transition variables, including some not measured in other surveys, such as the council tax band as a proxy for property wealth, there is always the possibility of unmeasured confounding as a possible explanation for the study findings. There is also the opposite problem that many of the confounding variables included in the analysis are highly correlated, which can result in model overfitting (Macintyre et al. 2002). We made the assumption that the administratively defined UK Census enumeration district was a good measure of a meaningful neighbourhood, at which small-area exposures can operate. Neighbourhoods should ideally be homogeneous with respect to socio-economic and other contextual characteristics but, in general, administrative boundaries tend to create non-homogeneous areas that can diminish between-area variability. We found the baseline social cohesion scores to fit into this pattern (Fone et al. 2007b), which tends to lead to conservative estimates (Blakely & Woodward, 2000; Stafford et al. 2008b). Therefore, our results are unlikely to overestimate the strength of the associations between change in mental health and the neighbourhood deprivation and social cohesion measures.

In summary, our results suggest that neighbourhood deprivation and neighbourhood social cohesion are important factors in understanding the social epidemiology of the mental health of individuals. Neighbourhood deprivation contributes to widening inequalities in mental health. We have found prospective evidence to support the hypothesis that high neighbourhood social cohesion significantly mitigates the adverse effect of neighbourhood deprivation on mental health. Policies and interventions to reduce mental health inequalities across the socio-economic gradient should recognize the importance of social context, and interventions should include components that operate not only for individuals but also at the neighbourhood level.

## Supplementary material

For supplementary material accompanying this paper visit http://dx.doi.org/10.1017/S0033291713003255.

## Acknowledgements

We thank the study participants and many of the staff of the former Gwent Health Authority, the Caerphilly Local Health Group and their successor organizations, and the Chief Executive's Office and the Electoral Services Department of Caerphilly county borough council for their help in running this collaborative study.

This work was supported by a Welsh Assembly Government/UK Medical Research Council Health Research Partnership Award (H07-3-030).

The work was undertaken at the Centre for the Development and Evaluation of Complex Interventions for Public Health Improvement, a UK Clinical Research Collaboration Public Health Research: Centre of Excellence. Funding from the British Heart Foundation, Cancer Research UK, the Economic and Social Research Council (RES-590-28-0005), the Medical Research Council, the Welsh Assembly Government and the Wellcome Trust (WT087640MA), under the auspices of the UK Clinical Research Collaboration, is gratefully acknowledged.

The funding organizations played no role in the design and conduct of the study; collection, management, analysis, and interpretation of the data; and preparation, review, or approval of the manuscript.

## **Declaration of Interest**

None.

## References

- Almedom A (2005). Social capital and mental health: an interdisciplinary review of primary evidence. *Social Science and Medicine* **61**, 943–964.
- Berwick DM, Murphy JM, Goldman PA, Ware Jr JE, Barsky AJ, Weinstein MC (1991). Performance of a five-item mental health screening test. *Medical Care* **29**, 169–176.
- Blakely TA, Woodward AJ (2000). Ecological effects in multi-level studies. *Journal of Epidemiology and Community Health* 54, 367–374.

- **Buckner JC** (1988). The development of an instrument to measure neighbourhood cohesion. *American Journal of Community Psychology* **16**, 771–791.
- **Cattell V** (2001). Poor people, poor places, and poor health: the mediating role of social cohesions and social capital. *Social Science and Medicine* **52**, 1501–1516.
- Connolly S, O'Reilly D, Rosato M (2007). Increasing inequalities in health: is it an artefact caused by the selective movement of people? *Social Science and Medicine* 64, 2008–2015.
- **Crockett A, Afkhami R, Rafferty A, Higgins V, Marshall A** (2011). *Weighting the Social Surveys*. ESDS Government. UK Data Archive and Institute for Social and Economic Research.
- Croft-Jefferys C, Wilkinson G (1989). Estimated costs of neurotic disorder in UK general practice. *Psychological Medicine* 19, 549–558.
- Cuijpers P, Smits N, Donker T, ten Have M, de Graaf R (2009). Screening for mood and anxiety disorders with the five-item, the three-item, and the two-item Mental Health Inventory. *Psychiatry Research* **168**, 250–255.
- De Silva MJ, McKenzie K, Harpham T, Huttly SRA (2005). Social capital and mental illness: a systematic review. *Journal of Epidemiology and Community Health* 59, 619–627.
- **Dewa CS, Thompson AH, Jacobs P** (2011). The association of treatment of depressive episodes and work productivity. *Canadian Journal of Psychiatry* **56**, 743–750.
- DWP (2012). Households Below Average Income (HBAI): An Analysis of the Income Distribution 1994/95 – 2009/10. National Statistics. Department for Work and Pensions: London (www.gov.uk/government/publications/ households-below-average-income-hbai-199495-to-200910). Accessed 2 December 2013.
- Fine B (2001). Social Capital Versus Social Theory. Political Economy and Social Science at the Turn of the Millennium, pp. 189–190. Routledge: London.
- Fone DL, Dunstan FD, Christie S, Jones A, West J, Webber M, Lester N, Watkins J (2006*a*). Council tax valuation bands, socio-economic status and health outcome: a cross-sectional analysis from the Caerphilly Health and Social Needs Study. *BMC Public Health* **6**, 115.
- Fone DL, Dunstan FD, John A, Lloyd K (2007*a*). Associations between the common mental disorders and the Mental Illness Needs Index in community settings: a multilevel analysis. *British Journal of Psychiatry* **191**, 158–163.
- Fone DL, Dunstan FD, Lloyd K, Williams GH, Watkins J, Palmer SR (2007b). Does social cohesion modify the association between area income deprivation and mental health? A multilevel analysis. *International Journal of Epidemiology* **36**, 338–345.
- Fone DL, Dunstan FD, White J, Kelly M, Farewell D, John G, Lyons RA, Lloyd K (2012). Cohort profile: the Caerphilly Health and Social Needs Electronic Cohort Study (E-CATALYsT). *International Journal of Epidemiology*. Published online: 5 November 2012. doi: 10.1093/ije/dys175.
- Fone DL, Farewell DM, Dunstan FD (2006b). An ecometric analysis of neighbourhood cohesion. *Population Health Metrics* **4**, 17.

Ford E, Clark C, McManus S, Harris J, Jenkins R, Bebbington P, Brugha T, Meltzer H, Stansfeld SA (2010). Common mental disorders, unemployment and welfare benefits in England. *Public Health* **124**, 675–681.

**Fryers T, Melzer D, Jenkins R** (2003). Social inequalities and the common mental disorders. *Social Psychiatry and Psychiatric Epidemiology* **38**, 229–237.

Harpham T, Grant E, Thomas E (2002). Measuring social capital within health surveys: key issues. *Health Policy and Planning* **17**, 106–111.

Hemingway H, Stafford M, Stansfeld S, Shipley M, Marmot M (1997). Is the SF-36 a valid measure of change in population health? Results from the Whitehall II study. *British Medical Journal* 15, 1273–1279.

Kawachi I, Berkman LF (2000). Social cohesion, social capital, and health. In *Social Epidemiology* (ed. L. F. Berkman and I. Kawachi), Ch. 8, pp. 174–190. Oxford University Press: New York.

Kawachi I, Kennedy BP, Glass R (1999). Social capital and self-rated health: a contextual analysis. *American Journal of Public Health* **89**, 1187–1193.

Kelly M, Dunstan F, Lloyd K, Fone D (2008). Evaluating cutpoints for the MHI-5 and MCS using the GHQ-12: a comparison of five different methods. *BMC Psychiatry* 8, 10.

LaPierre TA (2012). The enduring effects of marital status on subsequent depressive symptoms among women: investigating the roles of psychological, social and financial resources. *Journal of Epidemiology and Community Health* 66, 1056–1062.

Lloyd KR, Jenkins R, Mann A (1996). Long term outcome of patients with neurotic illness in general practice. *British Medical Journal* 313, 26–28.

Macintyre S, Ellaway A, Cummins S (2002). Place effects on health: how can we conceptualise, operationalise and measure them? *Social Science and Medicine* 55, 125–139.

Macintyre S, McKay L, Der G, Hiscock R (2003). Socio-economic position and health: what you observe depends on how you measure it. *Journal of Public Health Medicine* **25**, 288–294.

Mair C, Diez Roux AV, Galea S (2008). Are neighbourhood characteristics associated with depressive symptoms? A review of evidence. *Journal of Epidemiology and Community Health* 62, 940–946.

Mair C, Diez Roux AV, Shen M, Shea S, Seeman T, Echeverria S, O'Meara ES (2009). Cross-sectional and longitudinal associations of neighborhood cohesion and stressors with depressive symptoms in the multiethnic study of atherosclerosis. *Annals of Epidemiology* **19**, 49–57.

McCabe CJ, Thomas KJ, Brazier JE, Coleman P (1996). Measuring the mental health status of a population: a comparison of the GHQ-12 and the SF-36 (MHI-5). *British Journal of Psychiatry* **169**, 517–521.

McKenzie K, Whitley R, Weich S (2002). Social capital and mental health. *British Journal of Psychiatry* 181, 280–283.

McManus S, Meltzer H, Brugha T, Bebbington P, Jenkins R (2009). Adult Psychiatric Morbidity in England, 2007. Results of a Household Survey. The NHS Information Centre for Health and Social Care: London. Merlo J (2003). Multilevel analytical approaches in social epidemiology: measures of health variation compared with traditional measures of association. *Journal of Epidemiology and Community Health* **57**, 550–552.

Muntaner C, Lynch J, Davy Smith G (2000). Social capital and the third way in public health. *Critical Public Health* **10**, 107–124.

Murayama H, Fujiwara Y, Kawachi I (2012). Social capital and health: a review of prospective multilevel studies. *Journal of Epidemiology* **22**, 179–187.

Norman P, Boyle P, Rees P (2005). Selective migration, health and deprivation: a longitudinal analysis. *Social Science and Medicine* 60, 2755–2771.

**Pevalin DJ, Goldberg DP** (2003). Social precursors to onset and recovery from episodes of common mental illness. *Psychological Medicine* **33**, 299–306.

Propper C, Jones K, Bolster A, Burgess S, Johnston R, Sarker R (2005). Local neighbourhood and mental health: evidence from the UK. *Social Science and Medicine* 61, 2065–2083.

Putnam RD (1993). Making Democracy Work: Civic Traditions in Modern Italy. Princeton University Press: Princeton, NJ.

Rai D, Kosidou K, Lundberg M, Araya R, Lewis G, Magnusson C (2012). Psychological distress and risk of long-term disability: population-based longitudinal study. *Journal of Epidemiology and Community Health* **66**, 586–592.

Rasbash J, Steele F, Browne W, Goldstein H (2009). A User's Guide to MlwiN version 2.10. Centre for Multilevel Modelling: University of Bristol.

Rumpf H-J, Meyer C, Hapke U, John U (2001). Screening for mental health: validity of the MHI-5 using DSM-IV Axis I psychiatric disorders as gold standard. *Psychiatry Research* **105**, 243–253.

Russ TC, Stamatakis E, Hamer M, Starr JM, Kivimäki M, Batty GD (2012). Association between psychological distress and mortality: individual participant pooled analysis of 10 prospective cohort studies. *British Medical Journal* 345, e4933.

Sacker A, Wiggins RD (2002). Age-period-cohort effects on inequalities in psychological distress, 1981-2000. *Psychological Medicine* 32, 977–990.

Sampson RJ, Morenoff JD, Gannon-Rowley T (2002). Assessing 'neighbourhood effects': social processes and new directions in research. *Annual Review of Sociology* 28, 443–478.

Singer JD, Willett JB (2003). A framework for investigating change over time. In Applied Longitudinal Data Analysis. Modelling Change and Event Occurrence, pp. 9–10. Oxford University Press: New York.

Skapinakis P (2007). Commentary: Socioeconomic position and common mental disorders: what do we need to know? International Journal of Epidemiology 36, 786–788.

Stafford M, Bartley M, Sacker A, Marmot M (2003). Measuring the social environment: social cohesion and material deprivation in English and Scottish neighbourhoods. *Environment and Planning A* 35, 1459–1475.

**Stafford M, De Silva M, Stansfeld S, Marmot M** (2008*a*). Neighbourhood social capital and common mental disorder: testing the link in a general population sample. *Health and Place* **14**, 394–405.

Stafford M, Duke-Williams O, Shelton N (2008b). Small area inequalities in health: are we underestimating them? *Social Science and Medicine* 67, 891–899.

Stafford M, Gimeno D, Marmot MG (2008c). Neighbourhood characteristics and trajectories of health functioning: a multilevel prospective analysis. *European Journal of Public Health* 18, 604–610.

Subramanian SV, Kim DJ, Kawachi I (2002). Social trust and self-rated health in US communities: a multilevel analysis. *Journal of Urban Health* **79**, 521–534.

Subramanian SV, Lochner KA, Kawachi I (2003). Neighborhood differences in social capital: a compositional artifact or a contextual construct? *Health and Place* **9**, 33–44.

Thomas C, Benzeval M, Stansfeld SA (2005). Employment transitions and mental health: an analysis from the British household panel survey. *Journal of Epidemiology and Community Health* 59, 243–249.

Thomas C, Benzeval M, Stansfeld S (2007). Psychological distress after employment transitions: the role of subjective financial position as a mediator. *Journal of Epidemiology and Community Health* **61**, 48–52.

Wade TJ, Pevalin DJ (2004). Marital transitions and mental health. *Journal of Health and Social Behaviour* 45, 155–170.

Ware JE Jr., Gandek B (1998). Overview of the SF-36 Health Survey and the International Quality of Life Assessment (IQOLA) Project. *Journal of Clinical Epidemiology* 51, 903–912.

Ware JE Jr., Kosinski M, Dewey JE (2000). *How to Score Version 2 of the SF-36<sup>®</sup> Health Survey*. QualityMetric Incorporated: Lincoln, RI.

Ware JE Jr., Kosinski M, Gandek B (1993). SF-36<sup>®</sup> Health Survey: Manual and Interpretation Guide. QualityMetric Incorporated: Lincoln, RI.

Weich S (1997). Prevention of the common mental disorders: a public health perspective. *Psychological Medicine* **27**, 757–764.

Weich S, Twigg L, Lewis G, Jones K (2005). Geographical variation in rates of common mental disorders in Britain: prospective cohort study. *British Journal of Psychiatry* 187, 29–34.

White IR, Royston P, Wood AM (2010). Multiple imputation using chained equations: issues and guidance for practice. *Statistics in Medicine* **30**, 377–399.