RESEARCH REPORT

Education and income: which is more important for mental health?

R Araya, G Lewis, G Rojas, R Fritsch

Study objective: To assess which indicators of socioeconomic status are associated with an increased prevalence of common mental disorders.

Design: Cross sectional household survey.

Setting: Santiago, Chile.

Participants: Random sample of adults aged 16–65 residing in private households.

Main results: Less education (odds ratio 2.44, 95% confidence intervals 1.50 to 3.97), a recent income decrease (odds ratio 2.14, 1.70 to 2.70), and poor housing (odds ratio 1.53, 1.05 to 2.23), were the only socioeconomic status variables that remained significantly associated with an increased prevalence of common mental disorders after adjustments. The prevalence of common mental disorders was also higher among people with manual unskilled occupations, overcrowded housing, and lower per capita income but these associations disappeared after adjustment for other explanatory and confounding variables.

Conclusions: There is a strong, inverse, and independent association between education and common mental disorders. However, income was not associated with the prevalence of common mental disorders, after adjusting for other socioeconomic variables. Similar results have been found in other Latin American studies but British studies tend to find the opposite, that income but not education is associated with common mental disorders. Understanding the impact of socioeconomic factors on mental health requires research in poor as well as rich countries.

According to the World Health Organisation much of the global health burden is attributable to the non-psychotic and common mental disorders, such as depression and anxiety.1 Although some psychotic conditions might be more severe and burdensome on an individual basis, the volume and duration of the more common mental conditions yield a much larger aggregate burden.

Socioeconomic status is a complex concept that has been borrowed by medical researchers, often without due regard to its sociological inheritance. In epidemiology the concept is assessed indirectly using a variety of different measures with different implications for social and economic policy. Income, material possessions (or standard of living), occupational status, and education are the domains most commonly studied. Nevertheless, these measures are not equivalent and might have different meanings and represent different concepts of social position in different cultures. For instance, income changes throughout life while education remains comparatively “frozen” after early adulthood and educational attainments can have different meaning in different places. The association between relative or absolute income and health is among the most commonly reported in the scientific literature. However, several recent studies have found that this association is weakened or disappear when controlling for other socioeconomic variables, especially education.2,3 Understanding the relation between socioeconomic status and health depends upon distinguishing these various measures and examining for independent associations with health.

A large body of scientific literature, mainly from western countries, shows that social disadvantage, especially lack of material possessions, lower income, and financial strain, are associated with common mental disorders (CMD).4–10 However, much less is known of the socioeconomic determinants of CMD in non-western countries. Nevertheless most of the world’s population live outside the western countries and it is these developing countries that experience the largest degree of socioeconomic inequality. Psychiatric household surveys from these countries rarely get published in peer reviewed journals.11,12 There have been a few published from Latin America over the past 20 years13–15 but because of some methodological problems, especially the lack of adjustment for potential explanatory or confounding variables no firm conclusions can be reached. Thus it is still unclear which socioeconomic factors, if any, are independently associated with an increased prevalence of CMD in Latin America and elsewhere in the developing world.

There seems to be an important gap in our knowledge about the most accurate indicators of social position and the most health sensitive socioeconomic factors across cultures. To understand better the relation between socioeconomic status and mental health, we need to carry out research in rich as well as poor countries. We carried out a large cross sectional survey of the private households of Santiago, the capital of Chile to examine which indicators of socioeconomic status are most strongly associated with the prevalence of CMD.

METHODS
Participants and sampling
This paper describes data from the Santiago Mental Disorders Survey undertaken between 1996 and 1998 in Santiago, Chile.16 The sampling involved a three stage clustered design. Households within 200 sectors from all the 35 boroughs of Santiago, capital of Chile, were randomly chosen with a probability proportional to the population size. A larger sampling fraction was needed in the most affluent boroughs to permit testing for socioeconomic differences between groups. One person per household was chosen at random.17 The sampling framework was the total adult population living in private households of Santiago, representing 3,217,177 people. Further details of the sampling design can be requested from the authors.
Variables

Main outcome variable
CMD were assessed using the Revised Clinical Interview Schedule (CIS-R). This is a structured interview that has been fully standardised so that it can be administered by lay interviewers. People scoring 12 or above on the CIS-R total score (range 0–60) were regarded as suffering from a CMD. This threshold has been used extensively in Chile, the UK, and elsewhere. The CIS-R in its English and Spanish versions has been used extensively in primary care and community studies with validity and reliability comparable or better than other commonly used structured interviews in mental health.

Socioeconomic variables

The following socioeconomic variables were included:

Education
Subjects’ own educational level was subdivided in six categories according to whether participants had completed primary education (8 years), secondary education (12 years), or higher education (more than 12 years).

Monthly per capita income and recent income decrease
The sum of net monthly salaries and other income (pensions, dividends, interests, or rents) contributed by each household member was divided by the number of residents regardless of age to generate a per capita monthly income. People were requested to consider mean income over six months before the interview. Per capita income was treated both as a continuous and categorical (quartiles and deciles) variable. Also people were asked if that house had “experienced a significant drop in income over the past six months”.

Quality, tenure, and overcrowding of housing
The location, size, and quality of repairs of the property were assessed through visual inspection by the interviewer. Properties were classified in three categories: good, fair, and poor. Tenure of the house was divided into owned, rented, or other form of tenure such as squatter or lending. Crowding was estimated as the number of bedrooms divided by the number of people living in the premises.

Occupational status of the main breadwinner
We used the Chilean Institute of National Statistics scale with four categories: (1) Low status but stable occupation: involving employed manual non-specialised workers. (2) Low status and unstable occupation: involving casual manual non-specialised workers. (3) Middle status occupation: involving non-manual workers, with no professional qualifications. (4) High status occupation: involving non-manual professional or business people with prestigious posts. For households where nobody was currently employed, coding was based on the last business people with prestigious posts. For households where

Results were adjusted by potential confounding variables such as age and sex and other possible explanatory variables such as working status (employed, unemployed, economically inactive, unable to work for health reasons) and the presence of a self reported physical disease in response to an open-ended question (do you suffer from any physical problem or disability at present?). Two independent physicians assessed if the physical problem would require medical attention in which case they classified this according to the bodily system involved. The self reported number of friends or relatives who could provide emotional or practical support if needed was determined with a single, open-ended question.

Several pilot studies were carried out to study the validity, reliability of the psychiatric interview and the feasibility of the procedures used.

Local bodies granted ethical approval.

Statistical analysis
In view of the multistage random sampling design, prevalence estimates with their corresponding confidence intervals were calculated using the survey commands of the program STATA, which takes into account the effect of the sampling strategy (stratification and clustering) and sampling weights. The association between CMD and each socioeconomic variable was examined by calculating odd ratios and their 95% confidence intervals using logistic regression models, both before and after adjusting simultaneously for sex, age, working status, physical illness, and social support. Subsequently variables were adjusted for the potential confounding effect of the other variables representing socioeconomic status (education, income, occupational status, and quality of housing) in other logistic regression models. We tested for linear relations and interactions using likelihood ratio tests.

RESULTS
The sample framework comprised 4693 addresses, 393 of which were declared unusable because they were non-residential or contained only residents over 65. So effectively 4300 private households were approached. Altogether 3870 subjects were interviewed, a response rate of 90%. The mean age of the sample was 35.6 (standard deviation 14.11), 53% were women, most people were married (55%), one third single (34%), the rest were separated (5%), cohabiting (4%), and widowed (3%). Unemployment was lower (4.2%) than official statistics for the country (5%). However 8.5% of the sample reported freelance work, many of whom were effectively unemployed. Forty seven per cent were fully employed, 22% were housekeepers, 16% students, and less than 1% were permanently out of work for health reasons. Further details about the sample characteristics can be found elsewhere. Missing data on the main independent variables were minimal for education, income decrease, and total household income (37, 18, and 56 respectively).

Lower income and recent income decreases were more common among less educated people (table 1). Similarly those

<table>
<thead>
<tr>
<th>Table 1: Educational level, per capita income quartiles, and income decrease. Weighted percentages and 95% confidence intervals</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Educational level</strong></td>
</tr>
<tr>
<td>------------------------</td>
</tr>
<tr>
<td>Per capita income quartiles</td>
</tr>
<tr>
<td></td>
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<td></td>
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<tr>
<td></td>
</tr>
<tr>
<td>Income decrease</td>
</tr>
</tbody>
</table>
Table 2  Association between educational levels and common mental disorders (CMD). Percentage weighted prevalence, unadjusted, and adjusted odds ratios (95% confidence intervals). Odds ratios refer to baseline group for each category.

<table>
<thead>
<tr>
<th>Educational level</th>
<th>Weighted sample %</th>
<th>Prevalence (95% CI)</th>
<th>Unadjusted odds ratios</th>
<th>Model 1*</th>
<th>Model 2†</th>
<th>Model 3‡</th>
</tr>
</thead>
<tbody>
<tr>
<td>Higher</td>
<td>15.7</td>
<td>11.1 (8.7 to 14.1)</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Secondary complete</td>
<td>31.2</td>
<td>22.3 (19.3 to 25.7)</td>
<td>2.31 (1.67 to 3.19)</td>
<td>1.82</td>
<td>1.63</td>
<td>1.51</td>
</tr>
<tr>
<td>Secondary incomplete</td>
<td>29.4</td>
<td>26.5 (22.5 to 30.9)</td>
<td>2.89 (2.06 to 4.05)</td>
<td>2.30</td>
<td>1.90</td>
<td>1.83</td>
</tr>
<tr>
<td>Primary complete</td>
<td>9.3</td>
<td>33.5 (25.2 to 42.9)</td>
<td>4.04 (2.45 to 6.64)</td>
<td>2.73</td>
<td>2.49</td>
<td>2.16</td>
</tr>
<tr>
<td>Primary incomplete</td>
<td>14.4</td>
<td>37.8 (31.7 to 44.3)</td>
<td>4.87 (3.39 to 7.00)</td>
<td>3.35</td>
<td>2.69</td>
<td>2.44</td>
</tr>
</tbody>
</table>

*Model 1 adjusts for age, sex, physical disease, working status, and social support. †Model 2 adjusts for per capita income, income decrease, and housing variables. ‡Model 3 adjusts for all variables.

Table 3  Association between income, recent income decrease, occupational and working status, and common mental disorders (CMD). Percentage weighted prevalence, unadjusted, and adjusted odds ratios (95% confidence intervals). Odds ratios refer to baseline group for each category.

<table>
<thead>
<tr>
<th>Income per capita quartiles</th>
<th>Weighted sample %</th>
<th>Prevalence (95% CI)</th>
<th>Unadjusted odds ratios (95% CI)</th>
<th>Model 1* odds ratios (95% CI)</th>
<th>Model 2† odds ratios (95% CI)</th>
<th>Model 3‡ odds ratios (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Highest</td>
<td>25.0</td>
<td>14.8 (14.7 to 14.8)</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Middle high</td>
<td>25.5</td>
<td>23.2 (23.1 to 23.3)</td>
<td>1.75 (1.32 to 2.31)</td>
<td>1.52 (1.11 to 2.08)</td>
<td>1.24 (0.91 to 1.70)</td>
<td>1.17 (0.82 to 1.66)</td>
</tr>
<tr>
<td>Middle low</td>
<td>24.7</td>
<td>29.0 (28.9 to 29.2)</td>
<td>2.36 (1.71 to 3.26)</td>
<td>1.88 (1.35 to 2.62)</td>
<td>1.35 (0.94 to 1.94)</td>
<td>1.21 (0.83 to 1.76)</td>
</tr>
<tr>
<td>Lowest</td>
<td>24.9</td>
<td>33.4 (33.2 to 33.5)</td>
<td>2.89 (2.16 to 3.87)</td>
<td>1.93 (1.40 to 2.68)</td>
<td>1.37 (0.97 to 1.98)</td>
<td>1.03 (0.70 to 1.53)</td>
</tr>
<tr>
<td>Income decrease</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>No</td>
<td>69.1</td>
<td>19.3 (17.2 to 21.5)</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Yes</td>
<td>30.9</td>
<td>38.0 (33.8 to 42.4)</td>
<td>2.57 (2.08 to 3.17)</td>
<td>2.29 (1.84 to 2.86)</td>
<td>2.18 (1.74 to 2.74)</td>
<td>2.14 (1.70 to 2.70)</td>
</tr>
<tr>
<td>Occupational status of main breadwinner</td>
<td></td>
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<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>High</td>
<td>13.9</td>
<td>12.6 (9.93 to 15.9)</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Medium</td>
<td>36.6</td>
<td>23.4 (19.9 to 27.4)</td>
<td>2.13 (1.55 to 2.93)</td>
<td>1.68 (1.21 to 2.33)</td>
<td>1.01 (0.70 to 1.47)</td>
<td>0.98 (0.67 to 1.43)</td>
</tr>
<tr>
<td>Low</td>
<td>39.5</td>
<td>28.6 (25.4 to 31.9)</td>
<td>2.78 (2.03 to 3.81)</td>
<td>1.93 (1.37 to 2.73)</td>
<td>0.98 (0.65 to 1.49)</td>
<td>0.91 (0.60 to 1.40)</td>
</tr>
<tr>
<td>Low unstable</td>
<td>10.3</td>
<td>34.7 (28.2 to 41.8)</td>
<td>3.67 (2.47 to 5.47)</td>
<td>2.47 (1.60 to 3.82)</td>
<td>1.22 (0.74 to 2.01)</td>
<td>1.18 (0.72 to 1.94)</td>
</tr>
</tbody>
</table>

*Model 1 adjusts for age, sex, physical disease, working status, and social support. †Model 2 adjusts for per education, capita income, income decrease, and quality of housing. ‡Model 3 adjusts for all variables.

People with lower income were over twice as likely to report experiencing income decreases recently than those in the highest income group (43% (42.7 to 43.1) compared with 20% (19.7 to 19.9%). As expected there was some intercorrelation between the main independent variables but none of the variables was rejected from the regression models due to collinearity.

In the fully adjusted models, only less education (table 2), a recent income decrease (table 3), and housing of poorer quality (table 4) remained significantly (p<0.05) associated with CMD after adjustment for all the other socioeconomic status variables. Adding income drop to a model that included per capita income but did not include any other socioeconomic...
variables did not eliminate the statistically significant association found between income and CMD. However, it was with the addition of other socioeconomic variables to the same model that this statistically significant association between income and CMD disappeared. Different analyses were carried out using per capita income grouped in 10, 6, and 4 categories and education in 6, 5, and 4 categories. Virtually the same pattern of results emerged from these analyses with education, income drop, and poorer housing quality showing the only statistically significant associations (p<0.05) with CMD.

We found no statistically significant interactions (p<0.05) between education, income, and income decrease.

**DISCUSSION**

In keeping with most previous studies from developed as well as developing countries we found a higher prevalence of CMD among the most socially disadvantaged groups defined according to a variety of socioeconomic indicators. Nevertheless, only a recent income drop, lower level of education, and poorer housing quality showed independent and statistically significant associations with an increased prevalence of CMD after adjusting for other explanatory variables. In contrast with findings from other studies in Europe and America, income was not associated with the prevalence of CMD after adjusting for other socioeconomic variables. Some factors, such as acute financial strain, might have a more universal impact on mental health; whereas others, such as income or education, might be more country specific. The cross sectional design of this study precludes us from reaching any conclusions about the direction of causality.

Chile is one of the 10 countries in the world with the most unequal distribution of income. Within these 10 countries, Chile has the smallest proportion of people living on less than US$1/day and one of the highest levels of education. Over the past four decades successive governments have declared education is essential to eradicate poverty. Parents have to ensure their children receive at least eight years of basic education, which is free in state schools. Thus illiteracy is fairly uncommon, especially among younger generations. Nevertheless access to higher education is increasingly more difficult for the poor. Poor families incur large debts to pay for a university degree that no longer gives certainty of a high paid job in the future.

Our results showed a robust, dose-response, and independent inverse association between education and CMD. Other Latin America studies had shown similar findings but reaching conclusions was limited because of methodological problems, such as the use of questionnaires rather than interviews, limitations in the sampling, poor response rates, and, most importantly, unadjusted results. With this study most of these problems were successfully overcome. A British study, which used similar methodology to ours, did not find a statistically significant association between education and CMD after adjustment for other socioeconomic measures. It is unclear why education could be more closely associated with the prevalence of CMD in a poor Latin American country rather than in the UK. The lack of association between CMD and education in the British sample could have arisen because most people achieved a higher level of education and the sample might have been too homogeneous to detect any differences. However, this did not seem to be the case, the British sample had a reasonable spread among categories denoting different educational levels according to qualifications achieved (no qualifications, GCSE, and A levels). The educational level in the Chilean study was measured according to whether people had completed primary, secondary, or higher education. Thus the way we estimated educational achievement was different but we think both represented a similar concept, namely the educational level achieved by people. A number of explanations could account for this association between CMD and education in poorer Latin American countries.

Poor education could be a marker for childhood adversity, which could hypothetically be more common among Chileans and play a fundamental part in determining susceptibility to depression or other diseases later in life. It would have been useful to have more precise information on early life circumstances but this kind of information when collected retrospectively is usually affected by a strong recall bias. Thus, a longitudinal design would be required to test this hypothesis. Education could be perceived or objectively represent a more sensitive indicator of lack of opportunity or low social position in developing rather than developed countries. It is probable that educational attainment is more closely related to CMD in the UK and USA found clear and significant independent associations between income and CMD, whereas in Brazil, another middle income country with great inequalities, the association between CMD and income drop, was weaker and after adjustment for other socioeconomic variables. In Brazil, another middle income country with great inequalities, the association between CMD and income decrease was also weaker after adjustment for other socioeconomic variables. It is possible that those people with higher income could have under-reported their income and this would have attenuated the differences in the prevalence of CMD between various income groups. However, our measures of income and education showed great resemblance with the official government statistics and we have no other reason to believe that a large measuring bias might have been introduced. We used different ways of grouping the income and education variables but our results remained unchanged. Again adjusting for “working status” did not influence greatly the association between income and CMD, even though lower income usually increases psychological distress and it is more common among the unemployed or unable to work for other reasons. The absence of a large confounding effect of “working status” on the association between income and CMD can be partly explained by the small proportion of people who were unemployed or out of work reporting their income and this would have attenuated the association.
unable to work. Furthermore, we tested for interactions between income and income decrease with “working status” but none were statistically significant (p<0.05). Nevertheless, the association of material factors, wealth, and CMD seems to have been captured better by the quality of the housing, a variable that might represent a more stable picture of wealth than income.

Far too often research findings on the association between health and socioeconomic indicators in the western world are extrapolated to poorer countries as if no important differences exist. However, these socioeconomic indicators represent complex concepts, subject to cultural as well as social influences. We invite readers to be more cautious when using these socioeconomic indicators outside the western world. Understanding the impact of socioeconomic factors on mental health, or indeed any other health outcome, requires research in poor as well as rich countries.

ACKNOWLEDGEMENTS
We would like to thank all the interviewers who participated in this study. Also our thanks to Dr David Gunnell, Social Medicine Department, University of Bristol, for reading this paper and providing some useful comments.

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Funding: This study was funded by the European Community (EC).

Conflicts of interest: none.

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