

Epidemiological correlates of overweight and obesity in the Northern Cape Province, South Africa

Mackenzie H Smith¹, Justin W Myrick², Oshiomah Oyageshio³, Caitlin Uren^{4,5}, Jamie Saayman⁴, Sihaam Boolay⁴, Lena van der Westhuizen², Marlo Möller^{4,5}, Brenna M Henn², Austin W Reynolds^{Corresp. 1}

¹ Department of Anthropology, Baylor University, Waco, United States

² Department of Anthropology and UC Davis Genome Center, University of California, Davis, Davis, United States

³ Center for Population Biology, University of California, Davis, Davis, United States

⁴ DSI-NRF Centre of Excellence for Biomedical Tuberculosis Research, South African Medical Research Council Centre for Tuberculosis Research, Division of Molecular Biology and Human Genetics, Faculty of Medicine and Health Sciences, University of Stellenbosch, Cape Town, South Africa

⁵ Centre for Bioinformatics and Computational Biology, University of Stellenbosch, Cape Town, South Africa

Corresponding Author: Austin W Reynolds
Email address: austin_reynolds@baylor.edu

Background. In the past several decades, obesity has become a major public health issue worldwide, associated with increased rates of chronic disease and death. Like many developing nations, South Africa is experiencing rapid increases in BMI, and as a result, evidence-based preventive strategies are needed to reduce the increasing burden of overweight and obesity. This study aimed to determine the prevalence and predictors of overweight and obesity among a multi-ethnic cohort from the rural Northern Cape of South Africa. **Methods.** These data were collected as part of a tuberculosis (TB) case-control study, with 395 healthy control participants included in the final analysis. Overweight and obesity were defined according to WHO classification. Multivariate linear models of BMI were generated using sex, age, education level, smoking, alcohol consumption, and diabetes as predictor variables. We also used multivariable logistic regression analysis to assess the relationship of these factors with overweight and obesity. **Results.** The average BMI in our study cohort was 25.1. The prevalence overweight was 18.6% and the prevalence of obesity was 23.5%. We find that female gender, being older, and having more years of formal education are all positively associated with BMI in our dataset. Women (OR = 5.4, CI = 3.2-9.4), older individuals (OR = 1.02, CI = 1-1.04), and those with more years of education (OR = 1.18, CI = 1.07-1.29) were all more likely to be overweight or obese. Alternatively, being a smoker is negatively associated with BMI and decreases one's odds of being overweight or obese (OR = 0.25, CI = 0.15-0.42). **Conclusions.** We observed a high prevalence of overweight and obesity in this study. The odds of being overweight and obesity was higher in women and those with more education and increases with age. Community-based interventions to control obesity in these

communities should pay special attention to these groups.

1 **Epidemiological Correlates of Overweight and Obesity in** 2 **the Northern Cape Province, South Africa**

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4 Mackenzie H Smith¹, Justin W Myrick², Oshiomah Oyageshio³, Caitlin Uren^{4,5}, Jamie Saayman⁴,5 Sihaam Boolay⁴, Lena van der Westhuizen², Cedric Werely⁴, Marlo Möller⁴, Brenna M Henn²,6 Austin W Reynolds¹

7

8 **Affiliations:** ¹ Department of Anthropology, Baylor University, Waco, Texas 76706, USA9 ² Department of Anthropology and UC Davis Genome Center, University of California, Davis,

10 CA 95616, USA

11 ³ Center for Population Biology, University of California, Davis, CA 95616, USA12 ⁴ DSI-NRF Centre of Excellence for Biomedical Tuberculosis Research, South African

13 Medical Research Council Centre for Tuberculosis Research, Division of Molecular Biology and

14 Human Genetics, Faculty of Medicine and Health Sciences, Stellenbosch University, Cape

15 Town, South Africa.

16 ⁵ Centre for Bioinformatics and Computational Biology, Stellenbosch University,

17 Stellenbosch, South Africa

18

19 Corresponding Author:

20 Austin W Reynolds¹

21 One Bear Place #97173

22 Waco, TX 76798

23 Email address: Austin_Reynolds@baylor.edu

24

25 **Abstract**

26 **Background.**

27 In the past several decades, obesity has become a major public health issue worldwide,

28 associated with increased rates of chronic disease and death. Like many developing nations,

29 South Africa is experiencing rapid increases in BMI, and as a result, evidence-based preventive
30 strategies are needed to reduce the increasing burden of overweight and obesity. This study
31 aimed to determine the prevalence and predictors of overweight and obesity among a multi-
32 ethnic cohort from the rural Northern Cape of South Africa.

33 **Methods.**

34 These data were collected as part of a tuberculosis (TB) case-control study, with 395 healthy
35 control participants included in the final analysis. Overweight and obesity were defined
36 according to WHO classification. Multivariate linear models of BMI were generated using sex,
37 age, education level, smoking, alcohol consumption, and diabetes as predictor variables. We also
38 used multivariable logistic regression analysis to assess the relationship of these factors with
39 overweight and obesity.

40 **Results.**

41 The average BMI in our study cohort was 25.1. The prevalence overweight was 18.6% and the
42 prevalence of obesity was 23.5%. We find that female gender, being older, and having more
43 years of formal education are all positively associated with BMI in our dataset. Women (OR =
44 5.4, CI = 3.2-9.4), older individuals (OR = 1.02, CI = 1-1.04), and those with more years of
45 education (OR = 1.18, CI = 1.07-1.29) were all more likely to be overweight or obese.

46 Alternatively, being a smoker is negatively associated with BMI and decreases one's odds of
47 being overweight or obese (OR = 0.25, CI = 0.15-0.42).

48 **Conclusions.**

49 We observed a high prevalence of overweight and obesity in this study. The odds of being
50 overweight and obesity was higher in women and those with more education and increases with

51 age. Community-based interventions to control obesity in these communities should pay special
52 attention to these groups.

53

54 **Introduction**

55 Obesity, a medical condition characterized by excessive fat accumulation, can have
56 severe consequences for health, including increased risk of cardiovascular disease, cancer,
57 stroke, and type 2 diabetes [1]. Obesity is most commonly determined through the measurement
58 of Body Mass Index (BMI), a metric calculated by dividing an individual's weight in kilograms
59 by their height in meters squared. From 1990 to 2015, global mortality associated with elevated
60 BMI increased by 28.3%, with the majority of these deaths being caused by cardiovascular
61 disease [2]. In addition to its effects on individual health, high obesity incidence can have
62 substantial economic implications, with a global impact of approximately \$2.0 trillion annually
63 [3]. The costs linked to obesity range from medical expenses and pharmaceuticals to absenteeism
64 and premature mortality [4].

65 Global obesity prevalence increased almost threefold between 1975 and 2016 [5]. A 2015
66 study showed that 603.7 million adults globally were classified as obese, representing 12.0% of
67 the adult population [2]. Based on current trends, it has been estimated that 1.12 billion adults
68 worldwide will be classified as obese by 2030, and an additional 2.16 billion as overweight [6],
69 roughly 40% of the world population. Sub-Saharan Africa in particular has shown staggering
70 increases in obesity prevalence in the past several decades. Among seven countries in this region
71 in 2009, an average of 31.4% of women were classified as overweight or obese [7]. In Malawi,
72 18% of men and 44% of women in urban areas were categorized as overweight or obese [8].

73 In accordance with both global trends and those observed in sub-Saharan Africa, the
74 population of South Africa has also demonstrated a rapid increase in average BMI over the past

75 several years [9–11]. As of 2002, 29.2% of men and 56.6% of women in South Africa were
76 categorized as overweight or obese [12]. These values are higher than what has been reported in
77 other African nations [13]. Various behavioral factors have also been studied in South African
78 populations, with several studies indicating an inverse correlation between smoking and BMI
79 [14, 15]. There is also evidence for higher obesity incidence among women as compared to men
80 in South Africa, a trend often observed in developing nations [6, 13, 16].

81 It is important to note, however, that the majority of research on BMI and obesity in
82 South Africa has been conducted on urban cohorts, particularly in the Eastern and Western Cape
83 Provinces [17, 18]. Although national statistics on measures of obesity are available [19],
84 research on rural and peri-urban South African areas, such as those in the Northern Cape
85 Province, remains limited, restricting our understanding of how overweight and obesity manifest
86 in these communities. This gap in the literature is important given the observed differences in
87 obesity trends between urban and rural communities in lower- and middle-income nations [20].
88 In South Africa specifically, women in urban areas were 1.6 times more likely, and urban men
89 2.3 times more likely, to have excessive BMI than those in rural areas [16]. However, the
90 average BMI of both men and women is increasing much more rapidly among rural communities
91 than urban communities globally [21]. Further investigation in these smaller communities is
92 crucial in identifying risk factors and working towards improved public health initiatives and
93 education.

94 In this study, we analyze some of the demographic and behavioral factors related to BMI
95 in the Northern Cape of South Africa. Our sample consists of 395 individuals from rural areas,
96 small towns, and one large municipality in the province. Investigating these factors will lay the

97 groundwork for the development of specialized public health and education initiatives to reduce
98 the prevalence of obesity and overweight among this region.

99

100 **Materials & Methods**

101 Study Design and Sampling Procedure

102 The data used in this study was collected as part of the Northern Cape Tuberculosis
103 (NCTB) Project, a case-control study on host susceptibility to TB. Between 2017 and the
104 beginning of the SARS-CoV-2 pandemic in 2020, data was collected on 1,095 individuals (N_{men}
105 = 544; $N_{\text{women}} = 551$) recruited from 12 community (public) health clinics in the Northern Cape
106 Province, South Africa that serve populations with high TB rates, mostly in rural towns with
107 populations under 10,000. The Northern Cape Province has the largest area, lowest population
108 size and lowest population density in South Africa. Data was collected via participant interviews,
109 medical histories, saliva samples, and anthropometric measurements. In this study, only
110 participants with a negative TB result were included due to a symptomatic effect of TB on BMI
111 [22]. Participants were partitioned into the case or control group based on a decision tree
112 considering previous TB diagnosis, treatment for TB, and TB test results obtained during the
113 course of the study (Oyageshio, in prep.).

114

115 Ethics and Informed Consent

116 Data collected from study participants was approved by the Health Research Ethics
117 Committee of Stellenbosch University (Project number: N11/07/210A) and the Institutional
118 Review Board of the University of California, Davis (IRB number: 1749073-1). Participation in
119 this study was voluntary, with the ability to withdraw at any time. Written informed consent was
120 obtained and subsequent medical and demographic questionnaires were conducted in the local

121 language of Afrikaans by trained research assistants from the community. All data was kept
122 confidential with no connections to participant names. Deidentified variables were stored in a
123 secure RedCap database following data collection.

124

125 Demographic and Socio-economic Factors

126 Demographic information was collected through interviews and recorded on a data
127 collection sheet. Participants were asked to provide their town of residence, highest level of
128 education achieved, and the ethnic group with which they self-identify.

129

130 Medical History

131 During interviews, participants self-reported diabetes, HIV, asthma, and TB status. If
132 participants reported having diabetes, they were asked to identify if they were a type 1 or type 2
133 diabetic. HIV status was self-reported as positive, negative, or unknown. Asthma was recorded as
134 no, yes, or unknown. In addition to asking participants whether they had TB at the time of the
135 study, they were also asked if they were currently taking TB medication, whether they had TB in
136 the past, and if so, how many TB episodes they had experienced. Following the self-reporting of
137 TB information, TB tests were administered to all participants using GeneXpert, Auramine O
138 Stain, GeneXpert Ultra, SMEAR, or Culture tests.

139

140 Behavioral Factors

141 Alcohol use was evaluated by asking participants if they drink alcohol. If yes, further
142 information was collected regarding the specifics of their alcohol consumption. Participants were
143 asked whether they drink beer, wine, liquor, or ginger beer, and how much of each they consume

144 during the week and weekend. Smoking behavior was categorized as yes or no. If participants
145 answered yes to smoking, they were asked to provide the age at which they began smoking, as
146 well as the average amount they smoke per day.

147

148 Anthropometry

149 Height was measured using a stadiometer. Participants were asked to look straight ahead
150 and stand with their heels against the wall while measurements were being taken. One height
151 measurement was recorded for participants in pilot data collection and two measurements were
152 taken upon initiation of the primary study. For individuals with two recorded measurements for
153 height, the average of these values was used. Weight was measured with a digital scale and
154 recorded once. Height and weight measurements were used to calculate BMI for each participant
155 using the equation: $BMI = \text{kg}/\text{m}^2$. Waist and hip circumference measurements were self-
156 administered by participants using a measuring tape, and two measurements of each were
157 recorded. Following calculation of BMI, participants were assigned to weight categories based
158 on the guidelines set forth by the Centers for Disease Control and Prevention: those with a BMI
159 < 18.5 were classified as “Underweight”, 18.5 to <25 as “Healthy,” 25 to <30 as “Overweight,”
160 and 30 or higher as “Obese” [23]. Waist-to-hip ratio was calculated by dividing average waist
161 circumference (cm) by average hip circumference (cm).

162

163 Quality Control

164 Participants who were TB positive, HIV positive, or both ($N = 642$), were removed from
165 this study due to an expected effect of disease on BMI. All hip and waist measurements
166 supervised by community health care worker #5 were also removed due to errors in measurement

167 technique. BMI could not be calculated for an additional 18 participants due to an absence of
168 height measurements, weight measurements, or both, and these individuals were removed. A
169 total of 395 individuals were included in BMI analyses.

170

171 Statistical Analyses

172 Quality control and data analysis were performed in R 4.0.2. The packages effects [24,
173 25] was used for data analysis, and ggplot2 [26] was used for data visualization.

174 Six covariates (alcohol intake (yes or no), smoking (yes or no), diabetes (yes or no), sex,
175 age, and years of education) were entered into a generalized linear model to further characterize
176 their association with BMI. As observed in other studies, BMI was not normally distributed
177 among our sample, therefore we used log-transformed BMI as the outcome variable. Bivariate
178 effect plots were generated between BMI and each individual covariate included in the model.
179 Additional generalized linear models were generated by partitioning male and female
180 participants to examine sex-specific effects.

181 We also performed statistical analyses using waist-to-hip ratio (WHR) in place of BMI. A
182 generalized linear model was created using WHR as the outcome variable and included alcohol
183 intake, smoking, diabetes, sex, years of education, and age. This model included 161
184 participants. Bivariate effect plots were created between WHR and alcohol intake, smoking,
185 diabetes, sex, age, and highest qualification. Our generalized linear models were supplemented
186 by the generation of odds ratios using the questionr package [27] to investigate the relationship
187 between individual variables and BMI outcome (overweight/obese versus not).

188

189 Results

190 Participant Demographics

191 Our final sample (**Table 1**) included 395 participants ($n_{\text{male}} = 145$; $n_{\text{female}} = 250$). The
192 median age was 43 years, with all participants falling between 18 and 86 years of age. Our
193 dataset was collected across twelve study sites, representing rural areas, small towns, and one
194 small city from the Northern Cape Province, South Africa. The most common self-identified
195 ethnicity among participants was Coloured (85%), followed by Tswana (5%). Approximately 7%
196 of participants reported having no education, 27% reported attending or completing primary
197 school, and 66% reported attending or completing secondary school.

198 Behavioral Factors

199 The majority (62%) of participants were smokers. Among women, 54% reported that
200 they smoked, compared to 75% of men. Only 39% of participants reported consuming some
201 quantity of alcohol. This also differed by sex, with 50% of men and 34% of women in our
202 sample reporting alcohol consumption.

203

204 Anthropometry and BMI

205 Median height and weight were 159cm (range: 109.4-184.9cm) and 60kg (range: 27-
206 134kg) respectively. The median BMI was 23.6, and the mean BMI was 25.3. Nearly a quarter
207 (22%) of participants were classified as underweight, 35% as healthy, 18% as overweight, and
208 25% as obese. The distribution of individuals between these classifications differed substantially
209 between male and female participants, with 36% of female participants being placed in the obese
210 category compared to only 5% of male participants (**Figure 1; Table 1**).

211

212 Factors associated with BMI, WHR, and obesity

213 We fit a generalized linear model to investigate the relationship between BMI and the
214 factors of sex, age, education, smoking, drinking, and diabetes status. This model showed
215 significant relationships between BMI and smoking, gender, education level, age, and diabetes
216 (**Table 2**). Together, these factors explained 26% of the variance in BMI among our cohort.
217 Being female showed a strong positive correlation with BMI, as did age and years of education.
218 The relationships between each of these variables and BMI were observed in our bivariate effect
219 plots (**Figure 2**), which indicated a 5-point BMI increase for women, a 4-point decrease for
220 smoking, and a 6-point increase for having 14 years of education as compared to zero.

221 We find evidence that women in our sample had over 5 times greater odds of being
222 overweight or obese compared to men (OR = 5.4, CI = 3.2-9.4; **Table 3**). In our dataset, every
223 one-year increase in age is associated with a 2% increase in the odds of being overweight or
224 obese, while each additional year of school completed increased those odds by 18%. Smoking,
225 on the other hand, showed a significant negative correlation with BMI ($p = 5.85 \times 10^{-8}$). In fact,
226 smoking decreased the odds of being overweight or obese by nearly 75% in our study population
227 (OR = 0.25, CI = 0.15-0.42).

228 Because of the strong sex effect on BMI, we also fit models for women (Table S1; $n =$
229 250) and men (Table S2; $n = 145$) separately to investigate potential sex-specific factors
230 associated with BMI. The strong negative correlation between BMI and smoking seen in our
231 primary model was maintained in both sex-specific models ($p = 4.4 \times 10^{-5}$ and 3.2×10^{-3} ,
232 respectively). The significant positive correlations between years of education, age, diabetes and
233 BMI are maintained with the women-only model, however these relationships were notably
234 weaker than in the primary model. None of these significant correlations are seen in the men-
235 only model.

236 We also explored the factors associated with waist-to-hip measurement ratios in our
237 dataset. This model was fit on a subset of the dataset (Table S3; n=160), as waist and hip
238 measurements were not available for all participants. We find a slight positive correlation
239 between WHR and years of education ($p = 0.0412$), but together these variables only explain
240 ~2% of the variance in WHR in our dataset. No other covariates showed a significant association
241 with WHR.

242

243 Discussion

244 In this study we analyzed the demographic and socio-behavioral factors related to BMI in an
245 adult cohort from the Northern Cape Province, South Africa. Our results suggest a significant
246 difference in obesity trends between men and women in the study population. Sex-based
247 differences were observed not only in obesity incidence, but also in the degree to which various
248 factors impact BMI outcomes. While smoking has a strong negative correlation with BMI in
249 both men and women, we find that age and years of education are only significantly correlated
250 with BMI in women. Although we had significantly more female participants (250 women, 145
251 men), our results are consistent with previous research. Higher obesity incidence among women
252 as compared to men is observed in other studies on South African populations [9, 14–16]. While
253 data regarding the effect of socioeconomic status on BMI outcomes in males and females shows
254 variability depending on the metric used to represent socioeconomic status, Wagner et al. found
255 that primary and tertiary education were associated with greater BMI values in females only
256 [14]. Interestingly, while we find an association between age and BMI among females only,
257 previous research has demonstrated this correlation for males only [15, 16].

258 One limitation of the current study is the relatively limited sample size of waist and hip
259 measurements. While BMI is a widely used metric for obesity that is correlated with body fat
260 content, past research has shown that researchers must be careful not to over-interpret BMI in
261 studies of obesity, as it does not distinguish between excess fat, muscle, or bone mass, or provide
262 any indication of the distribution of fat among individuals [28, 29]. Several studies have
263 suggested that WHR are better at predicting visceral adiposity and cardiometabolic disease risk
264 [30, 31]. Future work incorporating these and other more sensitive measures of adiposity will
265 provide greater detail into the risk-factors and consequences of obesity in this population.

266 In line with prior research on sex differences in BMI and obesity, we find 57% of female
267 participants were overweight or obese compared to only 19% of male participants. One such
268 study found that in 5 out of 6 sampling locations in sub-Saharan Africa, a higher percentage of
269 females were categorized as obese as compared to males [32]. A study in Soweto, South Africa
270 showed that the proportion of women classified as obese was twice that of men [15]. There are
271 several potential explanations for the extreme obesity risk associated with being female in our
272 sample population. There is some evidence to suggest higher BMI is desirable in some South
273 African communities [33, 34]. It has also been shown that having control over household food
274 spending is associated with greater incidence of obesity in women [33].

275 Previous work has shown that socioeconomic status, including educational level, is
276 positively correlated with BMI, for both men and women, in South Africa [32, 35, 36], as well as
277 other low- and middle-income nations [32, 37]. This aligns with our result demonstrating the
278 more education one has, the higher their BMI – each year of education increases their odds of
279 being overweight or obese by 18%. This trend may be explained by cultural factors, as being
280 overweight or obese reflects wealth in low-income communities where many lack adequate

281 access to food. A study on 37 lower- and middle-income nations showed that BMI was positively
282 associated with wealth in all 37 countries surveyed [38]. Previous work has also found that a
283 person's daily amount of vigorous physical activity also decreases with increasing wealth [39,
284 40], suggesting a multifactorial relationship between socioeconomic status and BMI.
285 Interestingly, we do not see a significant relationship between age and years of education in our
286 model of men-only. This pattern has been reported elsewhere [33] and suggest that other factors
287 may be more important in explaining BMI variation in men in the Northern Cape.

288 Our results indicating that smoking decreases odds of being overweight or obese by 75%
289 are also in line with previous research. Nicotine use suppresses appetite and causes an increased
290 resting metabolic rate, resulting in weight loss [41]. Smoking has previously been associated
291 with lower BMI in South African adults [15, 32].

292

293 **Conclusions**

294 Overall, our results suggest that smoking, sex, education level, age, and diabetes all
295 influence BMI outcomes among the rural population of the Northern Cape of South Africa,
296 consistent with previous research in other South African cohorts and developing nations. These
297 results deepen our understanding of the factors contributing to obesity risk in this region. A
298 necessary future direction of this work will be to include additional measures of adiposity,
299 behavior, and socio-economic status to obtain a more complete understanding of the risk-factors
300 for obesity in the Northern Cape. This will allow for better identification of high-risk groups for
301 behavioral interventions, ultimately mitigating the public health and economic burdens created
302 by the global obesity epidemic at the local level.

303

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310

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312

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Figure 1

Histogram of BMI values in our dataset, partitioned by sex

Pink bars represent the distribution in males, while blue bars represent the distribution in females. We find that the average BMI for females is higher than that of males.

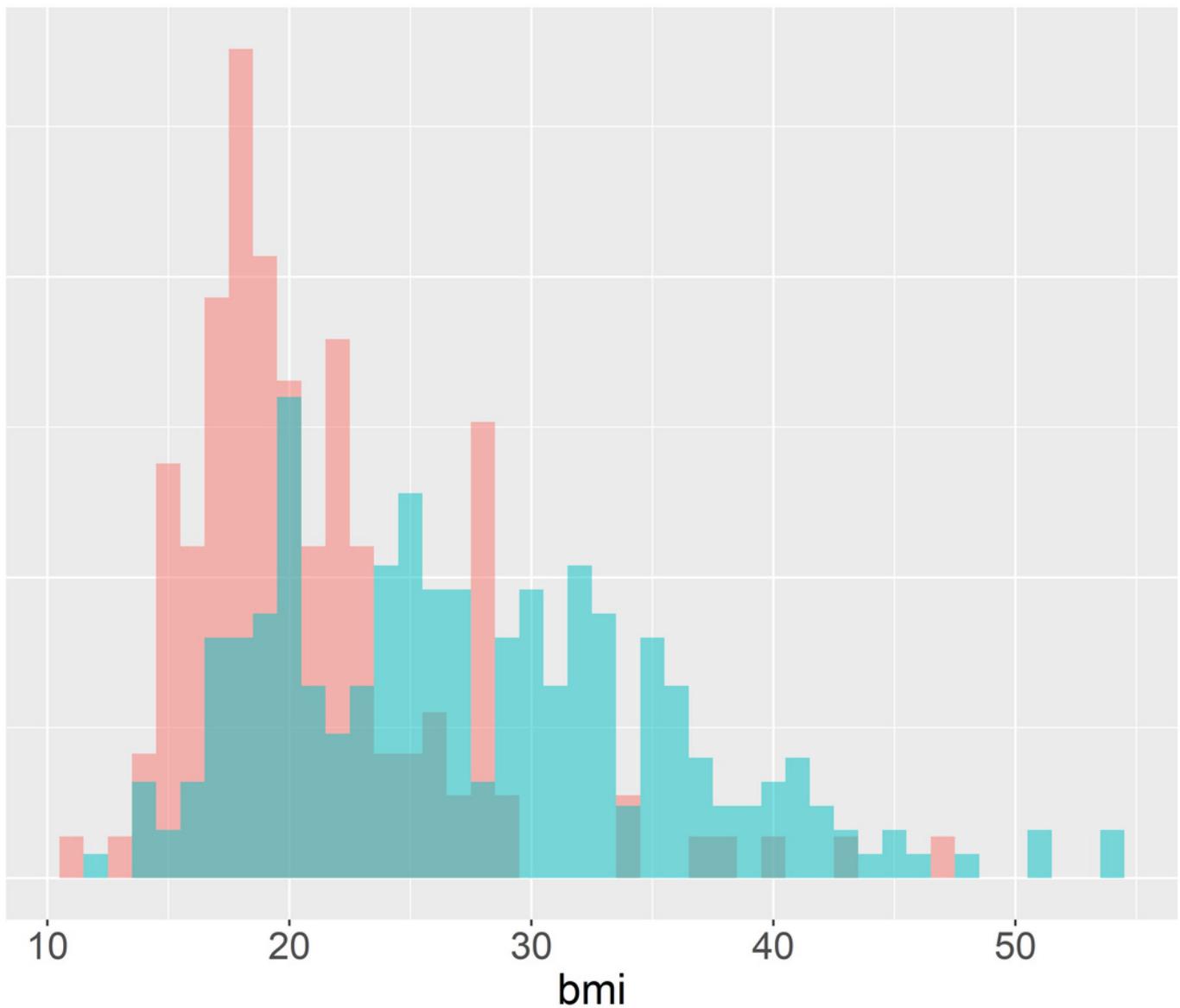


Figure 2

Bivariate effect plots demonstrating the relationship between BMI and the covariates included in Model 1

We find a 5-point BMI increase associated with being female, a 6-point increase for having 14 years of education, and a 4-point decrease for smoking.

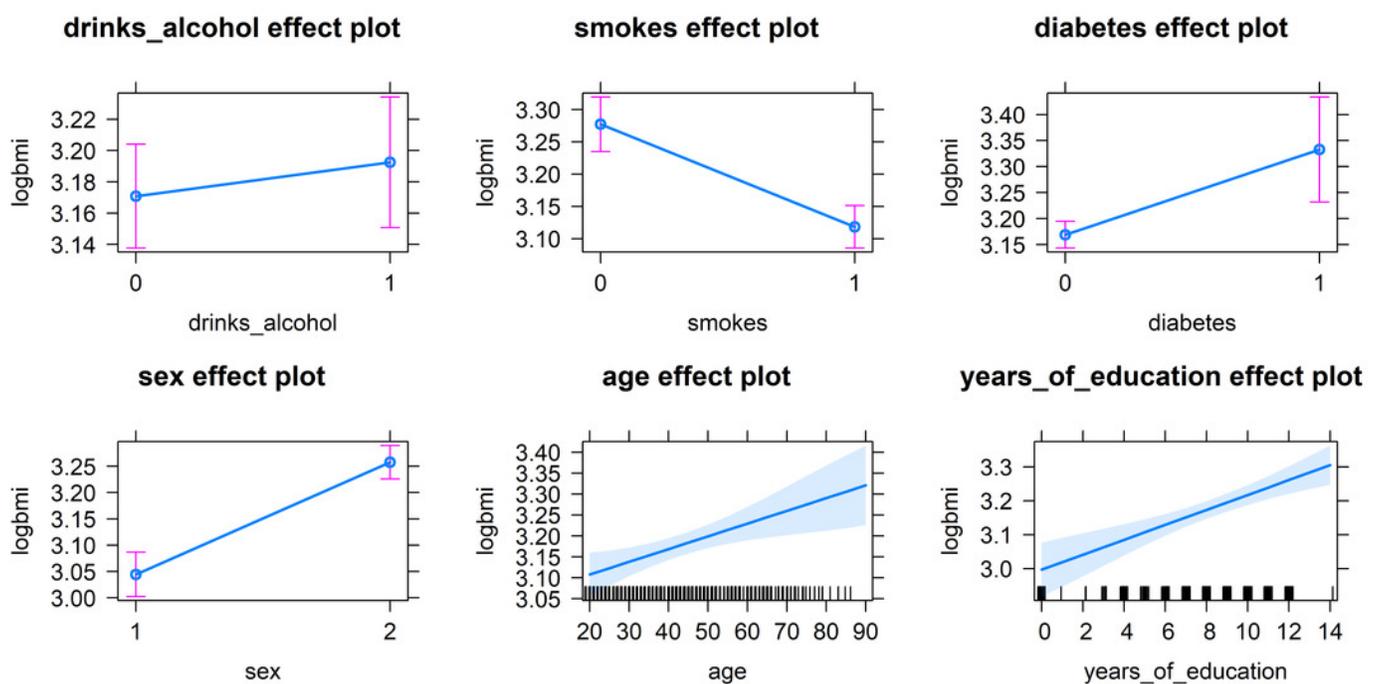


Table 1 (on next page)

Summary of participant demographics, behavioral factors, anthropometry, and BMI.

Summary statistics are provided for men, women, and the total sample. We find substantial differences in average BMI and overweight/obesity incidence between men and women.

**WHR only available for 160 participants.*

Variable	Men	Women	Total
Participants (N)	145	250	395
Age (mean)	44	43	43
Years of education (mean)	8	8	8
Smokes	75%	54%	62%
Drinks Alcohol	50%	34%	39%
Height (cm) (mean)	165.2	155.7	159.2
Weight (kg) (mean)	57	67.3	63.5
BMI (mean)	21	27.8	25.3
WHR (mean)*	0.84	0.88	0.87
Underweight	39% (n=57)	12% (n=31)	22% (n=88)
Healthy weight	43% (n=62)	31% (n=77)	35% (n=139)
Overweight	13% (n=19)	21% (n=52)	18% (n=71)
Obese	5% (n=7)	36% (n=90)	25% (n=97)

1 **Table 1:** Summary of participant demographics, behavioral factors, anthropometry, and BMI. Summary
2 statistics are provided for men, women, and the total sample. We find substantial differences in average
3 BMI and overweight/obesity incidence between men and women. *WHR only available for 160
4 participants.

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Table 2 (on next page)

Model 1 results showing the impact of six covariates on BMI outcomes in our sample of 395 participants.

We find significant positive correlations between BMI and age, years of education, and being female. We find a significant negative correlation between smoking and BMI.

Variable	Estimate	se	z	Pr(> z)
(Intercept)	2.808286	0.081325	34.532	2.00E-16
Alcohol	0.02173	0.028067	0.774	0.43929
SexF	0.213366	0.027302	7.815	5.22E-14
Age	0.003044	0.001008	3.021	0.00268
Years of education	0.022012	0.004646	4.738	3.03E-06
Smokes	-0.158903	0.028365	-5.602	4.02E-08
Diabetes	0.163951	0.053133	3.086	0.00218

1 **Table 2:** Model 1 results showing the impact of six covariates on BMI outcomes in our sample of 395
2 participants. We find significant positive correlations between BMI and age, years of education, and
3 being female. We find a significant negative correlation between smoking and BMI.

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Table 3 (on next page)

Odds ratio results for the covariates included in Model 1.

We find that women have 5x greater risk of being overweight or obese, each additional year of school increases risk by 18%, and smoking decreases risk by 75%.

Variable	OR	Lower CI (2.5%)	Upper CI (97.5%)	p
(Intercept)	0.041617	0.0081968	0.1969	8.49E-05
Alcohol	1.2391987	0.7352237	2.1097	0.4241174
Sex	5.4056422	3.1941477	9.4323	8.94E-10
Age	1.0234617	1.0047186	1.043	0.0147148
Years of education	1.1763391	1.074636	1.294	0.0005895
Smokes	0.2535469	0.1520819	0.4172	9.23E-08
Diabetes	4.221319	1.4584546	14.5498	0.0126958

1 **Table 3:** Odds ratio results for the covariates included in Model 1. We find that women have 5x greater
2 risk of being overweight or obese, each additional year of school increases risk by 18%, and smoking
3 decreases risk by 75%.
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