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# Measuring the level of social support using latent class analysis



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#### ABSTRACT

Different instruments have been used to measure social support in epidemiological studies of which the most widely used is the Medical Outcomes Study Social Support Scale (SSS-MOS). However, these studies lack measures of the level of social support on health risks. We used latent class analysis (LCA) to distinguish subgroups with different levels of perceived social support and tested the consistency of these subgroups by their associations with the prevalence of Common Mental Disorders (CMD). This is a cross-sectional study of 1013 mothers living in the city of Salvador, Brazil in which psychosocial data were collected through home visits using the SSS-MOS and the Self Reporting Questionnaire-20. For each dimension of social support analysed here, we selected models with two classes using LCA. Multivariate logistic regression models were used to estimate the association between participants' perceived social support and the prevalence of CMD to verify the consistency of the groups defined by LCA. There was a clear difference in the reporting of perceived social support between those classified as high or low using LCA. The probability of perceiving several types of social support was lower in the subgroup classified as low level of social support (13.7-59.8%), and it was much higher in the group classified as high level of social support (84.3-98%). A greater prevalence of CMD was found among mothers with lower levels of social support. LCA seems to be a useful tool to improve measurement of perceived social support by separation into two levels in which the lower level is associated with an increased prevalence of CMD.

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# 1. Introduction

Social support refers to the provision of psychological and material resources by a social network, aimed at improving an individual's ability to cope with stress (Cohen, 2004). It is an important predictor of physical and psychological health and well-being and plays a significant causal role in chronic non-communicable and infectious diseases, as well as in their treatment and rehabilitation (Cohen and Syme, 1985; Cohen, 2004; Uchino et al., 2012). Social support is also associated with reductions in cancer mortality rates (Pinquart and Duberstein, 2010), a reduced risk of coronary heart disease (Uchino et al., 1996), improved prognosis and survival for cardiovascular and brain diseases (Berkman and Glass, 2000), protection

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against depression and anxiety (Kawachi and Berkman, 2001; Stansfeld, 2002; Cohen, 2004), and an improved ability to cope with stress resulting from chronic illness (Stansfeld, 2002).

There is evidence that social support and social ties benefit an individual's mental health and psychological well-being (Turner and Marino, 1994; Kawachi and Berkman, 2001). Several studies have emphasized the effect of social support on reducing the symptoms of depression, anxiety, social phobia, and Common Mental Disorders (CMD) (Costa and Ludemir, 2005; Maulik et al., 2009; Moak and Agrawal, 2009; Kim, 2010). The positive effects of social support on mental health may be observed at any age, independent of the source that provides it, as long as the individual perceives him/herself to be socially integrated and supported (Cornwell and Waite, 2009a).

Over the years, increasing efforts have been made to achieve better measures of social support, disentangling a set of dimensions, which include informational support, tangible help, emotional support and integration (Krause and Markides, 1990), social disconnectedness and perceived isolation (Cornwell and Waite, 2009b), and social health (Cella et al., 2007). However, debate about the quality of instruments that measure social support is minimal (Uchino et al., 1996; Gonçalves et al., 2011). Few studies have analysed the quality of social support measures used in survey research (Sherbourne and Stewart, 1991; Berkman and Glass, 2000; Griep et al., 2005; Gonçalves et al., 2011; Uchino et al., 2012) and when such studies have been conducted, analysis has been restricted to an evaluation of the instrument's dimensionality, and occasionally to aspects related to construct validity (Sherbourne and Stewart, 1991; Griep et al., 2005; Gonçalves et al., 2011).

Furthermore, studies do not adequately respond to issues related to the choice of cut-off points in quantitative scales with different levels of risks for health or behavioural outcomes (Uchino et al., 1996; Gonçalves et al., 2011). This hampers recognition of specific characteristics of any given subgroup that could be at a greater risk of health problems (Corin, 1994). A more refined analysis of social support measures is necessary to improve recognition of their impact on health.

One of the most widely used instruments in epidemiological research is the perceived social support scale (SSS-MOS) developed for the Medical Outcomes Study (Sherbourne and Stewart, 1991; Berkman and Glass, 2000; Griep et al., 2005). This scale was constructed at the end of the 1980s and was designed to cover five dimensions of social support, such as emotional support (expressions of positive affect, empathy, care and trust); affective support (expressions of love and affection); material support (provision of material help); informational support (provision of information that helps the individual to deal with difficulties); and positive social interaction (the availability of company for leisure activities) (Sherbourne and Stewart, 1991).

The scale is considered an adequate measure of perceived social support (Berkman and Glass, 2000), and it has been adapted and validated for the Brazilian population (Griep et al., 2005). Several studies on the SSS-MOS conducted in different countries (Sherbourne and Stewart, 1991; Mahmud et al., 2004; Griep et al., 2005; Espínola et al., 2007; Robitaille et al., 2011) focus on presenting its psychometric properties, but nobody has yet used it to provide a clear definition for cut-off points in the original scale for the identification of levels of perceived social support associated with different risks to mental or physical health. Instead of choosing a cut-off point for the sum of responses for each item, some studies use a continuous score to represent social support (Aflakseir, 2010), while others classify scores below the median (Griep et al., 2005) or the first tertile (Andrade et al., 2005; Pinto et al., 2006) as representative of a low level of perceived social support. There are also studies that merely provide the chosen score for cut-off – 75 points (Hasselmann et al., 2008), 57 points (Salinero-Fort et al., 2011), 33 points (Costa and Ludemir, 2005) – with no explanation about how the authors reached these scores or what they represent. Thus, the criteria for classifying the scores of the SSS-MOS and characterizing levels of perceived social support seems to be somewhat arbitrary.

From this perspective, latent class analysis (LCA) is a useful statistical technique for clustering individuals into subtypes within a population when there is no prior knowledge about which individual belongs to which subpopulation. This method is used to analyse multivariate categorical data and model associations between observed variables that provide an imperfect measure of a non-observable (latent) variable. The discrimination of subgroups (also called classes) arises from an evaluation of response patterns and the probabilities associated with each of the observed variables. This allows us, for example, to identify a group of subjects with a high probability of reporting low levels of social support, according to their responses to items on a scale (Collins and Lanza, 2010).

LCA is increasingly used in public health studies to identify metabolic syndrome (Boyko et al., 2012), immunological patterns (Figueiredo et al., 2013) and incidence of respiratory symptoms in children (Spycher et al., 2008), and has proven to be an efficient method for disentangling groups and creating more homogeneous subgroups. In relation to instrument consistency, LCA has been used to identify the subtypes of Antisocial Personality Disorder (Bucholz et al., 2000), Attention Deficit/Hyperactivity Disorder (Rasmussen et al., 2002), and patterns of problem behaviour (Lang et al., 2006). These studies aim to improve diagnostic validity by relying on empirical evidence, and moving towards more valid and useful classification criteria.

Working with a large sample of women to whom the Medical Outcomes Study Social Support Scale was applied, this study aims to distinguish subgroups with different levels of perceived social support by using the LCA technique. We also seek to analyze the agreement between the classifications as defined by an analysis of response patterns using LCA, compared to that obtained using the median as a cut-off point of the original score, as used by many investigators. Finally, we test whether the different levels of perceived social support identified here are associated with prevalence of Common Mental Disorders (CMD).

#### 2. Methods

### 2.1. Study design and participants

This is a cross-sectional study of 1013 mothers of children aged between 4 and 12 years old, part of a population cohort in the city of Salvador. Participants were randomly selected from a set of 20,000 residences within the city of varying economic levels and environmental conditions, organized into 24 geographic areas (Barreto et al., 2006). Of the 1445 families initially recruited for the study, 259 were excluded because a guardian who was not the child's biological mother answered their psychosocial questionnaires. Moreover, it was not possible to locate 55 families due to changes of address, while 118 questionnaires lacked sufficient data, leaving 1013 mothers to form the sample used in this work. As this work is part of a series of previous publications (Barreto et al., 2006), to maintain consistency in the subpopulation studied in the project, we only selected the mother's dataset, on the assumption that the mother's characteristics have a strong influence on the child's health status.

### 2.2. The Medical Outcomes Study Social Support Scale

This scale is composed of 19 items with positive descriptions and responses on a five-point scale, where 1 = never, 2 = rarely, 3 = sometimes, 4 = almost always and 5 = always, to identify the frequency with which social support is perceived (Sherbourne and Stewart, 1991). The score was calculated for each dimension through the sum of points attributed to the items divided by the highest possible score for each dimension. To standardize the score of social support scale, the results of the ratio (total points obtained/high score for the dimension) were multiplied by 100.

The scale has been adapted and validated in Brazil and presents high internal consistency levels and moderate item-scale correlation, enabling it to be used as a measure of perceived social support (Griep et al., 2005). Although the original scale was designed to cover five social support dimensions, only three dimensions have been identified in the Brazilian population: positive social interaction/affective support; emotional/information support; and material support (Griep et al., 2005).

Given these differences, we decided to identify scale dimensionality using exploratory factor analysis with the principal axis method for the extraction of factors and varimax orthogonal rotation. Following another Brazilian study (Griep et al., 2005), the items were grouped according to the three factors identified in our study: (1) Affective support and positive social interaction, including all the items of these sub-scales and adding an item from the emotional support sub-scale (someone who understands your problems); (2) Emotional and informational support, which involved all the questions from both sub-scales, except the one included in the first factor; and (3) Material support, for which we maintained the four items from the original scale. The goodness of fit indices demonstrated the adequacy of analysed data (KMO = .953; Bartlett's test of sphericity with a p value = .000).

### 2.3. Instruments used to identify LCA group consistency

The Self Reporting Questionnaire (SRQ-20) was used to identify mothers with suspected Common Mental Disorders (CMD). This inventory was developed by the World Health Organization and has been validated for the Brazilian population (Mari and Williams, 1986; Gonçalves et al., 2008). It is composed of 20 dichotomous (Yes/No) questions which enquire about the presence or absence of symptoms suggestive of the following disorders: depression; dysthymia; phobia/anxiety; somatisation and neurasthenia. Compared to the diagnostic psychiatric interview, Brazilian studies revealed high sensitivity (85%) and specificity (80%) for this instrument, and indicated that a cut-off point of 8 or more symptoms could be used to identify participants with suspected CMDs (Mari and Williams, 1986; Gonçalves et al., 2008).

### 2.4. Statistical analyses

LCA is one of the most widely used latent class models for categorical data and serves to cluster individuals into mutually exclusive classes that represent groups of individuals who are similar to each other, based on a set of observed categorical variables. The model estimates the smallest number of groups (latent classes) that describe the variability of response patterns within the population. Model interpretation is based on the estimated probabilities of membership of each class (class prevalence) and on the probabilities of a given response for the observed variables conditional on the class to which he/she belongs. Interpretation of the classes is carried out by comparing the probability of each response in each class relative to the overall population (Collins and Lanza, 2010).

The evaluation of the model's goodness of fit was achieved through entropy and the G-Lo-Mendell-Rubin likelihood ratio test (Collins and Lanza, 2010). The entropy index measures the quality of the resulting classification in terms of the separation of the latent classes. Its values range from 0 to 1, where high values (>0.90) indicate that the latent classes are highly discriminative.

In order to fit the latent variable models, the instrument items were organized according to the dimensions identified in the exploratory factor analysis. The responses were then dichotomized so that the response options "never", "rarely" and "sometimes" formed a new category indicative of a lower perception of social support; while the options "almost always"

and "always" formed another category which was indicative of a higher perception of social support. This re-arrangement was important in providing response categories with frequencies higher than 10% and improved our capacity to interpret the LCA findings. The frequency of answers "never", "rarely" and "sometimes" varied between 46.1% and 20.8%, which demonstrates that most answers were "almost always" and "always" options.

For each of the three dimensions of social support, we selected models with two latent classes. We used the same methodology to estimate the subgroups for the global social support scale. Conditional probabilities are presented for the final models; these are the probabilities of reporting specific social support items conditional on membership of that particular latent class (denoting level of perceived social support). These analyses were implemented using Mplus version 5 software (Muthèn and Muthèn, 1998–2007). Estimation in MPlus is obtained via the EM algorithm.

Furthermore, we used the original score from the social support scale to compute its median. Scoring below or above the median of the original scores on the scale is referred to in the literature as a cut-off point to classify individuals into low and high level social support groups. The Kappa index was used to measure agreement between the two different forms of classification based on the medians of the original scores and on the groups discriminated by LCA.

The consistency of the social support subgroups was also evaluated using multivariate logistic regression analysis to estimate the association between social support and CMD prevalence, adjusting for socio-demographic indicators, such as income, schooling, child's age, paved roads, piped water and access to the sewerage system. Odds ratios (ORs) and corresponding 95% confidence intervals were provided using STATA software program version 10.0.

#### 2.5. Ethical considerations

The study was approved in 2005 by the National Ethics Committee, under registration number 047-05/CEP-ISC FR-78168. Participants signed a Free and Informed Consent Form, which contained detailed information about the data collection procedures for the research. Participants were informed of any results that suggested a need for medical/psychological care and were provided with relevant information about the health service.

#### 3. Results

### 3.1. Sample description

The participants' average age was 32.2 years (SD 6.73), most had only one child (58.1%), were of mixed race (66.6%), had not completed secondary education (69.9%) and had an income of less than one minimum wage per month (54.3%). There were 9.1% of participants living without piped water, 16.9% did not have access to the sewerage system and 64.8% did not have access to paved roads. The prevalence of suspected CMD was 37.7%.

#### 3.2. LCA results

For the material support dimension, we observed that, among the participants classified as having high social support levels, the probability of perceiving this type of support in any of the items ranged from 85.6% to 93.6%. On the other hand, amongst those who demonstrated less perception of material support, the probability of such support ranged from 17.3% to 46.4% (Table 1).

A similar pattern was observed for the emotional and informational support sub-scales, where the probability of perceiving this type of support amongst participants classified with high levels ranged from 87.4% to 92.8%. On the other hand, for those classified as having lower perceptions of social support, there was a lower probability of such responses (from 20.3% to 43.7%). Furthermore, for the items that made up the dimension of affective support and positive social interaction, the probability of referring to this type of support ranged from 87.1% to 98% and from 20% to 59.8%, amongst women classified in the profiles of higher and lower perceptions of support, respectively (Table 1).

We observed the same pattern in the global social support analysis. We identified higher probabilities of perceiving social support amongst participants classified with high levels of global social support. These probabilities ranged from 87.2% to 98.4%. On the other hand, for those classified as having lower levels of global social support, the probabilities of perceiving social support were lower, ranging from 25.5% to 67.9%.

Despite the majority of participants being classified as perceiving high levels of social support, groups of women with both low and high levels of social support, were characterised for all dimensions analyzed. The entropy values for the three types of identified support and for the global social support scale were above 0.732, indicating adequate goodness of fit (Table 2).

# 3.3. Results of the evaluation of social support profile consistency

The results of the analysis of agreement between subgroups created by LCA and those created using the median as cut-off point are presented in Table 3. We observed that 100% of the participants classified as having a low level of social support using LCA were also classified as having a low level of social support using the median as a cut-off point, for each dimension

**Table 1**Description of levels of perceived social support using latent class analysis, according to three dimensions of support (material, emotional and information, and affective and positive social interaction) (*N* = 1013).

	Overall N(%)	High (%)	Low (%)
Material social support			
Availability of someone to:			
Help if confined to bed	765 (75.5)	91.4	46.4
Take to doctor	678 (66.9)	85.6	32.6
Prepare meals	636 (62.8)	87.5	17.3
Help with daily chores	710 (70.1)	93.6	27.0
Emotional and informational social support			
Availability of someone to:			
Listen to you	730 (72.1)	91.3	41.7
Give you good advice	747 (73.7)	92.8	43.7
Give you information	705 (69.6)	91.1	35.7
Confide in	710 (70.1)	92.0	35.5
Give advice you really want	698 (68.9)	92.5	31.7
Share worries with	624 (61.6)	87.4	20.9
Turn to for suggestions	650 (64.2)	92.0	20.3
Positive social interaction and affective support			
Availability of someone to:			
Show love and affection	865 (85.4)	98.0	59.8
Have good time with	723 (71.4)	87.8	38.1
Hug you	789 (77.9)	92.5	48.2
Get together for relaxation	671 (66.2)	87.1	23.9
Help you to get your mind off things	702 (69.3)	89.3	28.6
Do something enjoyable with	693 (68.4)	91.4	21.7
Understand your problems	639 (63.1)	84.3	20.0
Love you	809 (79.9)	94.9	49.3

**Table 2** Social support latent classes for different dimensions of perceived social support (N = 1013).

Social support latent classes	N	%	Entropy®
Material social support			0.732
High	667	65.8	
Low	346	34.2	
Emotional and information social support			0.858
High	627	61.9	
Low	386	38.1	
Affective social support and positive social interaction			0.844
High	671	66.2	
Low	342	33.8	
Global social support			0.925
High	581	57.3	
Low	433	42.7	

<sup>\*</sup> Model fit index.

 Table 3

 Levels of agreement between the subgroups of social support, classified according to LCA technique and using the median as cut-off point (N = 1013).

Subgroups created using LCA technique	Subgroups created using median as cut-off point		Overall% of agreement	Kappa
Material support	High N (%)	Low N (%)	83.6	0.67*
High	501 (75.1)	166 (24.9)		
Low	0	346 (100)		
Emotional and informational support	High N (%)	Low N (%)	90.9	0.82*
High	535 (85.3)	92 (14.7)		
Low	0	386 (100)		
Positive social interaction and affective support	High N (%)	Low N (%)	78.4	0.58*
High	452 (67.4)	219 (32.6)		
Low	0	342 (100)		
Global social support	High N (%)	Low N (%)	90.2	0.81*
High	491 (98.2)	90 (17.5)		
Low	9 (1.8)	423 (82.5)		

<sup>\*</sup> p < 0.001.

**Table 4**Crude and adjusted associations between participant perceived social support level as identified by LCA and suspected Common Mental Disorder (CMD) (*N* = 1013).

Variables	CMD Suspected N (%)	OR crude (CI95%)	OR adj <sup>*</sup> (CI95%)
Material support			
High	233 (34.9)	1.00	1.00
Low	149 (43.1)	1.41 (1.08-1.84)	1.38 (1.06-1.81)
Emotional and informat	tional support	,	,
High	205 (32.7)	1.00	1.00
Low	177 (45.8)	1.74 (1.34-2.26)	1.77 (1.35-2.31)
Positive social interaction	on and affective support	, ,	, ,
High	219 (32.6)	1.00	1.00
Low	163 (47.7)	1.88 (1.44-2.45)	1.85 (1.41-2.43)
Global social support	, ,	, ,	, ,
High	187 (32.2)	1.00	1.00
Low	195 (45.1)	1.73 (1.34-2.24)	1.73 (1.33-2.26)

<sup>\*</sup> OR adjusted for income, schooling, age of the children, paved roads, piped water and access to the sewerage system.

analyzed separately. However, 1.8% of the participants classified as having a low level of global social support using LCA were classified as having a high level of social support using the median as a cut-off point.

The agreement in the subgroups of higher levels of social support was smaller compared to agreement in the lower levels of social support, although most participants were equally classified using both techniques for all dimensions. Nevertheless, the overall percentage of agreement was high for all dimensions, varying between 78.4% for positive social interaction and affective support, to 90.9% for emotional and informational support. Generally, Kappa values were statistically significant and indicated substantial agreement, except for positive social interaction and affective support, where the Kappa indicated moderate agreement (Table 3).

We also observed that those participants classified as having lower levels of social support across all dimensions had a greater chance of presenting CMD, compared to those classified as having higher levels of support. This increased chance of CMD ranged from 38%, for those with lower perceptions of material support, to 86% for those with lower levels of affective support and positive social interaction. (Table 4).

# 4. Discussion

This paper presents an application of the LCA technique for the classification of groups according to perceived levels of social support in order to identify more accurate levels of social support, which represents risks to health. Two groups of high and low social support, respectively, were identified for all the three dimensions of social support investigated. There was a clear pattern in the probability of reporting social support stated in the two subgroups; a lower probability of reporting in the group labelled low level of social support, and a much higher one for the subgroup characterised with a high level of social support. Further, higher prevalence of CMD was observed amongst those participants with low levels of social support, as identified by LCA.

The application of LCA to distinguish levels of social support goes beyond other techniques because it precisely identifies the contribution of each indicator to the frame of the subgroups created. This means that, in addition to classifying the individuals according to the level of social support, one can identify the array of observed variables that represent each dimension of social support and characterize what is called high or low level of social support (Collins and Lanza, 2010).

Although we cannot discard the use of any variable-based method (as median) to define the level of response to SSS-MOS, based on our findings, we strongly recommend the use of the LCA technique to identify patterns of social support perception. This technique enables one to have precise and theory-based classes of social support extracted from SSS-MOS. This is especially relevant because social support is a complex construct that cannot be directly observed and has potentially different effects on health (Berkman and Glass, 2000; Kawachi and Berkman, 2001; Stansfeld, 2002). There is evidence that material and informational support can have a strong effect on individual physical health, since it can improve the access to resource and tangible aids (Berkman and Glass, 2000). On the other hand, emotional support may help the individual to deal effectively with stressful situations, preventing a cascade of negative emotional and behavioural responses, and the subsequent repercussions on individual psychological status and mental health (Kawachi and Berkman, 2001).

Investigators are increasingly using latent class analysis as an improved approach to identify underlying subgroups of individuals who share important characteristics and behaviours (Bucholz et al., 2000; Rasmussen et al., 2002; Lang et al., 2006). When using categorical data in this approach, one advantage is not have to make assumptions about the distributions of indicators, except in relation to local independence; in other words, there is an assumption that within a latent class the indicators are independent. This approach may be useful not only when data can only be captured categorically, but also may qualitatively describe how the groups are different, based on their responses.

However, there are limitations to using LCA arising from sample size requirements. LCA requires a sufficient sample size because of sparse cells, or small cell counts in the contingency table, which may lead to unstable results (Collins and Lanza, 2010). This is one reason for our decision to dichotomize the variables for each item on the social support scale. In order to consider the five categories of each item, a larger sample size is required to avoid unreliable results. The model's results focus on a description of estimated probabilities and one major aspect of an assessment of the model fit is whether the model is supported by related theory and whether the classes may be interpreted as meaningful. A model should not only be selected because it fits the data according to a certain statistical criterion, but due to the interpretability of the parameters, which is related to model plausibility.

To our knowledge, this is the first study of its type, which is why we verified the consistency of the social support groups created using LCA. We found general agreement between the classification created by LCA and that using the median as cut-off point for the original score. Furthermore, for all the social support dimensions analysed, participants with lower social support levels presented a higher prevalence of CMD, corroborating other studies that have reinforced the relationship between social integration and mental health (Kawachi and Berkman, 2001; Costa and Ludemir, 2005; Cornwell and Waite, 2009; Moak and Agrawal, 2009; Kim, 2010). Although we cannot generalize these results beyond the mental health domain, this method may be useful for other health domains (i.e. physical health) not tested in the current study.

We should emphasize that social support is transactional in nature; it involves reciprocity and is constructed over the life course. Not all social ties are supportive, and even with those that are, the support offered varies in type, frequency and intensity (Berkman et al., 2000; Stansfeld, 2002). So, contexts reflecting diverse life course stages, socioeconomic status and culture may therefore affect the thresholds found in our study, which limits our findings.

Another limitation refers to the cross-sectional nature of the design, which does not allow us to confirm the temporality of the relationships established here. However, our study did not seek to identify determinant factors for CMD; the estimated associations were relevant for a verification of the consistency of the subgroups distinguished by LCA and defining different levels of social support. Further, despite the use of validated questionnaires it is known that measurements of social support and mental health may be subject to respondent bias, since anxious or depressed individuals, with low incomes and with less access to social and health services tend to treat reality with less optimism and probably also perceive the level of social support around him/her lower than effectively it is (Kawachi and Berkman, 2001; Stansfeld, 2002).

The application of LCA in our work provides a useful identification and characterization of different levels of social support, allowing this population to be described according to a level of social support that may represent health risks. The refinement of social support measurements achieved here provides a better understanding of the social determinants of health, as well as theoretical underpinnings for the formulation of health care policies, enabling a new approach to the health/disease/care process. From the health promotion point of view, it enables one to identify groups exposed to low levels of social support and thus direct policy-related efforts to improving the social ties of such groups, ameliorating their state of health, including common mental disorders.

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#### Appendix A. Supplementary material

Supplementary data associated with this article can be found, in the online version, at http://dx.doi.org/10.1016/j.ssresearch.2014.11.009.

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