1	Cardboard floor: About the barriers for social progression and their impact on the
2	representativeness of epidemiological studies.
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23 Abstract (111 words)

The most disadvantaged extreme of the social continuum is usually underrepresented in 24 epidemiological studies. We discuss the consequences of excluding this segment of the 25 population and suggest different approaches for addressing this issue. In particular, we 26 27 describe/analyse a barrier that tends to perpetuates people in the most disadvantaged 28 extreme of the social continuum, hereinafter referred to as the "cardboard floor". Besides, we propose different approaches to accessing to the least favoured, segment in order to 29 study the cardboard floor. The adoption of these strategies could help to visualize this 30 barrier, allowing to better monitoring social mobility and their expected health 31 32 improvements, as well as increasing the representativity of population health studies.

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A continuum of socioeconomic status ranging from the least to the most privileged 34 35 persons is evidenced in population studies, with profound implications for health and care [1]. Individuals in the most disadvantaged social group suffer from extreme poverty and 36 face several specific challenges to their health and health care [2]. They frequently cannot 37 meet their most basic needs (including their physiological needs, most acutely 38 exemplified by homelessness) and are at a higher risk of health problems and accelerated 39 40 aging due to unhealthy habits (e.g. unhealthy diet and drug consumption), harmful environmental and biological factors, and social isolation [1–4]. As a result, the most 41 socially disadvantaged persons have higher rates of premature mortality, especially 42 43 caused by suicide and violence, and higher prevalence of all types of diseases, particularly infectious diseases and mental disorders [2,5]. Besides, care for chronic conditions is 44 compromised for this population group, which relies to a substantial degree on emergency 45 46 care, particularly in health systems that do not guarantee universal health coverage [5].

Even considering the relative size of the most deprived extreme of the social continuum 47 48 (e.g. about 0.5% of the U.K. adult population in 2018 was considered homeless) [6], the scale of unmet health and health care needs would imply that improving their social 49 mobility might have a significant impact on the overall health status of the population. 50 51 However, several barriers significantly hinder this upward mobility. If a glass ceiling is used as a metaphor for the barrier to higher achievement, success, or recognition for 52 individuals of certain groups within different careers or industries (e.g. women becoming 53 CEOs), an even more appropriate one in this case would be a cardboard floor, making 54 55 reference to the surface that is a daily experience for many extremely deprived people [7]. 56 Studying the impact of this barrier on health, could help to understand it better, hopefully favouring social mobility. Conducting such studies, however, is not exempt from 57

difficulties, one which being particularly relevant: the lack of access to data from personsin the least favoured extreme of the social continuum.

The most disadvantaged group is very unlikely to be included in research and, as a result, is usually inadequately represented in health studies. This recruitment bias has important implications [3,4,8]. It limits the representativeness and external validity of surveys and population health studies and, furthermore, results in underestimation of the health risks, morbidity and mortality across the entire population. Importantly, it also hides the true scope of the specific issues affecting this group from researchers, policy makers and the public.

Different approaches focusing on improved sampling strategies to guarantee the representation of this group in population studies could be used. [3,8,9]. Proposed complementary strategies include: assigning greater sampling weights to individuals in this group, targeted over-recruitment, and/or intensifying fieldwork in marginal areas or suburbs through involvement of social organisations at local level. Nevertheless, these methods require some a priori knowledge of the number of people in this situation when defining the reference population for a specific study.

The use of data from administrative data and Electronic Health Records (EHR), such as 74 75 the Medicaid claims data in the U.S. and the Clinical Practice Research Datalink (CPRD) in the U.K. [9,10], could also be a suitable way to access to the most socially 76 77 disadvantaged persons. Relevant health and healthcare information for this population are often registered within these data sources. Some limitation of this data needs to be 78 79 acknowledged in relation to their completeness, and ability to capture circumstances of 80 maximum vulnerability and the inclusion of information on key mediating mechanisms relevant to determine biological, behavioural, and psychosocial pathways. However, such 81 data also have strengths: they are in many cases mandatory, population-wide and usually 82

contain relevant information on different health outcomes, such as mortality or hospital 83 84 admissions. Besides, most such data are potentially linkable to other relevant datasets for the study of this population (e.g. social care or demographic records) bringing together 85 their strengths and, in some cases, allowing to overcome the abovementioned limitations 86 [9,11]. Hence, the use of linked data from EHR could be a suitable way to capture relevant 87 aspects of the most socially vulnerable individuals and, furthermore, might represent an 88 89 adequate approach to obtain valid and reliable estimations of the health status in this part of the population. In addition linked EHR data, would allow estimating the relative 90 numbers of the most disadvantaged group, providing relevant additional information on 91 92 morbidity and outcomes and facilitating the implementation of improved sampling strategies [9]. 93

The access to the most disadvantaged extreme segment of the social continuum remains 94 a challenge for population health studies. Using a combination of approaches based on 95 96 the use of HER linked data and strengthening the sampling strategy for the specific 97 studies, might be a synergistic way to improve the validity of population health 98 estimations. The adoption of these strategies could help to visualize the barriers for social mobility and the access to the most disadvantaged social groups. This will help to better 99 100 understand the phenomena that perpetuate the cardboard floor and to shape care systems 101 that truly "do not leave one behind" [12].

102 Contributorship Statement

All authors (JA-T, JMV, FGV and JA) were involved in all phases of the development of
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106 **Competing interests**

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116	References			
117	1	World Health Organization. Social Determinants of Health: Key Concepts. Who		
118		2014.		
119	2	Fazel S, Geddes JR, Kushel M. The health of homeless people in high-income		
120		countries: descriptive epidemiology, health consequences, and clinical and policy		
121		recommendations. Lancet 2014;384:1529-40. doi:10.1016/S0140-		
122		6736(14)61132-6		
123	3	Burnam MA, Koegel P. Methodology for Obtaining a Representative Sample of		
124		Homeless Persons. Eval Rev 1988;12:117–52.		
125		doi:10.1177/0193841X8801200202		
126	4	Pizzi C, De Stavola B, Merletti F, et al. Sample selection and validity of		
127		exposure-disease association estimates in cohort studies. J Epidemiol Community		
128		Heal 2011;65:407-11. doi:10.1136/jech.2009.107185		

129	5	Koh HK, O'Connell JJ. Improving health care for homeless people. JAMA - J.
130		Am. Med. Assoc. 2016. doi:10.1001/jama.2016.18760
131	6	Ministry of Housing Communities & Local Government. Live tables on
132		homelessness. Homelessness Stat. 2018.
133	7	Baldacci E, de Mello L, Inchauste G. Financial crises, poverty, and income
134		distribution. In: Macroeconomic Policies and Poverty Reduction. 2005.
135		doi:10.4324/9780203005804
136	8	Toro PA, Wolfe SM, Bellavia CW, et al. Obtaining representative samples of
137		homeless persons: A two-city study. J Community Psychol 1999;27:157-77.
138		doi:10.1002/(SICI)1520-6629(199903)27:2<157::AID-JCOP4>3.0.CO;2-2
139	9	Vickery KD, Shippee ND, Bodurtha P, et al. Identifying Homeless Medicaid
140		Enrollees Using Enrollment Addresses. <i>Health Serv Res</i> 2018; 53 :1992–2004.
141		doi:10.1111/1475-6773.12738
142	10	Herrett E, Gallagher AM, Bhaskaran K, et al. Data Resource Profile: Clinical
143		Practice Research Datalink (CPRD). Int J Epidemiol Published Online First:
144		2015. doi:10.1093/ije/dyv098
145	11	Padmanabhan S, Carty L, Cameron E, et al. Approach to record linkage of
146		primary care data from Clinical Practice Research Datalink to other health-related
147		patient data: overview and implications. Eur J Epidemiol Published Online First:
148		2019. doi:10.1007/s10654-018-0442-4
149	12	Griggs D, Stafford-Smith M, Gaffney O, et al. Policy: Sustainable development
150		goals for people and planet. Nature. 2013. doi:10.1038/495305a
151		