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Efficacy of “Mentoring to Be Active” on Weight Loss, Body Mass Index, and Body Fat among Obese and Extremely Obese Youth in Rural Appalachia

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Abstract

Purpose: This study compares the efficacy of a behavioral skills mentoring program (*Mentored Planning to Be Active [MBA]*) to a teacher-led program (*Planning to Be Active [PBA]*) for increasing physical activity in Appalachian teens on health outcomes (weight loss, body mass index (BMI), and body fat).

Methods: Secondary analysis of a larger group-randomized controlled trial was conducted in 20 rural Appalachian schools. Descriptive Pearson correlations and multivariate analyses with between-subject effects were conducted. Effect sizes (ES) using Cohen’s *d* and odds ratios (OR) with 95% confidence intervals were calculated.

Findings: The obese MBA group lost 77.5% more weight by T3 compared to the PBA group; T2 was ($F = 8.51, P = .000$) and T3 was ($F = 7.62, P = .000$). ES was 0.34. OR = 1.45 (95% CI: 0.558–3.792) at T2 and OR = 3.32 (95% CI: 1.103–9.978) at T3. Extremely obese in the MBA group lost 80.0% more weight compared to the PBA group; T2 was ($F = 5.23, P = .025$) and at T3 ($F = 6.33, P = .015$) ES was 0.58. OR = 4.36 (95% CI: 0.981–19.34). Extremely obese females lost more weight compared to males ($F = 4.75, P = .034$). BMI and body fat had similar results; youth in the MBA group had the most improvement.

Conclusions: Rural Appalachian youth are disproportionately extremely obese. BMI does not capture adiposity or cardiovascular risk. BMI, BMI percentile, raw weight, fat mass, and percent body fat are more complete analyses of adiposity and cardiovascular risk.

Keywords

body mass index; extremely obese; health promotion; obesity; weight loss

The majority (72%) of the United States’ land area is rural.¹ Residing in a rural area, coupled with poverty, is associated with disparate suffering from high rates of chronic disease and disability related to the early onset of obesity.² Compared to other sub-

populations, rural Appalachian youth are disproportionately obese or extremely obese. In a sample of 333 ninth-grade rural Appalachian students, Smith and associates found obesity rates were 16.5% and 13.2% of youth were extremely obese.³ According to National Health and Nutrition Examination Survey [NHANES] data, national rates of extreme obesity range from 5.9% to 7.8%, whereas the obesity rate of high school youth is approximately 20%.⁴ Children and youth residing in rural areas are 25% more likely to be overweight or obese, inclusive of extremely obese, compared to those residing in metropolitan areas.^{5,6} Obesity-related medical costs are estimated to cost the United States up to \$210 billion annually.^{6,7}

The high prevalence of extreme obesity combined with high rates of sedentary behaviors and poor nutritional quality place Appalachian teens at increased risk for diabetes and cardiovascular disease earlier in life, resulting in a poorer quality of life and earlier mortality.^{2,3} A body mass index (BMI) over 40 kg/m² reduces life expectancy by 6.5–13.7 years in adults.⁸ Heart disease, diabetes mellitus type II, and certain cancers increase as BMI increases.^{8,9} Extremely obese youth have earlier onset of illness and earlier death due to early acquisition of comorbidities such as nonalcoholic fatty liver disease, musculoskeletal problems, and long-term risk of disease.^{4–6,10} Cardiovascular and metabolic profiles of youth suffering from extreme obesity are poorer compared to even less obese counterparts.^{11–14} Extremely obese youth have early onset of hypertension, dyslipidemia, inflammation, and hyperinsulinemia.^{12–14} Ramifications of teen extreme obesity are a serious concern in rural Appalachia and pose important clinical and public health challenges.

Appalachian resident's personal health behaviors are related to the region's poor health outcomes.³ Appalachians have poorer health and fewer positive health-related behaviors compared to other US populations.^{4–6} Compared to other Americans, Appalachians are less physically active in their leisure time.^{15,16} Appalachian youth also consume up to 4 times the amount of sugared-sweetened beverages (SSBs) compared to non-Appalachian youth.^{16–18} Cultural norms found within Appalachia fuel this disproportionate consumption of SSBs. These norms include “sipping soda or pop throughout the day,” a preference for acidic beverages, and skepticism of outsiders delivering health-related programming.^{5,6,17} Youth residing in high-poverty areas such as rural Appalachia are affected more frequently by poor dietary behaviors such as SSBs.^{6,18}

Effective interventions to improve overall health are needed in rural Appalachia because overcoming health disparities for low-income Appalachian teens has not been adequately addressed.^{6,19} In rural, under-resourced areas such as Appalachia, physical inactivity is significantly higher than national levels.^{2,15,16} Appalachian youth are among the most sedentary.^{3,16,20} One explanation may be circumstances in Appalachia. For example, relying on organized sports to promote physical activity (PA) behaviors has not been effective for Appalachian youth.¹ Opportunities to participate in sports are limited due to inadequate school resources, lack of transportation, and limited availability of school teams.^{3,6,15,16} School-sponsored programs enroll a small percentage of students. For Appalachian youth, the motivation to be physically active must shift from competitive sports to personal fitness grounded in personal motivation. Using trained peer mentors can reach more youth and is more likely to carry over into adulthood. Our experience of working in rural Appalachia suggests that many students do not have access to exercise and fitness facilities characteristic

of urban and suburban settings. Lack of transportation and distance to facilities are barriers for many Appalachian youth. In Appalachia, interventions should emphasize activity done in rural neighborhoods and at home. These skills are useful for sustaining goal-directed behavior change.

School-based interventions have also been limited in their scope and effect on obesity prevention especially targeting teens.^{6,20} Low efficacy of these programs may be due to unique cultural challenges such as a preference for informal sharing of information among local residents rather than health content delivered by professionals.⁶ Though school-based interventions increase health knowledge, there is less evidence of effectiveness for health behavior changes and improving health outcomes.^{19–24} School-based interventions typically deliver programs as part of a course such as health or physical education via teachers or adults. Low efficacy of these programs may also be due to cultural challenges in Appalachia.^{6,25} Other factors include varying fidelity of program delivery by classroom teachers, school environments not consistent with program goals, and a lack of theoretical programing targeting health behaviors.^{20,21,26} In rural Appalachia, resources for preventive care are lacking. Resources to engage in healthful behaviors such as recreational facilities, organized sports, or neighborhood parks or facilities for PA are inadequate in rural Appalachia.⁵ To better understand the unique challenges to healthy lifestyle behaviors in rural Appalachia, community-based studies are needed.²⁰

Behavioral interventions are usually the first line of treatment for obese adolescents.²⁷ Few intervention studies with behavioral and lifestyle interventions have focused on extremely obese youth.^{27–29} For youth suffering from obesity and extreme obesity, interventions have mostly resulted in modest reductions in weight and BMI because baseline BMI values are so high.²⁹

A promising new approach for the delivery of school-based interventions may be to use trained peer mentors. Rarely have school-based interventions used peer mentors to deliver content and offer support for building skills to change behavior. Mentoring has been effective at addressing health risk behaviors among Appalachian youth,³⁰ including overweight and obesity.^{23,24} Smith and Holloman found that Appalachian peer mentors helped elementary-aged mentees improve short-term dietary behaviors. Appalachian youth assigned to peer mentors (versus adult leaders) demonstrated improved BMI and increased PA on a short-term basis.²⁴

Mentored Planning to Be Active [MBA] was adapted from *Planning to Be Active [PBA]* for delivery by trained peer mentors. MBA emphasizes the social determinants of health by using a social networking approach that trains peer mentors to support teens. PBA is a 10-lesson unit delivered over 10 weeks and designed to teach self-regulation of PA in the home and neighborhood among teens. PBA's overall objective is to help youth develop and implement a personalized PA program at home with no exercise equipment. The personalization occurs with weekly goal setting and tracking of behaviors. A subsequent review of weekly goals and activities are reviewed the following week with feedback and revision, if needed. Expanding PBA for delivery by trained peer mentors via MBA may better promote and sustain adoption of daily regular PA through self-regulation of PA in

discretionary time. By addressing personal concerns as well as using existing social networks or not using health professionals to deliver the message, MBA may strengthen the tailoring of PA to personal interests, talents, and neighborhood.

Social Cognitive Theory (SCT) is extensively used for determinants of PA^{31–33} and is the framework for the PBA and MBA curriculum. PBA was developed and tested over the course of prior studies.^{31–33} Studies revealed that SCT variables were strongly related to moderate and vigorous PA; PBA increased self-regulation of PA; and a 10-lesson dose led to the greatest improvement in PA. The PBA curriculum addresses psychosocial determinants, self-regulation, and environment affecting PA adherence (Figure 1). Psychosocial determinants include outcome expectancies and exercise self-efficacy. Self-regulation includes goal setting, self-monitoring, overcoming barriers, time management, self-reward, and social support. Environmental determinants are home, neighborhood, and school opportunities for PA.

The peer mentoring approach used for MBA builds and strengthens social networks that links people and provides social support.³⁴ One's social network provides emotional, informational, and appraisal support to create a sense of psychological safety and higher motivation to change behavior. Learning, domain-specific self-efficacy, and behavior change happen when people have a sense of psychological safety or the perception that attempts to change behavior can occur without fear or embarrassment.³⁵ Advantages include enhanced learning and behavioral change support resulting from the perceived social support (emotional, informational, and appraisal) and psychological safety promoted by peer mentoring. Providing intense social support via peer mentoring results in curriculum delivery that promotes better health outcomes compared to classroom teachers delivering the curriculum.

The purpose of this study was to compare the efficacy of a healthy lifestyle skills mentoring program (MBA) to a teacher-led program (PBA) for increasing PA in Appalachian teens on health outcomes (eg, weight loss, BMI, and percentage of body fat). We hypothesized that obese and extremely obese teens in the MBA group would have greater reductions in body weight, BMI, and body fat compared to teens in the PBA group at 3 months (T2) and 6 months (T3) postintervention.

For the main study, we tested the efficacy of a 10-week school-based intervention in promoting self-regulation of PA among youth. A PA program originally designed for classroom delivery (usual care) was adapted to also be delivered via trained peer mentors over 10, 40-minute sessions (intervention).²² Curricular workbooks, manipulation checks (homework), worksheets, reinforcement activities, and weekly goal setting were provided to all participating youth. Each week, participants considered ways to incorporate PA into discretionary time.²² Detailed intervention procedures and data protocols are described elsewhere.²²

Methods

This secondary analysis was from a larger group-randomized controlled trial (G-RCT) conducted in 20 rural Appalachian high schools in southern Ohio.²⁵ Rural areas are nonmetropolitan or nonurban areas based on population density, housing, and territory.^{1,2,9,36,37} Rural counties generally have smaller towns, are outside the boundaries of metropolitan areas, and have lower population density. Descriptive and inferential analyses were completed. Multivariate analyses with tests of between-subject effects using mixed model ANCOVA were conducted. Effect sizes (ES) were calculated using Cohen's *d*. When standard deviations were not similar, Glass' delta technique was performed; if sample sizes were not similar, then Hedges' *g* was used. Small ES was defined as <0.2; medium ES = 0.3–0.5; large ES = 0.6 or above. The level of significance was set at *P* .05. To further quantify the strength of the association between MBA and PBA on health outcomes, odds ratios with 95% confidence intervals were calculated.

Sample and Procedure

A systematic sample was drawn from schools meeting the following criteria: (1) located in a rural Appalachian county; (2) county-level obesity rates where schools were located were similar to other counties of included schools; and (3) at least half (50%) or more of students were on a free or reduced lunch program.²² From these criteria, a total sample of “obese” adolescents (*n* = 106) and “extremely obese” teens (*n* = 84) in grades 9–11 at baseline data collection were recruited. The sample for this analysis was drawn from a larger sample of participants (*n* = 519). Although the larger study included 329 healthy weight youth, this paper presents results from a subgroup analysis of youth classified as obese or extremely obese. Nearly 37% (36.6%) of the overall sample were obese or extremely obese at baseline. This rate far exceeds the current adolescent obesity rates for 12- to 19-year-olds estimated at 20.6% in the United States.^{13,38} Baseline (T1), 3-month follow-up (T2), and 6-month follow-up (T3) data were used. Prior to data collection, teen written assent and parental permissions were obtained. The Human Subjects Committee (Protocol # 2014B0094) at the host institution and each school approved the study. The study is registered on [ClinicalTrials.gov](https://clinicaltrials.gov/ct2/show/study/NCT02329262) (ID number: NCT02329262).

Measures

Raw Weight and BMI for Age and Gender (BMI)—Health researchers and health care providers measure BMI as an indicator of obesity.^{4,6,10} However, BMI does not fully capture adiposity or cardiovascular risk. BMI is highly correlated with many different components of body weight including lean mass, skeletal muscle mass, fat mass, and bone mass, yet BMI cannot distinguish between these different components.¹⁰ Studies have found a strong link between BMI and percentage of body fat.^{9,10,39} In research, BMI remains the most commonly used measure to assess general body composition.¹⁰ Using the Tanita DC-430U Body Composition Analyzer,⁴⁰ Raw Body Weight (RW) and BMI for age and gender were assessed.

The Tanita portable professional grade BIA analyzer has been valid and reliable in estimating body weight, BMI, and the percentage of body fat in adolescents when compared

to dual-energy X-ray (DXA).^{12,41} Procedures in the *Tanita User's Guide* were followed.⁴⁰ Height was measured by having the teen stand without shoes on a portable stadiometer facing forward. Age (in years), gender, and height were entered into the analyzer for calculations. Teens stood on the analyzer without shoes or socks, having their feet on the measuring pads and hands directed down the side of their legs.

We used actual BMI as the outcome of interest because this was a long-term trial. BMI *z*-scores have been widely used and are mostly recommended for cross-sectional comparisons. Recently, several authors have cautioned against the use of BMI *z*-scores for research using longitudinal designs or for teens with high BMI values, citing concerns that their use could result in spurious differences between groups.^{8,9,11} Children at the extreme ends of BMI distribution require substantially greater changes in weight than their thinner counterparts for the same change in *z*-score. Because the BMI *z*-score curves were constructed using only data between the 3rd and 97th percentiles, caution when using growth curves outside this range is recommended.^{8,11} Berkey and Colditz noted that the difference between *z*-scores reflects larger differences in BMI in older compared to younger children.⁹

Body Fat Percentage—Using standard body fat ranges for children,^{14,39,40} body fat percentage was measured as the amount of body fat as a proportion of body weight by the Tanita DC-430 Body Composition Analyzer. Body fat percentage estimates used the Bioelectrical Impedance Analysis Method (BIA).¹² Although DXA is the best individual method to measure body composition, the use of DXA is not feasible in community-based or school-based settings; in these settings, the BIA methods used by the analyzer have shown acceptable reliability and validity when used with adolescents.^{12,41} In community studies, the Tanita Body Composition Analyzer demonstrated acceptable accuracy for estimating percentage of fat when compared to DXA in adolescent samples.¹² Standard measurement modes were applied for the best reliability.

BMI Percentile for Age and Gender—BMI percentile is used in clinical practice and public health research as an indicator of growth trajectory, growth abnormality, and growth variability.¹⁰ Percentiles describe growth patterns over time and allow one to rank youth relative to others based on age and gender.¹⁰ Using the CDC Calculator, each participant's date of birth, gender, day of data collection, weight to the nearest one-eighth pound, and height in feet and inches to the nearest one-eighth inch were used. Using sex-specific CDC guidelines by age and gender, we calculated BMI percentiles, with healthy weight status defined as between the 5th and 85th percentile; overweight status was defined as between the 85th and 95th percentile; obese was defined as above the 95th percentile; and "extremely obese" was defined as at or above the sex-specific 120% of the 95th percentile.^{13,28,42} Using data from 6 countries, these BMI cutoffs have been found to be statistically equivalent to adult cutoffs for classifying weight status.^{10,14,39} Measuring adiposity via BMI, BMI percentile, changes in weight, fat mass, and body fat percentage provides a more complete analysis.

We conducted a G-RCT to evaluate mentored delivery of a school-based intervention to improve health outcomes in teens from Appalachia Ohio.²² G-RCTs randomly assign identifiable social groups to study conditions. Measures taken on group members assess the

impact of the intervention.^{43,44} G-RCTs are appropriate to evaluate an intervention that manipulates the physical or social environment, involves social processes, or cannot be delivered to individuals without risk of contamination. In our situation, students attending the same school socialize together; a G-RCT was necessary to avoid the risk of cross-contamination. We recruited high schools in 3 waves, with 4 in Wave 1, 8 in Wave 2, and 8 in Wave 3, for a total of 20 schools. For each wave of schools, we randomly assigned half of the schools to each condition—intervention (MBA) and comparison (PBA)—for 10 schools in each of the 2 conditions by study's end. We collected data at baseline (T2), 3 months follow-up (T2), and 6 months follow-up (T3).²²

Sample Power Calculation for Primary Study

We conducted a careful analysis of sample size calculations for primary analysis of BMI in teens at T3. Data from the Ohio Family Health Survey were used to estimate the school-level ICC for BMI among 9th graders in Appalachian Ohio counties at 0.023. We estimated the over-time correlation of BMI measurements at 0.70, and adjusting for age and gender would explain approximately 30% of the variance in BMI. With these assumptions, 10 schools per condition had 82% power to detect a modest intervention effect (0.2 standard deviation difference between groups). This ES would correspond to a difference in mean BMI between groups of 1.04 kg/m² if the observed variation in BMI is similar to that of all Appalachian ninth graders (mean BMI from = 23.41, SD = 5.2). Teens each of the 20 schools are included in this analysis.

Results

Most participants were Caucasian (87.9%) and male (52.6%). See Table 1 for a demographic description of the sample. Teens ranged in age from 14 to 17 years old with the mean age being 15.03 (SD = 0.836) years. Due to the recruitment criteria of the main RCT study, the majority of the sample were in ninth (56.8%) or 10th (41.1%) grade. Students in either ninth or 10th grade (n = 111) were recruited to participate in the intervention (MBA) arm of the main study. Students enrolled in a health course at PBA schools were recruited for participation (n = 76). Most students in health courses at the comparison schools also were in either ninth or 10th grade.

At the beginning of the study, youth were well below the recommendations of 60 minutes of moderate to vigorous physical activity (MVPA) most days of the week. At baseline, both groups' average time in moderate to vigorous activity was only 20.9 minutes per day. At follow-up, obese mentees increased MVPA slightly by 1.2 minutes per day. Extremely obese mentees had the greatest improvement with an increase of MVPA of 3.35 minutes per day. It is important to note that this increase in activity was during student's free time after school. Classroom students decreased MVPA at follow-up. Obese classroom students had 1.78 minutes per day decrease and extremely obese classroom students' MVPA did not change from baseline to follow-up.

Although the focus of PBA and MBA was improving and sustaining MVPA, reduced SSB consumption and increased water consumption were found, especially among the extremely obese students. Extremely obese students decreased SSBs by consuming them fewer days

per week and increased water consumption from less than 5 servings per day to nearly 7 daily servings. Given the results for PA rates, the changes in health outcomes may be partially attributed to these and other dietary changes.

Descriptive Results of BMI, BMI Percentile, and Body Fat Percentage

At baseline, RW ranged from 147.4 to 429.4 pounds. Body fat percentage (BF %) ranged from 20.2% to 56.8%. BMI ranged from 26.8 to 62.5. Over 44% of the subsample were extremely obese at baseline. Of the teens classified as extremely obese, most (58.3%) were in the Class 2 Risk category. Over 34% were in the Class 3 Risk category and the remainder (7.1%) were at the highest risk (Class 4). At baseline, no differences by gender were found for the rates of extremely obese ($P = .885$) or for rates of obesity ($P = .211$). At baseline, the rate of extremely obese youth did not differ between the MBA group and the PBA group ($t = 0.575$, $df[82]$, $P = .567$).

H1: The MBA Group Will Have More Weight Loss at T2 and T3 Compared to the PBA Group

For all youth, an average of 7.3 pounds was lost from T1 to T2 regardless of group assignment. At T3, 10.81 pounds (additional 3.5 pounds) were lost from baseline regardless of group.

Obese (N = 106)—The MBA mentees lost more raw weight at both follow-up time points compared to the PBA group (see Table 2). The MBA group lost 77.5% more total weight compared to the PBA group. These group differences in weight loss were significant. The difference in weight loss between the MBA and PBA groups was clinically meaningful. Obese teens in the MBA group were over 3 times more likely to lose weight relative to the PBA group. Obese mentees lost 4.5% of weight; classroom students lost 2.5% of weight.

Extremely Obese (N = 84)—The MBA group lost more weight at both follow-up time points compared to the PBA group (see Table 2). The MBA group lost nearly 80% more total weight compared to the PBA group. Group differences in weight loss were significant. The difference in weight loss between the MBA group and the PBA group was statistically and clinically meaningful. The odds of extremely obese teens in the MBA losing weight relative to the PBA group were over 4 times ($OR = 4.34$) more likely for mentees. In both groups, females lost more weight at T2 compared to males ($F = 4.75$, $P = .034$). At T3, gender differences were not found ($P = .72$). Youth most at risk for clinical disease, the extremely obese, had the greatest reduction in weight in the MBA group.

H2: The MBA Group Will Have a Greater Reduction of BMI at T2 and T3 Compared to the PBA Group

Regardless of group, the change in BMI was significant ($F = 3.65$, $P = .029$) but also quadratic ($F = 6.5-7$, $P = .013$). Some teens reduced BMI while others increased BMI.

Obese—The MBA group had greater reductions in BMI at both time points. The odds of the MBA group reducing BMI relative to the PBA group were 29.2% more likely for mentees. At T3, the overall change in BMI for the MBA group was -1.17 compared to the PBA group. A meaningful effect on BMI reduction was not found. At initial follow-up,

72.3% of the MBA group reduced BMI, whereas 19.3% increased BMI. Similarly, 69.2% of PBA teens reduced BMI, whereas 28.2% increased BMI. At final follow-up, 66.7% of the MBA group reduced BMI, compared to 65.5% in the PBA group. Comparatively, 33.4% of the MBA group and 34.5% of PBA group increased BMI. In both groups, males had a greater decrease in BMI at initial follow-up (-0.88) compared to females (-0.70). At final follow-up, females had the greatest reduction in BMI (-1.06) compared to males ($-.82$), indicating a delayed response by females relative to males ($F = 9.68, P = .003$).

Extremely Obese—The MBA group had a greater reduction in BMI at both follow-up time points (see Table 2). The odds of the MBA group reducing BMI were over 3 times more likely relative to the PBA group. The difference in BMI reduction between the 2 groups had a clinically meaningful effect. Most importantly, extreme obesity reduced 32.7% in the MBA group compared to 21.9% in the PBA group. Although significantly more in the MBA group began the program as extremely obese ($n = 52$), compared to the PBA group ($n = 32$; $\chi^2 = 4.22, P = .004$), by T2 ($P = .06$) and T3 ($P = .22$) no group differences in extremely obese were found.

At initial follow-up, 84.2% of the MBA group reduced BMI, whereas 18.8% increased BMI. Only 63.3% of the PBA group reduced BMI, whereas 33.3% increased BMI. At final follow-up, 87.5% of the MBA group reduced BMI, compared to 72.7% of the PBA group. Comparatively, 12.5% of the MBA group and 27.3% of PBA group increased BMI. The rate of BMI increase was more than twice as high in the PBA group compared to the MBA group. Males had a greater reduction in BMI at T2 (-1.13) compared to females (-0.83). At T3, females had the greatest decrease in BMI (-1.52) compared to males (-1.17), indicating a delayed response from females. However, gendered differences were not significant ($P = .63$).

H3: The MBA Group Will Have More Fat Loss at T2 and T3 Compared to the PBA Group

Obese—The MBA group lost slightly more body fat compared to the PBA group. This health outcome trended in the hypothesized direction but differences were not significant. Percentage of body fat showed the same trends. The MBA group had a greater reduction of body fat percentage at both T2 and T3 compared to the PBA group. The MBA group who were obese had more than 2 times the odds of reducing their percentage of body fat relative to those who were obese and in the PBA group. See Table 2 for results.

Extremely Obese—The MBA group had more total fat loss compared to the PBA group. The difference between the 2 groups was small but clinically meaningful. At initial follow-up, the MBA group had a slight reduction in percentage of body fat, whereas the PBA group had an increase, but group differences were not significant. The MBA group lost a greater amount of fat pounds at both follow-up time points compared to the PBA group. The odds of the extremely obese assigned to the MBA group losing fat pounds were over twice as likely relative to extremely obese youth in the PBA group (see Table 2).

Discussion

Rural Appalachian teens face several health challenges. Compared to national rates, rural Appalachian adolescents are disproportionately classified as extremely obese. In our primary study, we found that 16.5% of our full sample were extremely obese at baseline. The rate of extreme obesity between males and females was nearly identical. These rates differed from national estimations for teens in 2 important ways: the overall prevalence rates and gender-specific estimations. The overall rates of extreme obesity in teen participants in our primary study exceed the national estimations for Caucasian teens by approximately 3 times the average for males and more than twice the national average for females. According to 2011–2014 NHANES data, the percentage of those extremely obese was 7.8%, with females slightly higher (8%) compared to males (7.7%).³⁸ Other national estimates for teen extreme obesity range from 4% to 6% of all youth.¹³ Nationally, rates of extreme obesity are lower among non-Hispanic teens compared to other teen populations, with males having the lowest rate (5.9%). Despite our population being predominately Caucasian and non-Hispanic, we found rural Appalachian teen extremely obese rates are disproportionately high regardless of gender.

In many communities such as rural Appalachia, intensive behavioral therapy and or medical management of youth obesity are not available or not feasible.^{28,45} Barriers to care include the lack of local program availability, difficulty with transportation to receive specialty care, and the lack of insurance. An alternative is tailored school-based interventions targeting behavioral change and health outcomes. Although our larger 20-school, G-RCT did not target youth with obesity and extreme obesity, baseline measurements indicated that a large proportion were either obese (24.2%) or extremely obese (16.5%). The nature of the program being available to all students avoided the stigma of obesity. Program adaptations such as physical activities, goal setting, and social support strategies accommodated the needs of these high-risk youth. Many began with walking goals and light activities rather than more intensive exercises. Our intervention focused on limiting sedentary activity rather than increasing vigorous activities. Peer mentors worked with mentees based on baseline mentee weight classification; obese and extreme obese youth were not co-mentees with those who were of a healthy weight classification.

Social relationships are a factor related to obesity risk.⁴⁶ Bruening and associates⁴⁶ found that BMI is associated with friendship selection. In a sample of emerging adults, rather than having friends with similar BMIs, college students avoided peers with higher BMI levels. To overcome the discrimination of friend selection for obese youth, peer mentoring is one method to strengthen social support and friendship networks. Mentors help youth overcome personal and social barriers, expose them to new relationships and opportunities, and assist in developing decision-making and problem-solving skills in everyday life.^{35,47} Mentoring relationships have positively influenced behavior change and health outcomes while promoting positive connections to parents and family, including PA^{23,24,30,35,48,49} and academic achievement.⁵⁰ Mentoring of teens has resulted in long-term and sustainable behavior change such as reduced substance use and smoking.^{48,49}

Although we assessed family structure and neighborhood characteristics, the social conditions affecting health behaviors of rural Appalachian communities need to be better understood. Rural Appalachia has environmental, economic, and social characteristics that influence health problems.^{18,19,51} Reduced educational opportunities and high unemployment have led to economic instability and persistent poverty, which is associated with poorer health outcomes.^{1,2,6,18} Recently, Dietz commented on the results of 2 rigorous trials targeting obesity prevention.⁵² He noted that social conditions might partially explain the results. Although these trials were conducted with low-income minority populations, similar social conditions exist within Appalachia. Rural Appalachia is plagued by the opioid crisis affecting many families. During our study, youth shared that parents were either incarcerated or deceased because of substance abuse/overdose. Many teens are responsible for the care of younger siblings. Others shared that they did not have a regular home but temporarily stayed with others. A school principal shared that most of her students lived in a small hotel close by. Our study attrition (N = 43) was mostly due to moving from the school prior to T2 data collection. We did not fully explore these broader structural barriers in our study, but they may have affected health behaviors and intervention sustainability. To overcome broader structural barriers to healthy behaviors, policy-based programs to improve economic conditions and community-driven approaches are needed.⁵² Without research and programing targeting the broader determinants of health-related behaviors, health disparities prevalent in rural Appalachia will persist.

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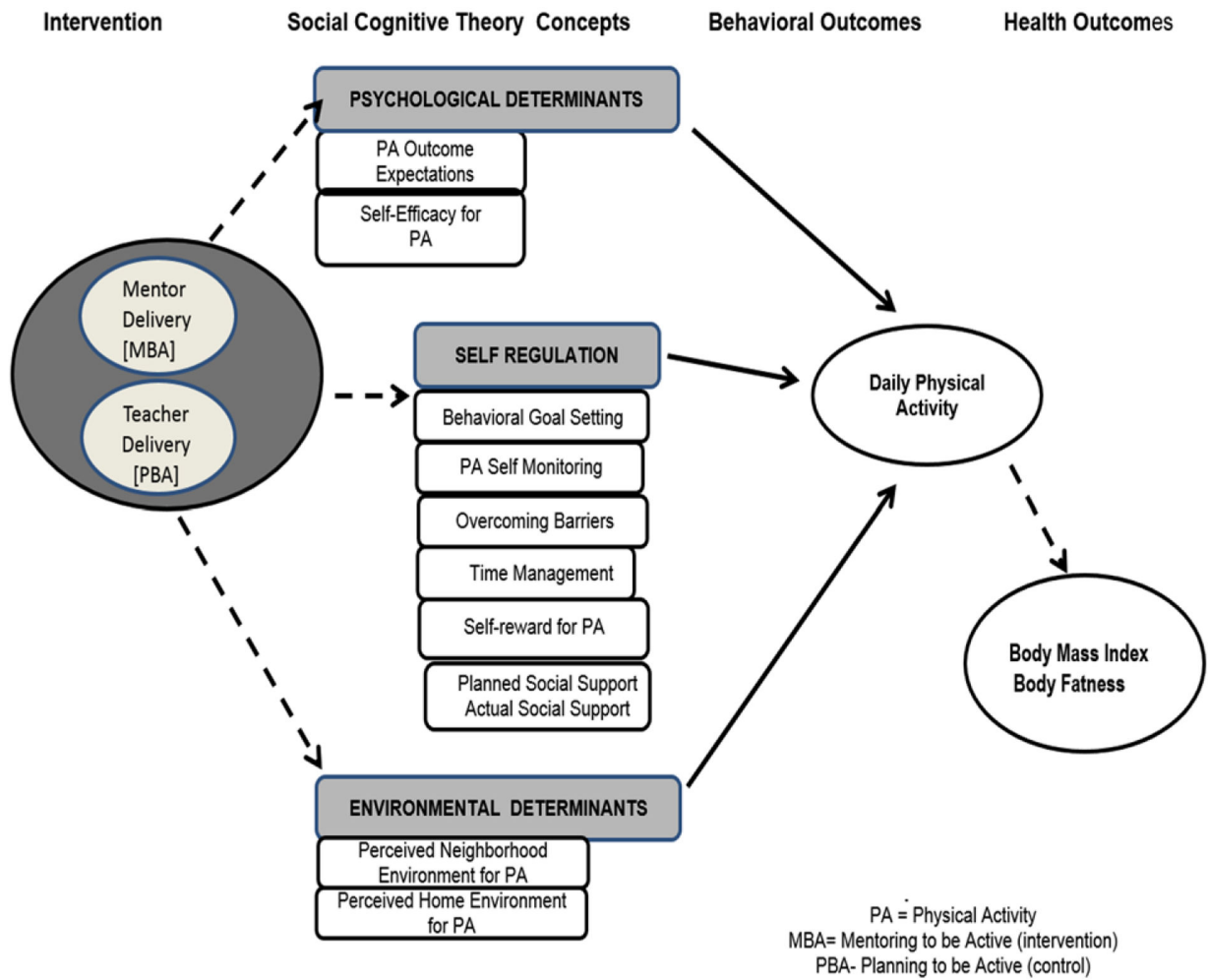


Figure 1. Curricular Components with Social Cognitive Theory Concepts and Outcomes.

Table 1

Demographic Description of the Sample (n = 190)

Demographic Variables	Number	Percent
Age in years		
14	55	28.9
15	82	43.2
16	46	24.2
17	7	3.7
Mean (SD) 15.03 years (0.826)		
Gender		
Male	100	52.6
Female	90	47.4
Grade in school		
9	108	56.8
10	78	41.1
11	4	2.1
Race/ethnicity^a		
White, not of Hispanic origin	167	87.9
American Indian or Native American	15	7.9
Black, not of Hispanic origin	9	4.7
Hispanic or Latino	2	1.1
Asian American, Pacific Islander	1	0.5
Household structure^b		
Lives with Mother	159	83.7
Lives with Father	109	57.4
Lives with at least 1 grandmother	41	21.6
Lives with at least 1 grandfather	29	15.3
Lives with foster parent or legal guardian	1	0.5
Baseline BMI classification		
Obese	106	55.8
Extreme obese	84	44.2
Class 2 ^c	49	58.3
Class 3 ^c	29	34.5
Class 4 ^c	6	7.1
Randomized study condition		
Mentoring to Be active (peer mentoring delivery)	114	60.0
Planning to Be active (classroom teacher delivery)	76	40.0

^aTwo participants selected “White” and “Hispanic or Latino”; percentages >100 due to rounding.

^bParticipants may select more than 1 response.

^cCalculated for those who were initially classified as “Extreme Obese” (n = 84); Obese is defined as (95th sex-specific percentile); Extreme Obese is defined as (120th sex-specific percentile). United States Centers for Disease Control and Prevention (CDC) *BMI-for-age Growth Charts*.

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Table 2
Weight Loss, Change in BMI, Fat Loss and Change in Percentage of Body Fat by Group

Group	Initial Follow-up	Final Follow-up	Group Differences	Effect size	Odds Ratios
Obese—weight loss					
Mentees	-6.1 pounds	-8.7 total lost	T2:($F=8.51, P^{***}$)	0.34	3.32* CI: 1.10–9.98
Classroom	-4.3 pounds	-4.9 total lost	T3:($F=7.62, P^{***}$)		
Extremely obese—weight loss					
Mentees	-10.2 pounds	-13.4 total lost	T2:($F=5.23, P^*$)	0.58	4.34* CI: 0.98–19.34
Classroom	-4.6 pounds	-7.4 total lost	T3:($F=6.33, P^{**}$)		
Obese—change in BMI					
Mentees	-0.843	-1.17 total	T2:($F=4.81, P^{**}$)	0.19	1.29 ns CI: 0.48–3.39
Classroom	-0.586	-0.751 total	T3:($F=1.89, ns$)		
Extremely obese—change in BMI					
Mentees	-1.17	-1.55 total	T2:($F=4.81, P^{**}$)	0.26	3.09* CI: 0.67–10.71
Classroom	-0.73	1.11 total	T3:($F=12.5, P^{***}$)		
Obese—fat loss					
Mentees	-2.1 pounds	-3.1 total	T2:($F=0.92, ns$)	0.11	.76 ns CI: 0.289–2.00
Classroom	-1.9 pounds	3.0 total	T3:($F=0.88, ns$)		
Extremely obese—fat loss					
Mentees	-4.74	-4.93 total	T2:($F=2.98, ns$)	0.20	2.38 ns CI: 0.69–8.26
Classroom	-1.60	-3.68 total	T3:($F=3.86, P^*$)		
Obese—body fat percentage					
Mentees	-0.53%	-0.73% total	T2:($F=8.51, P^{***}$)	0.12	2.19* CI: 0.879–5.46

Group	Initial Follow-up	Final Follow-up	Group Differences	Effect size	Odds Ratios
Classroom	-0.37%	-0.43% total	T3: ($F = 7.62, P^{***}$)		
Extremely obese—body fat percentage					
Mentees	-0.103%	-0.374% total	T2: ($F = 0.583, ns$)	0.12	.988 ns CI: 0.329–2.96
Classroom	+0.127%	-0.118% total	T3: ($F = 0.806, ns$)		

*** $P = .000$;

** $P = .01$;

* $P = .05$

ns = not significant