1	Clusters of health behaviours in Queensland adults are associated with different
2	socio-demographic characteristics
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28 Abstract

Background: The co-occurrence of unhealthy lifestyles, calls for interventions that target
 multiple health behaviours. This study investigates the clustering of health behaviours and
 examines demographic differences between each cluster.

Methods: 934 adults from Queensland, Australia completed a cross-sectional survey assessing multiple health behaviours. A two-step hierarchical cluster analysis using multiple iterations identified the optimal number of clusters and the subset of distinguishing health behaviour variables. Univariate analyses of variance and chi-squared tests assessed difference in health behaviours by socio-demographic factors and clusters.

37 Results: Three clusters were identified: the 'lower risk' cluster (n=436) reported the healthiest profile and met all public health guidelines. The 'elevated risk' cluster (n=105) 38 39 reported a range of unhealthy behaviours such as excessive alcohol consumption, sitting 40 time, fast-food consumption, smoking, inactivity and a lack of fruit and vegetables. The 'moderate risk behaviour' cluster (n=393) demonstrated some unhealthy behaviours with low 41 42 physical activity levels and poor dietary outcomes. The 'elevated risk' cluster were significantly younger and more socio-economically disadvantaged than both the 'lower and 43 44 moderate risk' clusters.

45 Discussion: Younger people who live in more deprived areas were largely within the
46 'elevated risk' cluster and represent an important population for MHBC interventions given
47 their wide range of unhealthy behaviours.

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49 Keywords: Public Health, Clustering, Health Behaviours, Multiple Health Behaviour

- 50 Change.
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54 Introduction

55 Modifiable health behaviours such as physical inactivity, excessive sedentary behaviour, 56 alcohol, smoking and a poor diet contribute to morbidity and mortality (Lim et al., 2012). 57 These health behaviours are the primary causes of non-communicable diseases, which 58 account for almost 70% of deaths globally and 91% of total deaths in Australia (World Health 59 Organisation, 2014). However, these health behaviours do not occur in isolation (Prochaska 60 and Prochaska, 2011). Therefore modifying one health behaviour in isolation (Busch et al., 61 2013) may not be an adequate health behaviour change strategy.

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Studies in the UK (Poortinga, 2007, Buck and Frosini, 2012) and Australia (Feng and Astell-63 64 Burt, 2013) have demonstrated that unhealthy behaviours co-exist. A study of English adults demonstrated that one in four individuals had three or more health risk behaviours 65 simultaneously (Poortinga, 2007). Similar findings were reported in Australian and Dutch 66 studies that reported clustering at both ends of the risk factor spectrum (all risk factors and 67 68 no risk factors) as well as clustering of smoking with other lifestyle risk behaviours (Schuit et al., 2002). This clustering of unhealthy behaviours may have important implications for 69 70 health promotion, particularly if health behaviours share underlying psychological 71 characteristics and can be changed using the same behaviour change techniques (e.g., self-72 monitoring) (Băban and Crăciun, 2007). To maximise change in behaviours in multiple 73 behaviour change interventions it is necessary to target multiple behaviours with specific 74 behaviour change techniques (Pronk et al., 2004).

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Interventions focusing on multiple health behaviours have demonstrated improved health outcomes compared with single behaviour interventions (Goldstein et al., 2004). However, reaching people through health behaviour change interventions remains challenging. It may be therefore more cost-effective to change multiple behaviours simultaneously once individuals have been reached (Prochaska and Prochaska, 2011, Prochaska et al., 2008).

This is important as more cost-effective health behaviour interventions are needed due to increasing pressures on health services (Australian Government Department of Health and Ageing, 2006). Furthermore, when two behaviours are closely related to one another, intervening on only one behaviour is not likely to generate lasting effects (Busch et al., 2013).

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87 Few studies have explored whether specific clusters of risk factors are more or less 88 prevalent in population subgroups. This is important to better target the right multiple health 89 behaviour change interventions to specific population subgroups. Previous research has 90 demonstrated how a 'one-size-fits-all' approach to health behaviour change is generally ineffective (Ball et al., 2006). A cluster analysis will help to identify individuals who express 91 92 similar health behaviours, therefore identifying potential target populations for health 93 promotion efforts. With these considerations in mind, this study aims to investigate the prevalence of individual health behaviours by socio-demographic factors such as age, 94 95 gender and socio-economic factors. Secondly, it aims to examine the clustering of these health behaviours. Thirdly, it aims to examine any between cluster differences in socio-96 97 demographic factors.

98 Methods

99 Study Population

Data were collected as part of the Queensland Social Survey (QSS) via computer-assisted 100 telephone interviews. The QSS is an omnibus survey of households in the state of 101 Queensland, Australia administered by the Population Research Laboratory at Central 102 Queensland University. For sampling purposes the state of Queensland was delineated into 103 104 two areas for telephone interviewing; first, South-East Queensland and second, the remainder of Queensland. A two-stage selection process was then employed; first, selection 105 106 of households and second, selection of respondent gender within each household. A sample of 1293 Australian adults were reached by randomly selecting households in the state of 107

108 Queensland (Australia) and then quota sampling by gender. Overall, there was variation from the Queensland population from which they were drawn with an over-sampling in the 109 55 and above age categories, and under sampling in the under 35 age categories. The QSS 110 included socio-demographic and health behaviour-related questions (Department of Health 111 and Ageing, 2013). The overall response rate of the QSS was 41.2% (n = 1,293). Ethical 112 approval was obtained through the Central Queensland University research ethics 113 committee and all participants provided informed consent. Data collection took place 114 115 between June and July 2013.

116 Health behaviours

117 <u>Physical activity</u>

Physical activity (PA) data were captured using the Active Australia Survey. The Active 118 Australia Survey is a brief PA questionnaire (Australian Institute of Health and Welfare 119 (AIHW), 2003) demonstrating acceptable validity compared with Actigraph accelerometry 120 121 (r=0.46-0.50) (Helmerhorst et al., 2012). Questions include items on duration and 122 frequency of walking and moderate and vigorous-intensity physical activity in the previous week. All activities had to be performed continuously for at least 10 minutes at a time. Total 123 duration of physical activity (TPA) was calculated using this formula: total walking minutes + 124 moderate activity minutes + (vigorous activity minutes * 2). In line with current public health 125 guidance (Australian Bureau of Statistics, 2013), to meet the physical activity guideline, 150 126 minutes of activity a week over 5 days were needed. A binary outcome was created with 127 ≥150.00 minutes of total activity in 5 or more session classed as meeting the guidelines. 128

129 <u>Sitting time</u>

Sitting time was calculated as the average daily time spent sitting in the past week, as reported on the Workforce Sitting Questionnaire (Chau et al., 2011). On this 10-item measure, participants were asked how much time they spent sitting on non-work and work days while working, commuting, using a computer, watching TV, and during other leisure-

time activities. This measure has demonstrated acceptable reliability (r=0.58, p<0.05) and validity (r=0.48, p<0.01) (Chau et al., 2012). As there is no concrete guideline (Department of Health, 2014); a recent meta-analysis (Chau et al., 2013) was used as a guide to dichotomise sitting time. Accordingly, a binary outcome was defined with sitting time of > 7 hours per day classed as excessive sitting time, given its association with increased risk in all-cause mortality.

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Fruit and vegetable consumption

Fruit and vegetable consumption was assessed by two items used previously in research (Smith et al., 2009b): 'How many serves of vegetables do you eat on a usual day?' and 'How many serves of fruit do you eat on a usual day?' In line with recent research (Australian Bureau of Statistics, 2015, Oyebode et al., 2014) a binary outcome was created based on whether or not participants were meeting the public health guidance recommendation of \geq 5 servings of vegetables and \geq 2 fruit (Department of Health and Ageing, 2013).

147 <u>Fast-food consumption</u>

Fast-food consumption was assessed using one item: "In the last 7 days, how many times 148 did you eat something from a fast-food restaurant like McDonald's, Hungry Jacks, KFC, etc? 149 This also includes other fast-food and takeaway such as fish and chips, Chinese food and 150 pizza." There was little literature to guide a binary outcome for fast-food, therefore a 151 threshold of ≥1 fast-food meal per week equated to not meeting public health guidelines 152 153 (Australian Government Department of Health, 2015). The Australian recommendation is to limit fast food as much as possible, so logically none in the last week is ideal (Australian 154 155 Government Department of Health, 2015). Fast-food consumption was used as a proxy for unhealthy food behaviour as it has been associated with weight gain (Paquet et al., 2010, 156 Thornton et al., 2009, Thornton et al., 2016) and deleterious health outcomes (Pereira et al., 157 2005, An, 2016). 158

159 <u>Smoking</u>

Smoking status was assessed using one item: 'Are you presently a smoker?' (yes/no). In line with current public health recommendations smoking is not recommended (Zwar et al., 2005).

163 <u>Alcohol consumption</u>

Participants were asked: "During the past 30 days did you consume at least one drink of any alcoholic beverage", and if yes, "how many drinks did you have on average each day". An estimate of alcohol consumption was created by multiplying the number of drinks per day in the last month and the average number of drinks per day divided by 30 to give an estimate of drinks per day. Based on established public health guidelines a binary outcome was created based on the guideline of ≥2 drinks per day (Australian Government National Health and Medical Research Council, 2014).

171 <u>Socio-demographic factors</u>

172 Socio-demographic factors measured included age group (18-34, 35-44, 45-54, 55-64, 65+), 173 gender (male, female), level of education (pre-school, primary or high school and college or university), BMI (underweight, healthy weight, overweight and obese) (calculated from self-174 report height (cm) and weight (kg)) and postcode. Post codes were linked to the socio-175 economic index for areas (SEIFA) developed by the Australian Bureau of Statistics 176 (Australian Bureau of Statistics, 2016). As part of this The Index of Relative Disadvantage 177 which ranks community areas in Australia according to relative socio-economic 178 disadvantage using census data on education, employment, occupation, housing and 179 180 English proficiency was used (Australian Bureau of Statistics, 2016). The Index of Relative Disadvantage was then split into four equal quartiles specific to this population (Q1 0-964; 181 Q2 965-1020; Q3 1021-1058; Q4 1059-1129) with a lower score meaning greater 182 disadvantage. 183

184 Statistical Analysis

Only individuals with complete data for all health behaviour and demographic variables were included in the final sample (n=934) resulting in the exclusion of 359 individuals. A sensitivity analysis showed no differences by socio-demographic factors and health behaviours in excluded data (Additional File 1). Prior to cluster analysis, engagement in the health behaviours were presented as simple proportions of those who meet the public health guidelines.

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192 To identify clusters of multiple health behaviours a two-step cluster approach was used 193 which ensures that the size of the distance matrix is determined by the number of preclusters rather than individual cases (Bitman and Gelbard, 2007). Firstly, based on the 194 distance criterion, cases were either assigned to an existing pre-cluster or assigned a new 195 196 pre-cluster. Pre-clusters were then clustered in the second step using the standard hierarchical clustering algorithm, which assessed multiple cluster solutions and automatically 197 determined the optimal number of clusters. The researchers made no assumptions 198 regarding cluster membership or number, as this was determined in the hierarchical two-step 199 200 clustering approach (Bitman and Gelbard, 2007).

201

The cluster analysis was conducted multiple times with multiple iterations undertaken to establish both the optimum number of clusters and the format of health behaviour variables to be included. This process initially included conducting the analysis with continuous behavioural variables and smoker or not, before multiple versions were conducted alternating an additional categorical variable. For example, the inclusion of fast-food as both continuous (servings per week) and categorical (meeting guidelines or not) was explored.

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The final combination of variables included within the cluster was determined using the Schwarz Bayesian Criterion (Bitman and Gelbard, 2007). Based on this analysis, the following variables were included in the final cluster analysis: physical activity, sedentary

212 behaviour, fruit and vegetable intake, fast-food intake, smoking and alcohol consumption. 213 The silhouette measure used to validate the cluster indicated 0.5, which demonstrates a 214 good level of both cohesion and separation, and provided a stronger solution than 215 comparative 2-cluster and 4-cluster alternative solutions. To examine differences between 216 clusters on socio-demographic factors Pearson chi squared was used with independent 217 variables as categorical predictors. For all tests, significance levels were set at p < 0.05. 218 Analyses were undertaken in SPSS v22 (IBM Corporation).

219 3.0 Results

220 **Demographic characteristics**

In the sample of 934 individuals, 495 (53.0%) were male and 439 (47.0%) were female (Table 1). The mean age of participants was 53.79 years (\pm 14.92). Using the index of relative disadvantage, and defining deprivation as living in the top 25% deprived areas (IRD <24 <964), 238 (25.5%) adults were categorised as living in the most deprived areas (Table 1). Most individuals were either college or university educated (64.5% n=602) and 35.0% (n=327) and 23.1% (n=215) were classified as overweight and obese, respectively.

227

228 INSERT TABLE 1 HERE

229

230 Behaviours

231 Physical activity

In total, participants reported an average of 317.72 (SD \pm 322.78) minutes of physical activity per week with 48.9% (n=457/934) categorised as not meeting the physical activity guidelines (Table 1). The percentage of participants meeting physical activity guidelines was not associated with gender (p>0.05), deprivation (p>0.05) or age (p>0.05).

236 <u>Sitting time</u>

A mean value of 413.37 (±193.59) minutes/day of sitting time was reported (≈7 hours/day) with 45.5% (n=425) of people categorised as sitting >7 hours/day. Sitting time was associated with gender (χ^2 [1] = 14.46, *p*<.001), with more males than females exceeding the guideline to sit >7 hours per day (50.5% vs. 39.9%) (*Cramer's V* = 0.124). Sitting time was not associated with deprivation (p>0.05). However, sitting time was lower in those with increasing age (χ^2 [5] = 25.77 *p*<0.001, *Cramer's V* = 0.17). Fewer people reported sitting for >7 hours/day as age increased (18-34 years, 55.8%, >65 years 34.5%).

244 Fruit and vegetable

A daily mean value of 1.95 (±1.27) fruit servings and 2.96 (±1.59) vegetable servings was reported; 86.1% (n=804) of participants were categorised as not meeting the recommended guidelines that combine >2 fruit and >5 vegetables. More females (63.1% vs. 45.1%) than males met the guidelines (χ^2 [1] = 30.46, *p*<0.001, *Cramer's V* = 0.18). There was no association between fruit and vegetable consumption and deprivation (p>0.05) or age (p>0.05).

251 Fast-food

Participants reported consuming an average of 1.61 (±0.79) servings of fast-food in the previous week (44.9% exceeded ≥1 fast-food meal per week). There was a significant association between fast-food consumption and gender (χ^2 [4] = 16.57, *p*<0.01): more males (48.5%) than females (40.8%) consumed ≥1 fast food meal per week. Deprivation was not associated with fast-food intake (p>0.05), whereas age was associated with intake (χ^2 [20] = 120.12, *p*<0.001). The percentage of individuals consuming at least one weekly serving of fast-food was significantly higher for those with a lower age.

259 <u>Smoking</u>

In total, 10.8% (n=101) of individuals smoked. There was no association between smoking (yes/no) and gender (p>0.05). However, there was an association of deprivation and

smoking; residents of the most disadvantaged quartile (Q1) reported higher levels of smoking (38.6% vs. 13.9%) than those in the least disadvantaged quartile (Q4 ; χ^2 [3] = 7.44, p<0.01, *Cramer's V* = 0.12). Smoking was also associated with age (χ^2 [3] = 15.17, p<0.01, *Cramer's V* = 0.13); A higher proportion of younger participants smoked compared to older participants.

267 <u>Alcohol</u>

Participants reported a mean value of 0.83 (±1.75) alcoholic drinks per day. On average, 14.0% of participants exceeded the guideline of ≥2 drinks per day. Alcohol consumption differed by gender (χ^2 [1] = 31.47, *p*<0.001): 20.0% of males and 7.3% of females exceeding alcohol guidelines. There was no association between alcohol and deprivation (p>0.05) or age (p>0.05).

273

274 Clustering of health behaviours

275 Descriptive characteristics of the cluster profiles can be seen in Table 2. Three clusters were identified; the 'lower risk behaviour' cluster included 46.7% (n=436) participants, 42.1% 276 277 (n=393) of participants were in the 'moderate risk behaviour' cluster and 11.2% (n=105) were in 'elevated risk behaviour' cluster. All 'lower risk behaviour' cluster members were non-278 279 smokers (100%) and sufficiently active (100%); they also consumed the most fruit and vegetables (5.34 servings/day) and had the lowest levels of sitting time (400 mins/day). This 280 differs considerably from the 'moderate risk behaviour' cluster which included insufficiently 281 active adults (100%), non-smokers (100%) and poorer dietary behaviours (4.56 servings of 282 283 fruit and vegetables per day). Most individuals in the 'elevated risk behaviours' cluster smoked (96.2%), were insufficiently active (61%), were sedentary for the highest amount of 284 285 time (437 minutes/day) and consumed the least fruit and vegetables per day (4.43 servings), 286 the most fast-food meals per week (1.68 meals) and the most alcohol per day on average 287 (1.83 drinks/day).

288

289 There was a statistically significant difference between clusters for smoking behaviours 290 (F(2,931) = 10,432, p<0.001). A Bonferroni post hoc test showed no difference between 'low risk behaviours' and 'moderate risk behaviours' (p>0.05). However, 'low risk behaviours' 291 292 were significantly different to 'elevated risk behaviours' (p<0.001). 'Moderate risk behaviours' 293 were significantly different to 'elevated risk behaviours'. Physical activity levels were significantly between all three clusters (F(2,931) = 3881.93, p<0.001). Another one-way 294 295 ANOVA (F(2,931) = 20.51, p<0.001) demonstrated drinks per day in 'low risk behaviours' 296 were significantly lower than in the 'elevated risk behaviours', and 'elevated risk behaviours' 297 were significantly higher than 'moderate risk behaviours'. (p<0.05). Despite this, there was no difference between 'low risk behaviours' and 'moderate risk behaviours' (p>0.05). Fruit and 298 vegetable intake also differed significantly between clusters (F(2,931) = 14.38, p<0.001). 299 Although post-hoc tests revealed no different between 'elevated risk behaviours' and 300 301 'moderate risk behaviours' in fruit and vegetable intake (p<0.05), there were statistically significant differences between 'low risk behaviours' and 'elevated risk behaviours' (p<0.001). 302 Moreover, there were also statistically significant differences between 'low risk behaviours' 303 and 'moderate risk behaviours' in fruit and vegetable intake (p<0.001). In contrast there were 304 305 no statistically significant differences for sitting time (F(2,931) = 2.32, p=0.10) or fast-food consumption (F(2,931) = 1.97, p=0.14) between clusters. 306

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308 INSERT TABLE 2 HERE

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Table 3 shows demographic differences between the three clusters. There was no significant gender differences for cluster membership; however, there was a significant effect for cluster membership by age ((χ^2 [8] = 17.95, *p*=0.02, *Cramer's V* = 0.10). The *'elevated risk behaviour'* cluster included significantly younger adults than the *'moderate risk behaviour'* cluster and *'lower risk behaviour'* cluster. Deprivation also differed as a function of cluster membership ((χ^2 [6] = 16.46, *p*=0.01, *Cramer's V* = 0.10). The *'elevated risk behaviour'* cluster was significantly more disadvantaged than the *'moderate risk'* and *'lower risk* *behaviour*' cluster. In contrast, there was no significant difference between clusters on education level (p>0.05).

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320 INSERT TABLE 3 HERE

- 321
- 322 Discussion
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324 Main findings of this study

There is a growing body of literature examining clusters of health behaviours. However, few present findings on a wide range of health behaviours by age, gender and area-level deprivation. Three clusters were identified within this study as '*lower risk'*, '*elevated risk'* and *'moderate risk'* behaviour clusters. The '*elevated risk'* cluster were younger and more socioeconomically disadvantaged. As such, this study is among the first to suggest different clusters of individuals may require different types of multiple health behaviour change (MHBC) interventions.

332

333 What is already known on this topic

This study supports previous evidence that shows how health behaviours are socio-334 demographically related (Buck and Frosini, 2012). Younger and more disadvantaged 335 individuals were more likely to smoke. Although smoking rates continue to decline 336 337 (Australian Institute of Health and Welfare, 2014), Australian data highlights that individuals 338 living in the low socio-economic (SES) areas are three times more likely to smoke (daily) than people within the highest SES areas (19.9% vs. 6.7%) (The Department of Health, 339 340 2015). Similar to previous research, 48.9% of adults were categorised as insufficiently active 341 (Duncan et al., 2012). However, in contrast to previous research (Department of Health, 2014, Feng and Astell-Burt, 2013) physical activity was independent of age, gender and 342 deprivation. Also similar to other population based data, males reported consuming more 343

344 alcohol, fast-food and less fruit and vegetables (Smith et al., 2009a, White, 2013). Finally, females and older participants were more likely to sit less than 7 hours per day. This is in 345 346 contrast to a plethora of previous research which suggests as people age they become more 347 sedentary (Hallal et al., 2012). Evidence of an association between deprivation and sitting 348 time remains equivocal (O'Donoghue et al., 2016). We suggest dissimilarities may occur due to the differences in arbitrary cut-offs used to define higher or lower levels of sitting time. 349 350 Findings within this study show that sitting time was independent of level of disadvantage. 351 This study also highlights how these behaviours converge by demonstrating three unique 352 and distinct clusters of health behaviours.

353

354 What this study adds

In public health terms, the behaviour of those in the 'elevated risk' cluster, who did not meet 355 356 most public health guidelines, is concerning. It is worth noting that while the 'elevated risk' cluster were significantly younger than the other clusters, the 'elevated risk' cluster was 357 balanced towards younger middle-aged people, rather than older middle-aged people. 358 Consistent with previous research (Poortinga, 2007, Schuit et al., 2002) our findings show 359 360 that excessive alcohol consumption, smoking, poor diet and to a lesser extent inactivity were found to cluster together within the 'elevated risk' cluster. A recent review (Noble et al., 361 2015) concluded that males and those with greater social disadvantage showed riskier 362 patterns of health behaviours. However, several of the included studies were poor quality. In 363 contrast, findings in this study showed no differences by gender, potentially as a result of the 364 wider range of behaviours considered (MacArthur et al., 2012). Considering the greater risk 365 posed by each of these behaviours and the likely exacerbated risk associated with engaging 366 in multiple risky behaviours this population subgroup is an obvious target for preventive 367 health initiatives. The high levels of inactivity and poor diet demonstrated within the 368 369 'moderate risk' cluster are also a public health concern.

370

371 Although both 'moderate' and 'elevated risk' clusters are unhealthy and at risk, there are important differences in health behaviours between them. For instance, compared to the 372 373 'moderate risk' cluster the 'elevated risk' cluster consumes more alcohol (1.83 drinks vs. 374 0.65 drinks/day per day), cigarettes (100% vs 3.8% were non-smokers) with differences in 375 physical activity too. In a recent study, deaths for all-cause mortality were advanced by 4.0 years for physically inactive adults. However, the rate advancement period for all-cause 376 377 mortality was 7.9 years among current smokers (Borrell, 2014). Those in the 'moderate risk' 378 cluster may represent very different attitudes and intentions towards health behaviours 379 (Prochaska et al., 2008, Prochaska and Prochaska, 2011). However, such between cluster differences are important and should not be ignored, as these two unhealthy clusters will 380 require different behaviour change interventions. They represent an important opportunity to 381 tailor public health interventions. Interventions have been successful in changing two, three 382 383 or even four or more health behaviours simultaneously, suggesting that MHBC is possible (Hyman et al., 2007). 384

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Due to both the wider range and greater extent of unhealthy behaviours, and perhaps more 386 387 serious consequences in terms of all-cause mortality (Borrell, 2014), individuals within the 'elevated risk' cluster may require a much more intense behaviour change intervention 388 compared to those who are within the 'moderate risk' cluster. In addition to this, given that a 389 dose-response relationship exists between health behaviours and health outcomes - where 390 391 there is the greatest benefit of improving health behaviours of individuals who are the most unhealthy (Lee and Skerrett, 2001) - the overall health of the population may be improved 392 393 most by focusing on those within the 'elevated risk' cluster. In contrast to this, a significant 394 proportion of the 'moderate risk' cluster may only require small changes to health behaviours 395 to meet public health guidelines, which is also a viable option for intervention.

396 *Limitations of this study*

A cross-sectional design remains strong for observational purposes however it does 397 encounter causality limitations. Subjective measures were used throughout the research 398 design and are subject to measurement error due to recall limitations and social desirability 399 400 bias (Atkin et al., 2012). Furthermore, alcohol data could have been standardised around 401 units per day rather than drinks per day. Moreover, while fast-food consumption has been 402 associated with deleterious health outcomes and weight gain it is also plausible that other 403 foods not captures within this study may be adverse for weight gain and health. Valid and 404 reliable self-report measures were used where available (Helmerhorst et al., 2012, Chau et 405 al., 2012). Although not always possible, future research should use objective measures of 406 health behaviours. Furthermore, there are currently no accepted thresholds for excessive 407 sedentary behaviour or fast-food; therefore classifications are arbitrary and may vary 408 considerably between studies (Owen et al., 2011). Results may also not be generalisable 409 beyond the study sample and area as the study is restricted to one geographical area, the 410 sample was divided into clusters based on a data driven approach and adults aged over 50 were oversampled. Future research should consider stronger research designs to build on 411 412 existing, tentative support for the implementation of MHBC into public health policy. Further research is needed to assess interactions between multiple health behaviours and any 413 mediation or relationships between different behaviours and chronic disease. 414

415 **Conclusion**

This study has progressed our understanding of the clustering of the most prevalent health behaviours in adults and is amongst the first to identify clusters of health behaviours within population subgroups. Younger people who lived in the more deprived areas were largely within the '*elevated risk*' cluster representing an important target group for MHBC interventions given their wide range of unhealthy behaviours. The '*moderate risk*' cluster still exhibited a range of unhealthy behaviours but may benefit from a less intensive MHBC intervention that focuses on smaller changes in health behaviours. This study supports

- 423 previous calls for a more comprehensive approach to behaviour change. Future
- 424 interventions and policies should acknowledge a range of behaviours when designing MHBC
- 425 interventions, particularly for those who are younger and reside within deprived
- 426 neighbourhoods.
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Table 1: Demographic characteristics of participant's and the percentage (% (n)) that meet public health guidelines

Demographic characteristic	•	Non-	n- Alcohol	Physically	<7 hours	Fruit and	Fast-food
		smoker	Alconor	Active	sitting time	vegetable	1 231-1000
Overall	100.0 (934)	89.2 (833)	86.0 (803)	51.1 (477)	54.5 (509)	13.9 (130)	55.1 (515)
Gender							
Male	53.0 (495)	88.3 (437)	80.0 (396)	49.5 (245)	58.6 (290)	45.1 (223)	51.5 (255)
Female	47.0 (439)	90.2 (396)	92.7 (407)	60.1 (264)	73.3 (322)	63.1 (277)	59.2 (260)
Age							
18-34	9.2 (86)	14.0 (12)	86.0 (74)	57.0 (49)	58.1 (50)	44.2 (38)	37.2 (32)
35-44	18.4 (172)	14.5 (25)	91.9 (158)	50.6 (87)	64.5 (111)	49.4 (85)	41.3 (71)
45-54	22.7 (212)	14.6 (31)	85.8 (182)	45.3 (96)	62.3 (132)	50.9 (108)	46.7 (99)
55-64	24.2 (226)	8.8 (20)	83.2 (188)	52.2 (118)	61.9 (140)	54.4 (123)	61.9 (140)
65+	25.5 (238)	5.5 (13)	84.5 (201)	53.4 (127)	75.2 (179)	61.3 (146)	72.7 (173)
SEIFA IRD Quintiles							
1 < 964 (Lowest IRD)	25.5 (238)	83.6 (199)	85.7 (204)	48.3 (115)	68.5 (163)	53.8 (128)	55.9 (133)
2 965 – 1020	24.8 (232)	90.9 (211)	87.5 (203)	49.1 (114)	71.1 (165)	49.6 (115)	49.1 (114)

3 1021- 1058	26.4 (247)	89.1 (220)	84.6 (209)	56.7 (140)	62.3 (154)	53.4 (132)	60.3 (149)
4 > 1059 (Highest IRD)	23.2 (217)	93.5 (203)	86.2 (187)	49.8 (108)	59.9 (130)	57.6 (125)	54.8 (119)
Education							
Pre, Primary or High School	35.5 (332)	87.3 (290)	84.0 (279)	47.9 (159)	70.8 (235)	57.2 (190)	57.2 (190)
College/University	64.5 (602)	90.2 (543)	87.0 (524)	52.8 (318)	62.6 (377)	51.5 (310)	54.0 (325)
Weight Status (BMI)							
Weight Status (BMI) Underweight (<18.5)	1.7 (16)	75.0 (12)	100.0 (16)	43.8 (7)	87.5 (14)	62.5 (10)	68.8 (11)
	1.7 (16) 34.5 (332)	75.0 (12) 87.0 (280)	100.0 (16) 88.5 (285)	43.8 (7) 54.0 (174)	87.5 (14) 69.6 (224)	62.5 (10) 55.9 (180)	68.8 (11) 59.3 (191)
Underweight (<18.5)			. ,		. ,		
Underweight (<18.5) Healthy weight (18.5-24.99)	34.5 (332)	87.0 (280)	88.5 (285)	54.0 (174)	69.6 (224)	55.9 (180)	59.3 (191)

Note: SEFIA= Socio-Economic Indexes for Areas; IRD = Index of relative disadvantage; IRD lower score = greater disadvantage

Table 2: Health behaviours by cluster membership612

	Cluster 1	Cluster 2	Cluster 3	Overall Sample
	'Lower risk	'Moderate risk	'Elevated risk	(n=934)
	behaviour'	behaviour'	behaviour'	
	(46.7%, n=436)	(42.1%, n=393)	(11.2%, n=105)	
Smoker	100	100	3.8	89.2
(% do not smoke)				
Physical Activity	100	0	61	51.1
(% active)				
Alcohol	0.75	0.65	1.83	0.83
(drinks/day)				
Fruit and	5.34	4.56	4.43	4.91
Vegetable				
(servings/day)				
Sitting time	400	422	437	413
(minutes/day)				
Fast-food	1.56	1.65	1.68	1.61
(meals/week)				

•		••••	• • • • • •
	Cluster 1	Cluster 2	Cluster 3
	Lower risk	Moderate risk	Elevated risk
	behaviours	behaviours	behaviours
Total	46.7 (436)	42.1 (393)	11.2 (105)
Gender			
Male	55.3 (241)	49.4 (194)	57.1 (60)
Female	44.7 (195)	50.6 (199)	42.9 (45)
Age years * (Mean, (SD))	53.89 (±15.56)	54.81 (±14.38)	49.52 (±13.53)
18-34	10.5 (46)	6.1 (28)	11.4 (12)
35-44	17.7 (77)	17.6 (69)	24.8 (26)
45-54	19.5 (85)	24.4 (96)	29.5 (31)
55-64	25.0 (109)	24.2 (95)	21.0 (22)
65+	27.3 (119)	26.7 (105)	13.3 (14)
IRD* (Mean, (SD))	1013.33, (±62.80)	1014.48 (±60.60)	991.40 (±66.51)
Quartile 1	23.4 (102)	24.4 (96)	38.1 (40)
Quartile 2	23.9 (104)	27.0 (106)	21.0 (22)
Quartile 3	29.4 (128)	23.2 (91)	26.7 (28)
Quartile 4	23.4 (102)	25.4 (100)	14.3 (15)
Education			
Pre, Primary or High school	32.6 (142)	37.4 (9)	41.0 (43)
College/University	67.4 (294)	62.6 (246)	59.0 (62)

Table 3: Between group differences in cluster demographics (% of participants (n))

BMI (Mean, (SD))	26.73 (±5.25)	27.88 (±5.83)	26.71 (±5.68)
Weight Status			
Underweight	1.6 (7)	1.0 (5)	3.8 (4)
Healthy weight	35.6 (155)	31.6 (124)	41.0 (43)
Overweight	38.8 (169)	33.8 (133)	23.8 (25)
Obese	17.6 (77)	28.2 (110)	26.7 (28)

Note: * significantly differences by cluster membership p<0.05. Figures are reported as % (n) unless stated as mean, SD.

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