BMI Trajectories During the Transition to Older Adulthood: Persistent, Widening, or Diminishing Disparities by Ethnicity and Education?

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**Title:** BMI Trajectories during the Transition to Older Adulthood: Persistent, Widening, or Diminishing Disparities by Ethnicity and Education?

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**Ethics Approval:** This study is based on publically available data and was classified as exempt by the University of South Carolina’s Institutional Review Board.
Abstract

Previous research has produced inconsistent results on whether education and ethnic disparities in BMI persist, widen, or diminish over time. We investigate how education and ethnicity, independently and conditionally, influence BMI trajectories during the transition to older adulthood. Employing random coefficient modeling, we analyzed 8 biennial waves of data (1992-2006) from the Health and Retirement Study, a nationally representative longitudinal study of individuals born between 1931 and 1941. After adjusting for health behaviors and health status, education and ethnic disparities in BMI persisted for most groups, but narrowed between high-educated white men and both low-educated Hispanic men and high-educated black men. As such, our findings generally support the persistent inequality interpretation. Therefore, even though interventions targeted at earlier points in the life course may be effective in reducing BMI disparities in later life, social and health policies directed at reducing obesity among older adults is also warranted.
Introduction

Obese persons are at increased risk of cardiovascular disease, hypertension, stroke, diabetes, disability and premature mortality (Alley and Chang 2007; Ferraro, Thorpe, and Wilkinson 2003; Thorpe and Ferraro 2004; Wannamethee, Shaper, and Walker 2005); as such obesity has become a major public health concern. The negative health consequences of obesity may be particularly pronounced among older adults. Prior research suggests that the transition to older adulthood is the period in the life course associated with the greatest risk of severe incident obesity (Ferraro et al. 2003), that obese individuals are at significantly higher risk of becoming disabled across older ages (Reynolds, Saito, and Crimmins 2005), and that obesity during older adulthood accelerates the onset of serious illnesses that place obese older adults at greater risk of early death (Thorpe and Ferraro 2004). Although obesity in mid to late life is likely to have significant health consequences, we know little about how social factors shape changes in body mass index (BMI) during the transition to older adulthood.

Prior studies find that gender, ethnicity, and education affect the rate at which BMI changes over the adult life course (Baltrus et al. 2005; Burke et al. 1996; Clarke et al. 2009; Freedman et al. 2005; Gordon-Larsen et al. 2004; McTigue, Garrett, and Popkin 2002; Mujahid et al. 2005). Most studies of BMI change in adulthood have focused on younger adults and we have only limited and conflicting information about how these social factors affect BMI change among older adults. For example, Burke et al. (1996) found that among young adults, black women were heavier at baseline and gained more weight than white women. Likewise, Clarke et al. (2009) reported that among individuals transitioning to mid-life, BMI increased at a faster rate among the less educated, blacks, and Hispanics. By contrast, Sheehan et al. (2003) found that among a national sample of adults aged 25-74 at baseline, black women experienced a slower
rate of increase in BMI than white women, but black men experienced a faster rate of increase in BMI than white men. Among middle-aged women, Lewis et al. (2005) found that neither race nor education was associated with BMI change over time, whereas Mujahid et al. (2005) found that education was positively associated with increases in BMI over time among black women transitioning to older adulthood. A few studies have also documented differences in the association between education and BMI by ethnicity and gender both cross-sectionally (Chang and Lauderdale 2005; Zhang and Wang 2005) and longitudinally (Burke et al. 1996; Mujahid et al. 2005). These prior findings indicate that ethnicity and education play an important role in how BMI changes across the life course.

Existing knowledge about how BMI disparities change or persist during the transition to older adulthood is hampered by the reliance on cross-sectional studies that confound period, cohort, and age effects, or by the use of longitudinal studies that utilize only two data points (He and Baker 2004; Kahn and Williamson 1991), and hence cannot disentangle random fluctuations from true change (Singer and Willet 2003). The majority of studies that do exploit the longitudinal nature of their data (i.e., assess intra-individual change) often assess change in BMI during early adulthood and middle age (Clarke et al. 2009; Kimm et al. 2005; McTigue et al. 2002; Mo-suwan, Tongkumchum, and Puelpaiboon 2000; Rzehak and Heinrich 2006; Wardle et al. 2006), but not during the transition to older adulthood. Moreover, the studies that have examined BMI change during the transition to older adulthood rely on community samples that may not be generalizable to the US population (Baltrus et al. 2005; Guo et al. 1999; Lewis et al. 2005; Mujahid et al. 2005; Wolinksy et al 2009). We therefore know little about how BMI changes as individuals enter older adulthood, a period of the life course in which individuals experience major life events, such as retirement and widowhood, and declines in health.
We seek to address these gaps in the literature by investigating whether ethnic and educational inequalities in BMI widen, diminish, or persist using a nationally representative sample of the US population making the transition to older adulthood. We also examine if the relationship between education and BMI differs by ethnicity and gender.

**Background**

Overall, studies investigating the education-health gradient have often been based on the assumption that the influence of education on health is static; that is, disparities between highly educated individuals and less educated individuals are maintained over the life course. Recently, however, this assumption has been called into question. Both cross-sectional (Miech and Shanahan 2000; Ross and Wu 1996) and longitudinal studies (Clarke et al. 2009; House et al. 1994; Lynch 2003; Lynch 2006; Walsemann, Gee, and Geronimus 2009; Walsemann, Geronimus, and Gee 2008; Willson, Shuey, and Elder 2007) have documented widening health disparities by levels of education, at least through middle-age. Often referred to as cumulative advantage/disadvantage (or CAD), Dannefer has described this phenomenon as the “systematic tendency for inter-individual divergence in a given characteristic (e.g., money, health, or status) with the passage of time” (2003: 327). As it relates to health, CAD contends that early life disadvantage leads to the accumulation of additional disadvantages which place individuals at greater risk for the earlier onset of illness, thus leading to widening health disparities as individuals age (Dannefer 2003; O’Rand 1996). Related to CAD is the theory of cumulative inequality proposed by Ferraro and Shippee (2009), which argues that the accumulation of disadvantage and its resulting effects are not merely the inverse of cumulative advantage. Rather the theory of cumulative inequality contends that advantage established in early life, such as being white or growing up in a wealthy family, results in the greater accumulation of
opportunities and resources that put already advantaged individuals in a better position to delay the onset of illness. In comparison, disadvantage established in early life, such as being an ethnic minority or growing up in a poor family, is more likely to result in the greater accumulation of disadvantages over the life course that accelerate the aging process and disease onset (Ferraro and Shippee 2009).

Although there has been a clear increase in the number of studies investigating the ways in which health disparities widen, diminish, or persist across the life course as a function of ethnicity or education, few of these studies have focused on BMI. Of the handful of studies that have explored the extent to which ethnic and educational disparities in BMI widen, persist, or diminish, the results have been mixed. For example, in a longitudinal study by Clarke et al. (2009), blacks, Hispanics, and high school graduates experienced greater increases in BMI from late adolescence to middle-age compared to whites and college graduates. As a result, the largest BMI disparities by ethnicity and education occurred at mid-life. Similarly, Kahn and Williamson (1991) reported greater increases in BMI over a ten-year period among black women as compared to white women, and lower educated women as compared to highly educated women. No such effects were found among men. In comparison, Lewis et al. (2005) found no change in BMI disparities by race (i.e., black versus white) or education among a community based sample of middle-aged women living in the United States.

In older adulthood, however, the cumulative effects of ethnicity and education on BMI may actually manifest as persistent or diminishing BMI disparities as the most disadvantaged groups experience earlier onset of disease and premature mortality (Ferraro and Shippee 2009; Lynch 2003). Unintentional weight loss often accompanies declines in health (Wallace and Schwartz 2002); thus, disadvantaged groups who earlier in the life course may have been at
greater risk of obesity compared to advantaged groups could actually see a narrowing of the BMI gap as they enter older adulthood. Additionally, as disadvantaged groups are selected out of the population through premature mortality, the remaining members of the population are typically healthier and more homogeneous. Indeed, some researchers contend that selective mortality is one of the main mechanisms explaining diminishing health disparities as individuals enter old age (Dupre 2007; Dupre 2008; Ferraro and Shippee 2009; Willson, Shuey, and Elder 2007). Thus, although CAD would predict widening disparities in a given health condition with age, cumulative inequality theory would contend that persistent or diminishing BMI disparities by ethnicity or education may actually reflect processes of cumulative inequality that create compositional changes in the population (Ferraro and Shippee 2009).

Declines in health may therefore play an important role in processes of cumulative inequality in BMI. Among younger aged samples, longitudinal studies typically do not adjust for health conditions that may be related to precipitous declines in health, such as cancer, stroke, or cardiovascular events (c.f. Baltrus et al. 2005; Burke et al. 1999; Clarke et al. 2009; Gordon Larsen et al. 2004; James et al. 2006; Kahn and Williamson 1991; McTigue et al. 2002), which are less likely to occur earlier in the life course. Similarly, longitudinal studies using middle-aged and older aged samples do not always fully adjust for these health conditions (c.f. He and Baker 2004; Mujahid et al. 2005; Sheehan et al. 2003), even though processes of cumulative health inequality between advantaged and disadvantaged groups manifest as early as the mid-30’s (Geronimus et al. 2001). Moreover, longitudinal studies that use only two observational points confound issues of mortality with compositional changes in the cohort or population (He and Baker 2004; Wolinsky et al. 2009). Thus, appropriate controls for health conditions and mortality selection are necessary in longitudinal studies of BMI change in older adults.
The theory of cumulative inequality also contends that inequality develops across multiple systems of stratification (Ferraro and Shippee 2009). However, most studies examining BMI disparities tend to treat education, ethnicity, and gender as independent social characteristics that contribute additively to these disparities. Yet, individuals sit at the nexus of multiple sources of inequality (Weber 2010), and it is the combination of these different sources of inequality that can lead to differential distribution of health risks in the population (Krieger et al. 1993; Williams and Collins 1995). By considering only the independent effects of these social characteristics, traditional approaches to studying disparities in BMI trajectories risk misestimating or misunderstanding the effects of ethnicity, education, and gender. For example, examining ethnic differences in BMI and BMI change without considering the additional component of gender variation would result in the mistaken conclusion that blacks, in general, gain weight faster than whites, when in fact this effect exists primarily among women (Baltrus et al. 2005; Burke et al. 1996; Freedman et al. 2005; Gordon Larsen et al. 2004; Kahn and Williamson 1991; McTigue et al. 2002). Processes of cumulative inequality are therefore relevant for understanding BMI trajectories because the experience of multiple systems of inequality can establish these trajectories early in life, but may change over the adult life course as a function of how individuals experience their social inequality. In fact, some prior studies of BMI trajectories among older adults suggest interactions between some combination of gender, race, and education (He and Baker 2004; Mujahid et al. 2005), but the findings are somewhat inconsistent across studies.

The purpose of this study is to examine how BMI changes during the transition to older adulthood as a function of gender, ethnicity, and education. We contend that it is the combination of these social characteristics, in conjunction with age, which determines both exposure and
vulnerability to factors thought to influence BMI. We examine two hypotheses. First, we hypothesize that ethnic and education disparities in BMI will persist or diminish as individuals enter older adulthood, given a higher prevalence of declining health among disadvantaged groups coupled with the expected compositional changes in our sample due to premature mortality among disadvantaged groups. Second, we hypothesize that education disparities in BMI will differ by ethnicity. We do not speculate as to the direction of the difference, as much of the research investigating ethnic differences in the relationship between education and BMI finds conflicting evidence as to the direction of the effect.

Our study builds upon prior work by 1) utilizing multiple waves of data from a longitudinal study spanning 14 years, 2) following a national sample of US adults making the transition to older adulthood, and 3) accounting for differences in health behaviors and health status that may be associated with changes in BMI during the transition to older adulthood.

Methods

Sample

We analyzed data from the Health and Retirement Study (HRS), a nationally representative longitudinal study of individuals born between 1931 and 1941. We utilized data from 8 biennial waves of the HRS (1992 – 2006) (RAND HRS Data 2008). We limited our analyses to respondents who self-reported as non-Hispanic white, non-Hispanic black, or Hispanic (any race) and who reported information on their weight and height at least once. After exclusions, our sample consisted of 9,825 respondents (7,166 non-Hispanic whites, 1,722 non-Hispanic blacks, and 937 Hispanics). Most respondents (70.6%) provided at least 7 waves of data, with approximately 60% providing data in all 8 waves. Only 1% of the sample was excluded due to item non-response.
Over the course of the study 17% of respondents died. Males, black females, the US-born, the less educated, current smokers, respondents in poor/fair health, and those with more than one health condition experienced higher rates of mortality than white females, immigrants, the highly educated, non-smokers, respondents in good/very good/excellent health, and those without a health condition. Baseline BMI was unrelated to mortality.

Analyses are unweighted because longitudinal weights were unavailable (Health and Retirement Study 2009); however, unweighted analyses produce unbiased coefficients if one includes the variables used to sample respondents (Winship and Radbill 1994).

**Measures**

**Dependent Variable**

Respondents’ self-reported their height and weight in each survey year. We used this information to calculate body mass index (BMI) as \([(\text{weight (kg)}/(\text{height (m})^2]])\).

**Ethnicity and Education**

To examine if the influence of education on BMI trajectories varies by ethnicity, we created 6 ethnicity/education dummy variables: White, ≥ 12 years of schooling (referent group); White, < 12 years of schooling; Black, ≥ 12 years of schooling; Black, < 12 years of schooling; Hispanic, ≥ 12 years of schooling; and Hispanic, < 12 years of schooling. More refined specifications of education – a four-category education variable (<12 years; 12 years; 13-15 years; and 16 or more years) and a three-category education variable (<12 years; 12 years; >12 years) - yielded similar results. We used the two category specification of education for parsimony and because the distinction between having a high school education or greater versus having less than high school education is a meaningful comparison for older adults who entered the workforce at a time when most jobs were in manufacturing and a high school education was
sufficient for securing middle-class jobs (Levy and Murnane 1992).

Covariates

Demographic, social, and economic characteristics include birth cohort (0=1931 to 1936 and 1=1937 to 1941), immigrant status (foreign-born versus US-born), and time-varying measures of retirement status (1=retired, 0=other), marital status (married, divorced/separated, widowed, other), and total household wealth (excluding IRAs) categorized as below the 25th percentile (<$43,920), between the 25th to 75th percentile ($43,920 to $329,131) and over the 75th percentile (>=$329,131). Total household wealth reflects the amount of total assets net of all debt. Missing values on individual items used to create the total household wealth measure were imputed using an imputation algorithm described in detail elsewhere (St. Clair et al. 2008). We adjusted total household wealth at each measurement occasion to 2006 dollars to account for inflation. Indicators of attrition include a measure of the number of waves of data collection each respondent missed (0 to 7 waves) and mortality (1=died after 1992, 0=alive).

Because health behaviors and certain health conditions can result in either weight gain or weight loss, and because these behaviors and conditions are more likely to occur among disadvantaged groups, we included the following time-varying measures of health status and behaviors: self-reported health (poor/fair versus good/very good/excellent), number of self-reported doctor diagnosed conditions (high blood pressure, diabetes, cancer, lung disease, heart disease, stroke, and arthritis), smoking status (current smoker versus other), and drinking behavior (ever drinks versus never drinks). The measures were assessed in each survey year and represent current health behaviors or conditions at the time of each survey. All covariates were interacted with age, our measure of time, to adjust for each covariate’s influence on the rate of BMI change.
Analytic Approach

To assess how multiple sources of inequality contribute to differences in BMI trajectories, we examined the effect of ethnicity and education on BMI trajectories in gender stratified analyses. Our analyses proceeded as follows. First, we examined the distribution of socio-demographic and health characteristics of the sample and present the results for each ethnicity and gender group. Next, we estimated two-level random coefficient models to investigate the extent to which ethnicity and education are associated with BMI over time. In our models, age in each survey year represents time. To facilitate interpretation of the model’s intercept, we centered age at 62, the median age of our sample, and modeled age as \([(age-62)/10]\). Thus, a one-unit increase in the age coefficient represents change in BMI over a decade. We tested for non-linearity of BMI change by including a quadratic age variable in our models, but this term was not significant.

Model 1 examined the association between our primary predictor variables (i.e., ethnicity/education, age, and ethnicity/education by age interaction terms), and our dependent variable, after adjusting for birth cohort and immigrant status. Model 2 included additional adjustment for time-varying measures of retirement status, wealth, and marital status to test if economic and social resources mediate the association between ethnicity/education and BMI trajectories. Given prior research that suggests that selective mortality can bias studies attempting to examine processes of cumulative inequalities in health (Dupre 2007; Dupre 2008; Ferraro and Shippee 2009; Ferraro, Shippee, and Schafer 2009), as well as research that finds certain health behaviors and health conditions contribute to weight gain or loss, our final model adjusted for time-varying measures of current smoking and drinking status, self-rated health, health conditions, as well as mortality and unit non-response to determine if the combined effects
of ethnicity and education on BMI at age 62 and over time were attenuated once other correlated measures of health behaviors, health conditions, and attrition were included (Model 3).

To examine how ethnicity and education are associated with BMI and changes in BMI over time, we included interactions between age and each of the ethnicity/education dummy variables. Non-significant differences in the interactions between age and ethnicity/education indicate persistent disparities in BMI; positive coefficients indicate widening disparities in BMI; and negative coefficients indicate diminishing disparities in BMI. We also determined if the effect of education on BMI and BMI change differed by ethnicity by examining the association of education with baseline BMI and BMI change within ethnic groups.

We employed the following random coefficients model:

\[ Y_{it} = \beta X_i + Z_{it}\lambda + \zeta_{0i} + \zeta_{1i}[((age_{it} - 62)/10)] + \epsilon_{it} \]  

(1)

where \( Y_{it} \) is BMI for respondent \( i \) at time \( t \) and assumes that conditional on \( \zeta_{0i} \) and \( \zeta_{1i} \), \( Y_{it} \) to \( Y_{in} \) are independent; \( t = 1, \ldots, T \) is the number of occasions on which respondent \( i \) was observed and \( i = 1, \ldots, n \); \( X_i \) is a vector of time-invariant covariates (e.g., ethnicity/education); \( Z_{it} \) is a vector of time-varying covariates (e.g., age, age x ethnicity/education); \( \zeta_{0i} \) and \( \zeta_{1i} \) are random effects that represent unobserved heterogeneity for respondent \( i \), and are assumed to be normally distributed with mean 0; and \( \epsilon_{it} \) is the random within-person error of prediction for respondent \( i \) at time \( t \). We also assume that the random effects \( \zeta_{0i} \) and \( \zeta_{1i} \) are independent of \( \epsilon_{it} \), and that all random components are independent of the vector of covariates (Singer and Willet 2003).

Random coefficient models were specified using xtmixed in Stata v10 (2007).

**Results**

We present sample characteristics by ethnic/gender groups in Table 1. Black and
Hispanic women reported significantly higher mean BMI scores compared to white women. We found no ethnic differences among men. Regardless of gender, blacks and Hispanics completed less education, had less total household wealth, were more likely to report fair/poor health, and were less likely to drink alcohol compared to whites. Black and Hispanic women and black men reported a higher mean number of health conditions and were less likely to be married compared to their white counterparts, whereas Hispanic men reported a lower mean number of health conditions compared to white men. Hispanic women were less likely to smoke than white females, and black males were more likely to smoke than white males. Hispanics were more likely to be foreign-born than whites. Regardless of ethnicity, men generally had more total household wealth than women. Additional details about the sample can be found in Table 1.

**Random Coefficient Models**

We first examined how ethnicity and education were associated with BMI at age 62 and over time among women (Table 2, Column 1), net of birth cohort and immigrant status. Whites with 12 years or more of schooling served as the referent group. At age 62, all ethnicity/education groups reported higher BMIs than white women with 12 or more years of schooling. Black and Hispanic women, regardless of education, reported higher BMIs than white women. Lower educated women (< 12 years of schooling) experienced higher BMIs than women with 12 or more years of schooling, regardless of ethnicity. These findings suggest that among women in our sample, the association between education and BMI at age 62 was equivalent across ethnicity. Over time, black women, regardless of education, experienced a slower rate of change in BMI than white women with 12 or more years of schooling, as did Hispanic women with less than 12 years of schooling (b=-0.53). We also found secular changes in BMI among our
sample; women in the younger cohort (1937-1941) had higher BMIs at age 62 (b=0.70) and experienced more rapid increases in BMI over a ten-year period (b=0.61) than women in the older cohort (1931-1936).

Next, given that processes of cumulative inequality are hypothesized to affect health through the accumulation of social and economic resources, we further adjusted for time-varying indicators of retirement status, total household wealth, and marital status, which were each interacted with age (Table 2, Column 2). Although the inclusion of these indicators did little to alter the relationship between ethnicity and education on BMI at age 62, it did attenuate the effect of ethnicity and education on BMI change among black women at each educational level. Hispanic women with less than 12 years of schooling continued to experience a slower rate of BMI change (b=-0.41, p<.05) compared to white women with 12 years or more of schooling.

Our final model (Model 3) included additional adjustment for time-varying measures of smoking, drinking, self-rated health, number of health conditions, as well as unit non-response and mortality. Inclusion of these covariates did little to explain the higher BMIs of black and Hispanic women at age 62 as compared to white women. The overall rate of change was attenuated from Model 2 (b=0.60) to Model 3 (b=0.53), as was the rate of change for Hispanic women with less than 12 years of schooling (b=-0.41, Model 2; b=-0.27, p>0.05, Model 3).

We followed the same model building approach for men as we did for women. Net of immigrant status and birth cohort, the only statistically significant difference in BMI at age 62 was among Hispanic men with less than 12 years of schooling. This group experienced higher BMIs (b=0.75) as compared to white men with 12 years or more of schooling (Model 1, Column 4). These findings suggest that neither ethnicity nor education were strongly related to BMI at
age 62 among white or black men. However, education was associated with BMI change among Hispanic men. Like women, men experienced an increase in BMI over a ten-year period (b=0.41), whereas black men, regardless of education, experienced a slower rate of change in BMI compared to white men with 12 years or more of schooling, as did Hispanic men with less than 12 years of schooling (b=-0.53). We also found a significant cohort effect; men in the younger cohort reported higher BMIs at age 62 and a more rapid increase in their BMIs over a ten-year period compared to men in the older cohort.

Estimates from Model 2 yielded comparable results to those found in Model 1, even after additional adjustment for time-varying indicators of retirement status, total household wealth, and current marital status. After further adjustment for time-varying measures of smoking, drinking, self-rated health, number of health conditions, as well as unit non-response and mortality (Model 3), the findings were generally equivalent, except that white men with 12 years of schooling experienced higher BMI at age 62 (b=0.34) than white men with 12 or more years of schooling, whereas BMI change among black men with less than 12 years of schooling was no longer significantly different from white men with 12 or more years of schooling. The results also revealed that education was more strongly associated with BMI change among Hispanic men as compared to white men.

[FIGURES 1 AND 2 ABOUT HERE]

Average BMI trajectories for the youngest cohort of women (Figure 1) and men (Figure 2) are presented by ethnicity/education groups. Disparities by education/ethnicity appear most prominent for women and reflect persistent BMI disparities during the study period. The trajectories also document the strikingly higher levels of BMI found among the most educated black women compared to the least educated white women. Among men, disparities between
education/ethnic groups were narrower than those found among women. The trajectories also show that the gap between Hispanics with less than 12 years of schooling and whites with 12 or more years of schooling narrowed over the study period, from 1.41 BMI units (i.e., 9.6 lbs/4.4 kg for an average height man) at age 51 to 0.60 BMI units (i.e., 4.1 lbs/1.9 kg for an average height man) at age 69 (p>.05 at age 69) (based on unit conversions from Williamson 1993). The gap between black men with 12 or more years of schooling and their white counterparts also narrowed over the study period, from 0.57 BMI units (i.e., 3.9 lbs/1.8 kg) at age 51 to -0.04 BMI units (i.e., -0.27 lbs/-0.13 kg) at age 69 (p>.05 at age 69).

**Discussion**

Our findings of ethnic and education disparities in BMI during the transition to older adulthood generally provide support for our two hypotheses. First, after adjusting for mortality, health conditions, and health behaviors, BMI disparities persisted across ethnic and education groups for women, but narrowed between white men with 12 or more years of schooling and both Hispanic men with less than 12 years of schooling and black men with 12 or more years of schooling. Thus, we find support for our hypothesis that BMI disparities persist or diminish during the transition to older adulthood. Second, the association between education and BMI trajectories differed only for Hispanic men. Thus, we found partial support for our second hypothesis that the relationship between education and BMI would vary by ethnicity.

Our findings can be situated within the larger body of work that has examined how health disparities change or persist as a function of ethnicity and education. The theory of cumulative advantage/disadvantage contends that as the advantaged age, they are more likely to accumulate a reserve of economic and social resources that can assist them in avoiding or delaying the onset of illness or disability (Ross and Wu 1996). As such, health trajectories between advantaged and
disadvantaged groups are expected to diverge with age (Dannefer 2003). Ferraro and Shippee (2009) have argued, however, that selective mortality results in greater homogeneity in a population. Because disadvantaged groups experience less longevity than advantaged groups, selective mortality is a likely explanation for diminishing education and ethnic disparities at older ages (Dupre 2007; Dupre 2008; Farmer and Ferraro 2005). Additionally, among older adults, higher rates of serious illnesses among obese individuals appear to increase the risk of premature mortality (Thorpe and Ferraro 2004).

Our results suggest that selective mortality, health behaviors, and health conditions may partially explain diminishing disparities in BMI between Hispanic women with less than 12 years of schooling and white women with 12 years or more of schooling. That is, after controlling for mortality, health behaviors, and health conditions, these disparities persisted, but did not diminish. Among men, however, adjusting for these additional covariates did little to explain the diminishing disparities found between white men with 12 or more years of schooling and their black counterparts or Hispanic men with less than 12 years of schooling. Although we adjusted for mortality that occurred during the course of the study, we could not account for selective mortality that occurred prior to baseline assessment. Thus, it is possible that our sample had already experienced some level of increasing homogeneity due to selective mortality. Additional studies that follow individuals from early adulthood through late-life are therefore needed to fully understand how education and ethnicity impact BMI disparities across the life course.

In addition to selective mortality, our findings may be influenced by higher levels of declining health among disadvantaged groups. Given that certain illnesses (e.g., cancer) are often accompanied by precipitous weight loss (Wallace and Schwartz 2002), the significantly slower
rate of increase in BMI that we found among blacks with 12 years or more of schooling and Hispanic men with less than 12 years of schooling may reflect the higher rates of certain illnesses experienced by ethnic minority groups in the US and individuals with lower education. We adjusted for a number of health measures (i.e., self-rated health, total number of health conditions); however, these measures may not adequately account for rapidly progressing illnesses that are associated with weight loss.

Among women, the association between education and BMI was fairly consistent across ethnicity. That is, education was associated with lower BMI at age 62, but was generally unrelated to BMI change. Lewis et al. (2005) also found that education was inversely associated with baseline BMI among a community sample of middle-aged black and white women, but was unrelated to changes in BMI over a four year period. Alternatively, findings from another longitudinal study of individuals making the transition to older adulthood documented an inverse association between education and baseline BMI for white and black women, but found that education was unrelated to BMI change for white women and positively associated with BMI change among black women (Mujahid et al. 2005).

Among men, the association between education and BMI varied by ethnicity. Education was inversely associated with BMI at age 62 among whites and Hispanics, and positively associated with BMI change only among Hispanics. Mujahid et al. (2005) reported similar findings for white men, but found that among black men education was positively associated with baseline BMI and this positive association was maintained over time. We know of no longitudinal studies that have examined differential effects of education among Hispanic men making the transition to older adulthood.

Inconsistencies between our findings and the findings reported by Mujahid and
colleagues (2005) may be in part due to differences in the sampling frame. Whereas we use a nationally representative sample of white, black, and Hispanic individuals making the transition to older adulthood, Mujahid et al. (2005) used a sample drawn from four community sites – two of the community samples were virtually all white and one of the community samples was entirely black. Research suggests that neighborhood and community contexts affect individual obesity risk (c.f. Do et al. 2007; Glass, Rasmussen, and Schwartz 2006; Grafova et al. 2008), thus, studies that rely on community samples may not be generalizable to the U.S. population.

When examining the joint effects of education and ethnicity, the results did not follow a consistent pattern. For more highly educated black men (e.g., 12 or more years of schooling), declines in BMI occurred more rapidly than for their white counterparts. If BMI declines are in part related to illness, one might expect that lower educated black men would experience more rapid declines in BMI, as is the case for lower educated Hispanic men. However, for some health outcomes researchers have documented that the greatest health disparities between black and white individuals occur at the highest levels of SES (Farmer and Ferraro 2005). While support for such effects is limited, it is possible that as individuals transition to older adulthood, education may provide greater protection against declining health among whites than among blacks, which could in turn prevent precipitous declines in BMI among whites.

Significant differences in BMI trajectories also emerged by birth cohort. While previous studies have reported similar trends (Flegal et al. 1998; Flegal et al. 2002; Ogden et al. 2006), our findings are quite striking given that birth cohorts in our sample at most differed by five years. Indeed, after 10 years, an “average” white woman with 12 years or more of schooling would be 6.8 pounds (2.2 kilograms) heavier if she were in the younger cohort rather than the older cohort. This gap would be 5.8 pounds (2.6 kilograms) for an “average” white man with 12
or more years of schooling. Thus, BMI prevalence among older adults is likely to increase rapidly as younger cohorts transition to older adulthood.

Although prior research suggests that BMI declines at older ages, our findings did not bear this pattern out (Flegal et al. 2002; Sheehan et al. 2003; Wallace and Schwartz 2002). One possible explanation for this difference is the age range represented in our sample. Half of our sample had not yet turned 70 at the last wave of data collection, the age at which population-level declines in weight become detectable (Wallace and Schwartz 2002).

Our results should be considered in light of study limitations. We were unable to adjust for differences in disability or physical activity, which are known correlates of obesity, because these items were not measured consistently across time in the HRS (St. Clair et al. 2008). Survey items pertaining to disability (e.g., ADL, IADL, and functional limitations), underwent significant modifications across the survey intervals. Rather than exclude data from waves with inconsistent disability measures, as other studies have done (Freedman et al. 2008), we chose to retain baseline observations and to not adjust for disability. Although it would have been preferable to adjust for disability status, we did not want to introduce other, unknown biases into our models by using inconsistent time-varying measures. Likewise, questions about physical activity underwent extensive changes across multiple waves of data collection. Including physical activity measures would have therefore violated a major assumption in longitudinal data analysis – that the variables were measured consistently across time (Singer and Willet 2003).

Our study also utilized self-reported measures of height and weight, which likely underestimates the true weight of respondents, particularly at the extreme of the weight distribution. Given that we compared individuals with themselves over time, the effects of underreporting weight should be minimized (Bowman and DeLucia 1992; Kuskowska-Wolk,
Bergstrom, and Bostrom 1992). Although prior longitudinal studies focusing on the transition to older adulthood have used clinically based measures of height and weight (Lewis et al. 2005; Mujahid et al. 2005; Sheehan et al. 2003; Wolinksy et al. 2009), these studies either relied on community samples that may not be generalizable to the U.S. population, or excluded Hispanics from the analyses (Sheehan et al. 2003). Thus, even though we were unable to utilize clinically based measures of height and weight, our use of a nationally representative sample provides insight into social processes affecting BMI change in the general U.S. population of older adults.

Finally, our results were based on respondents born between 1931 and 1941 who were living in the United States in 1992, and thus can only be generalized to this population. Because the meaning of education has changed over time (Levy and Murnane 1992) the extent to which education affects BMI trajectories may depend on birth cohort. Thus, the association between education and BMI trajectories that we observed in this study may not hold for younger or older cohorts of US adults.

Conclusion

The purpose of our study was to examine the influence of ethnicity and education on BMI and BMI change during the transition to late life. Our results highlight the continued need to investigate processes of cumulative inequality across a range of health behaviors and health outcomes in order to more fully understand how inequality gets “under the skin”. Indeed, given the increased risk of poor health and premature mortality in disadvantaged groups, cumulative inequality in health may manifest not only as widening health disparities as has been traditionally thought, but also manifest as persistent or diminishing disparities depending on the health indicator and the period of life under investigation. Moreover, our results emphasize the potential for accelerated increases in BMI during the transition to older adulthood among successive
cohorts, as well as the importance of ethnicity and education in BMI risk.

Our findings suggest that ethnic and education disparities generally persist during the transition to older adulthood. However, previous longitudinal studies have found widening BMI disparities by education and ethnicity among younger adults (Burke et al. 1996; Clarke et al. 2009). Taken together, this suggests that ethnic and education disparities are established early in life, and widen and persist across much of the life course. Thus, there are likely multiple points in the life course when population-level interventions could effectively reduce or eliminate social disparities in BMI. For example, increasing access to post-secondary education among young, disadvantaged adults may reduce obesity disparities later in life by allowing them the opportunity to accumulate social and economic resources that can promote healthy lifestyle behaviors. Given that a greater proportion of middle-aged adults are returning to school to attain higher education (Choy 2002), expanding access to post-secondary education among non-traditional students may also result in the reduction of obesity disparities at older ages. Additional longitudinal studies are needed, however, to establish a solid body of evidence that can be used to identify which points in the life course population-level interventions could be used to address BMI disparities at older ages.
References


Psychological Sciences and Social Sciences 58B:S327-S337.


Kimm, Sue Y. S., Nancy W. Glynn, Eva Obarzanek, Andrea M. Kriska, Stephen R. Daniels,


Miech, Richard A. and Michael J. Shanahan. 2000. "Socioeconomic Status and Depression over


Stata Corp. 2007. "Stata Statistical Software. Release 10.0." College Station, TX.


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Notes:

a) All indicators dummy coded and estimates can be interpreted as percentages, unless otherwise noted.
b) Time-varying indicators. Indicators are measured at every survey wave, but estimates presented in Table 1 are averaged across survey interval for ease of presentation.

† Significant (at p<.05) sex differences within race (ref=female), two-tailed test
‡ Significant (at p<.05) race differences within sex (ref=white), two-tailed test
Table 2: Estimates from Random Coefficient Models for BMI Stratified by Gender, HRS (1992-2006) a

| At Age 62 | Females (n=5,279) | | | | | | | Males (n=4,546) | | | | |
|-----------|------------------|------------------|------------------|------------------|------------------|------------------|------------------|------------------|------------------|------------------|
|           | Model 1          | Model 2          | Model 3          | Model 1          | Model 2          | Model 3          | Model 1          | Model 2          | Model 3          |
|           | b (SE)           | b (SE)           | b (SE)           | b (SE)           | b (SE)           | b (SE)           | b (SE)           | b (SE)           | b (SE)           |
| Intercept | 26.42 (0.13)***  | 26.41 (0.13)***  | 26.68 (0.14)***  | 27.44 (0.11)***  | 27.43 (0.11)***  | 27.52 (0.12)***  | 27.43 (0.11)***  | 27.43 (0.11)***  | 27.52 (0.12)***  |
| White, ≥12 years b | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- |
| White, <12 years | 1.02 (0.21)***  | 1.02 (0.21)***  | 1.13 (0.21)***  | 0.20 (0.17)     | 0.22 (0.17)     | 0.34 (0.17)*     | 0.20 (0.17)     | 0.22 (0.17)     | 0.34 (0.17)*     |
| Black, ≥12 years | 3.01 (0.26)***  | 3.07 (0.26)***  | 3.04 (0.25)***  | 0.03 (0.24)     | 0.11 (0.24)     | 0.20 (0.24)      | 0.03 (0.24)     | 0.11 (0.24)     | 0.20 (0.24)      |
| Black, <12 years | 4.10 (0.26)***  | 4.18 (0.28)***  | 4.17 (0.28)***  | -0.32 (0.25)    | -0.21 (0.25)    | -0.06 (0.25)     | -0.32 (0.25)    | -0.21 (0.25)    | -0.06 (0.25)     |
| Hispanic, ≥12 years | 1.31 (0.46)**   | 1.33 (0.46)***  | 1.41 (0.45)**   | 0.10 (0.38)     | 0.15 (0.38)     | 0.15 (0.37)      | 0.10 (0.38)     | 0.15 (0.38)     | 0.15 (0.37)      |
| Hispanic, <12 years | 2.54 (0.35)**   | 2.57 (0.35)***  | 2.61 (0.35)***  | 0.75 (0.31)*    | 0.81 (0.31)**   | 0.91 (0.31)**    | 0.75 (0.31)*    | 0.81 (0.31)**   | 0.91 (0.31)**    |
| Birth Cohort 1937-1941 c | 0.70 (0.15)***  | 0.70 (0.15)***  | 0.61 (0.15)***  | 0.66 (0.13)***  | 0.67 (0.13)***  | 0.59 (0.13)***   | 0.66 (0.13)***  | 0.67 (0.13)***  | 0.59 (0.13)***   |
| Rate of Change | | | | | | | | | |
| Age-62 d | 0.60 (0.07)***  | 0.60 (0.07)***  | 0.53 (0.08)***  | 0.41 (0.06)***  | 0.43 (0.07)***  | 0.48 (0.05)***   | 0.43 (0.06)***  | 0.43 (0.07)***  | 0.48 (0.05)***   |
| (White, ≥12 years) x Age | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- | ---ref---- |
| (White, <12 years) x Age | -0.17 (0.12)   | -0.09 (0.12)    | 0.01 (0.12)     | -0.06 (0.11)    | -0.05 (0.11)    | 0.04 (0.11)      | -0.06 (0.11)    | -0.05 (0.11)    | 0.04 (0.11)      |
| (Black, ≥12 years) x Age | -0.33 (0.14)*  | -0.27 (0.14)    | -0.24 (0.14)    | -0.41 (0.15)**  | -0.40 (0.15)**  | -0.34 (0.15)*    | -0.41 (0.15)**  | -0.40 (0.15)**  | -0.34 (0.15)*    |
| (Black, <12 years) x Age | -0.37 (0.16)*  | -0.24 (0.16)    | -0.11 (0.16)    | -0.32 (0.16)*   | -0.32 (0.16)*   | -0.20 (0.16)     | -0.32 (0.16)*   | -0.32 (0.16)*   | -0.20 (0.16)     |
| (Hispanic, ≥12 years) x Age | -0.26 (0.25)   | -0.22 (0.25)    | -0.12 (0.25)    | 0.10 (0.22)     | 0.08 (0.22)     | 0.11 (0.22)      | 0.10 (0.22)     | 0.08 (0.22)     | 0.11 (0.22)      |
| (Hispanic, <12 years) x Age | -0.53 (0.20)** | -0.41 (0.20)*   | -0.27 (0.20)    | -0.58 (0.19)**  | -0.58 (0.19)**  | -0.45 (0.19)*    | -0.58 (0.19)**  | -0.58 (0.19)**  | -0.45 (0.19)*    |
| Birth Cohort 1937-1941 x Age | 0.61 (0.08)*** | 0.57 (0.09)***  | 0.56 (0.09)     | 0.41 (0.08)**   | 0.32 (0.08)**   | 0.26 (0.08)**    | 0.41 (0.08)**   | 0.32 (0.08)**   | 0.26 (0.08)**    |
| Log Likelihood | -77933.8       | -77906.4       | -77725.4       | -54622.8       | -54575.1       | -54402.7        | -54622.8       | -54575.1       | -54402.7        |
| SD (u0i)    | 5.41           | 5.41           | 5.32           | 4.20           | 4.19           | 4.14            | 4.20           | 4.19           | 4.14            |
| SD (u1i)    | 2.26           | 2.25           | 2.20           | 1.92           | 1.91           | 1.86            | 1.92           | 1.91           | 1.86            |
| Person-Period Observations | 32682          | 25397          |               |               |               |               |               |               |               |
| Mean Number of Observations | 6.2 (min=1, max=8) | 5.6 (min=1, max=8) | | | | | | | |

Notes: a Model 1 also adjusted for immigrant status. Model 2 also adjusted for time-varying measures of current retirement status, current marital status, and current total household wealth. Model 3 also adjusted for time-varying measures of current smoking, current drinking, current self-rated health, and current number of health conditions, as well as unit non-response, and mortality. All covariates, except for immigrant status, mortality, and unit non-response, were centered at their grand means. All covariates were interacted with age. b ≥12 years and <12 years refers to years of schooling completed. c Reference group: 1931-1936 cohort. d Age is modeled as [(age-62)/10] for main effects and interactions. *p≤0.05, **p≤0.01, ***p≤0.001
Figure 1: Average BMI Trajectories for Female Respondents in 1937-1941 Cohort, by Ethnicity and Education (Model 3)

Notes: Average trajectories based on Model 3 estimates. All covariates except for immigrant status, mortality, and attrition were centered at their grand mean.
Figure 2: Average BMI Trajectories for Male Respondents in 1937-1941 Cohort, by Ethnicity and Education (Model 3)

Notes: Average trajectories based on Model 3 estimates. All covariates except for immigrant status, mortality, and attrition were centered at their grand mean.