Residential Property Values Predict Prevalent Obesity but Do Not Predict 1-Year Weight Change

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Objective: Lower socio economic status (SES) has been linked with higher obesity rates but not with weight gain. This study examined whether SES can predict short-term weight change.

Methods: The Seattle Obesity Study II was based on an observational cohort of 440 adults. Weights and heights were measured at baseline and at 1 year. Self-reported education and incomes were obtained by questionnaire. Home addresses were linked to tax parcel property values from the King County, Washington, tax assessor. Associations among SES variables, prevalent obesity, and 1-year weight change were examined using multivariable linear regressions.

Results: Low residential property values at the tax parcel level predicted prevalent obesity at baseline and at 1 year. Living in the top quartile of house prices reduced obesity risk by 80% at both time points. At 1 year, about 38% of the sample lost >1 kg body weight; 32% maintained (±1 kg); and 30% gained >1 kg. In adjusted models, none of the baseline SES measures had any impact on 1-year weight change.

Conclusions: SES variables, including tax parcel property values, predicted prevalent obesity but did not predict short-term weight change. These findings, based on longitudinal cohort data, suggest other mechanisms are involved in short-term weight change.

Introduction

Prevalent obesity in the US is associated with lower socio economic status (SES), especially among women (1,2). However, the same social gradient has not been observed for body weight gain (3). All segments of the US population are reported to have experienced the same rise in obesity rates, regardless of age, gender, race/ethnicity, education, or incomes (2,4,5).

Whether some SES groups have had higher rates of incident obesity than others cannot be readily determined from US federal surveys. The National Health and Nutrition Examination Survey (NHANES) (2,6) and the Behavioral Risk Factor Surveillance System (BRFSS) (4,7) are cross-sectional studies that are only able to compare different socio demographic groups at one point in time. To better assess the impact of SES variables on incident obesity or on short- or long-term weight gain, longitudinal cohorts are required (4).

However, longitudinal cohort studies on weight gain have produced inconsistent results. Some studies showed no significant links between SES variables and long-term weight gain (8,9), while others reported significant effects of SES (10-14). European studies, in particular, have noted that lower SES was associated with higher obesity rates and with greater weight gain among both children and adults (14-16). By contrast, US-based studies have reported only a limited impact of low SES on incident obesity and on body weight gain (17-19). The common consensus in the US is that disparities by income seem to be weakening with time (17).

Although cross-sectional studies have pointed to social disparities in obesity rates (20), they cannot prove that low SES is the direct cause of weight gain (21). Most of the published studies have relied, moreover, on self-reported SES, as indexed by education and incomes. The Seattle Obesity Study II (SOS II) was based on a well-characterized longitudinal cohort, with heights and weights measured over the period of 1 year. The present study is unique in multiple ways: (a) it introduced tax parcel residential property values from the county tax assessor as an objective measure of SES (22); (b) it examined whether multiple measures of SES (both self-reported and objective) predict prevalent obesity and longitudinal weight change using the same cohort.

Methods

Participant sampling and recruitment

The sample frame was established by selecting residential units from about 450,000 tax parcels (land lots) within King County Urban Growth Boundary (UGB). UGB refers to the urbanized area of King County, Washington, and is used by the local government for zoning and land use decisions. Cadastral land parcels (tax lots) are used by local government to collect real estate taxes. The provision of urban services: roads,
water, sewer, and fire and police services is limited by the UGB. Residential units in three bands of residential property values: <199k; >200k–<299k; and ≥300k were weighted to provide equal distributions by SES and were apportioned to correspond to the county distribution of 58% as single-family units and 42% as multifamily units. The property value cut offs were based on past results linking property values with income and health outcomes such as self-reported health and obesity rates (21,23,24). The selected residential units were sent to a commercial supplier for reverse telephone matching of addresses with telephone numbers. The matching rate was 55% for single and 40% for multifamily categories. After eliminating duplicates and incomplete records, 25,460 addresses and phone numbers were provided to the Battelle Memorial Institute Survey Research Group for telephone screening of potential study participants.

Battelle mailed out pre-notification postcards to potential participants and followed with up to three telephone calls. Eligible participants were English speaking, aged 18–55 years, were primary food shoppers in their household, and did not have mobility issues. Contact information for 712 potential eligible respondents who provided the verbal consent to participate was then provided by Battelle to SOS II research staff.

Data collection
Eligible persons, identified by Battelle, were contacted by phone and were invited to an in-person meeting. Of the eligible participants, 516 (72.5%) agreed to enroll in the study. To minimize response bias and to allow participation by working mothers, single parents, and lower-income groups, respondents were given the choice of location for the first in person meeting: the UW location or another location of their choice, including at his/her home. About 56% of study participants (n = 291) chose the latter option. At the first in-person visit, participants provided written consent before data collection and were then weighed and measured. Participants were compensated for completing all baseline protocols and for completing the second visit, 12 months later. The protocol was approved by the institutional review board (IRB) at the University of Washington.

Measured heights and weights
Height and weight was measured for each participant at baseline and at 1 year. Weight was measured in street clothes, without shoes, using a portable scale with a capacity of 350 lbs. Two participants (out of 440) exceeded the 350 lb limit (the maximum limit allowed on the weighing scale); in these cases, participant self-reported weight was used instead. Height was measured using a portable stadiometer. Following past studies using weight change over the period of 1 year as the outcome of interest (25), small weight gain was defined as ≥1 kg to <2 kg; large weight gain was defined as ≥2 kg; small weight loss as ≤−1 kg to >−2 kg; large weight loss as ≤−2 kg; and weight maintenance as ±1 kg.

Health behaviors survey
A computer-aided questionnaire was administered during the first in person visit. Data were collected on socio demographics, self-rated health and weight, and a variety of dietary, food shopping, and health behaviors. Many of the questions were based on the Behavioral Risk Factor Surveillance System (BRFSS). Demographic variables included age, gender, race/ethnicity, household size, and home ownership. For analyses, annual household income was grouped into three categories: low (<$50,000/year), medium ($50,000–<$100,000/year), and high (≥$100,000/year). Educational attainment was also grouped into three categories: <12 years (“high school or less”) 12-16 years (“some college”), and >16 years (“college graduates or higher”). Further questions addressed smoking: 72.8% of the sample never smoked; 25.3% were former smokers; and only 2.0% were current smokers.

Follow-up at 1 year
After 11 months, baseline participants were re-contacted by the same SOS II research staff member to schedule the follow-up visit. Participants who did not become pregnant and who still lived in King County were followed after 1 year. Participants were again given the choice of either meeting at the UW research site or at a location convenient to them. Of the 516 baseline participants, 478 participants completed the 2nd appointment, a lower attrition rate than expected. Participants completed a follow-up Health Behaviors Questionnaire (HBQ) and were weighed and measured for the second time. The follow-up HBQ asked participants if they had tried to lose weight over the previous year and if they had tried not to gain weight over the previous year. Data analyses were based on those participants for whom complete data at the two time points were available, including self-reported household incomes, property values, demographic, and behavior data from both questionnaires, and measured heights and weights (n = 440). Participants were compensated for completing all baseline protocols and for completing the second visit.

Residential property values
Data from King County tax assessor records for 2008 were used to develop the residential property value metric at the tax parcel level. When parcels had multiple units, the mean value per unit was used (21,23). The use of this method has been documented before (23).

Statistical analysis
We calculated unadjusted proportions, means, and standard deviations for prevalent obesity and longitudinal weight change variables, by key socio demographic variables. P-values were computed to test for the overall statistical significance of the bivariate associations. A series of linear and logistic regressions were used to examine the associations between SES measures, prevalent obesity, and weight change. Income, education, and residential property values were used as the three primary independent variables. Prevalent obesity at baseline and at 1 year, and percent weight change over 1 year were each used as the dependent variables of interest. First, we fit three logistic regression models assessing the bivariate relationship between income, education, and residential property values on prevalent obesity, after adjusting for age, gender, and race/ethnicity. Second, we fit one regression model to evaluate whether income and education were independently associated with prevalent obesity. The last and the final model further included residential property values to evaluate which of the three SES measures are independently associated with prevalent obesity. All results presented were calculated using robust standard errors. The same series of models were replicated for three other outcome measures; prevalent obesity at 1 year and percent weight change at 1 year. All analyses used Stata 10.1 (College Station, TX). An z level of 0.05 was used to define statistical significance.
Table 1 shows that prevalent obesity at baseline was strongly and inversely linked to the three SES measures: residential property values, education, and incomes. Similarly, prevalent obesity at 1 year was strongly and inversely linked to baseline residential property values, education, and incomes. No short-term weight gain (1 year) was observed in this population sample; on the contrary, there was a minor short-term weight loss (mean 0.46% body weight). In bivariate analyses, percent weight change was unrelated to SES.

Figure 1 shows large weight changes occurring in the short term. About 38% of the study sample lost >1 kg body weight, 32% maintained weight (±1 kg), and about 30% gained >1 kg body weight. About 25% of the sample had a large weight loss of >2 kg whereas 21% had a large weight gain of > 2 kg. When questioned about their weight behaviors over the previous year, 233 participants reported trying to lose weight, whereas 222 said they were not trying to lose weight. The majority of study participants reported trying not to gain weight over the previous year (291 yes; 165 no). The direction or the magnitude of 1-year weight change was not linked to residential property values.

As shown in Table 2, multivariable regressions explored the association between SES and prevalent obesity at baseline. In adjusted models, property values showed the strongest association with prevalent obesity, followed by education. Self-reported household income was no longer significant once property values were included in the model. Living in the highest quartile of house values was associated with an 80% drop in obesity rates.

Table 3 shows the associations between SES and prevalent obesity at 1 year. In adjusted models, property values showed the strongest association with prevalent obesity, followed by education. Self-reported household income was no longer significant once property values were included in the model. Living in the

### Table 1 Characteristics of the SOS II population sample

| Total, n | Prevalent obese at baseline, n (%) | Prevalent obese at 1 year, n (%) | % weight change at 1 year, mean (SD) |
|----------|-----------------------------------|-----------------------------------|-------------------------------------|
| N (%)    | 438                               | 145 (33.1)                        | 133 (30.3)                          | −0.46 (4.53) |
| Gender   |                                   |                                   |                                     |               |
| Men      | 136                               | 43 (31.6)                         | 42 (30.8)                           | −0.28 (3.93) |
| Women    | 302                               | 102 (33.7)                        | 91 (30.1)                           | −0.54 (4.77) |
| P-value  |                                   | 0.657                             | 0.875                               | 0.583        |
| Age categories (years) |                                   |                                   |                                     |               |
| 21-39    | 70                                | 18 (25.7)                         | 14 (20.0)                           | −0.05 (3.92) |
| 40-49    | 198                               | 63 (31.8)                         | 58 (29.2)                           | −0.24 (4.59) |
| ≥ 50     | 170                               | 64 (37.6)                         | 61 (35.8)                           | −0.89 (4.67) |
| P-value  |                                   | 0.064                             | 0.015                               | 0.128        |
| Race/ethnicity |                               |                                   |                                     |               |
| Whites   | 351                               | 107 (30.4)                        | 101 (28.7)                          | −0.30 (4.62) |
| Non-Whites | 87                            | 38 (43.6)                         | 32 (36.7)                           | −1.12 (4.08) |
| P-value  |                                   | 0.020                             | 0.147                               | 0.129        |
| Annual household income ($) |                                   |                                   |                                     |               |
| <35,000  | 77                                | 34 (44.1)                         | 31 (40.2)                           | −1.00 (5.35) |
| 35,000-<50,000 | 51                           | 21 (41.1)                         | 19 (37.2)                           | −0.27 (3.83) |
| 50,000-<100,000 | 157                        | 53 (33.7)                         | 46 (29.3)                           | −0.63 (4.94) |
| ≥100,000 | 153                               | 37 (24.1)                         | 37 (24.1)                           | −0.09 (3.80) |
| P-value  |                                   | 0.001                             | 0.007                               | 0.200        |
| Education |                                   |                                   |                                     |               |
| High school or less | 46                          | 24 (52.1)                         | 21 (45.6)                           | −0.42 (6.14) |
| Some college | 115                          | 54 (46.9)                         | 52 (45.2)                           | −0.56 (5.26) |
| College 4 years | 154                          | 43 (27.9)                         | 36 (23.3)                           | −0.39 (3.89) |
| College 5 years or more | 123                         | 24 (19.5)                         | 24 (19.5)                           | −0.47 (3.83) |
| P-value  |                                   | <0.0001                           | <0.0001                             | 0.935        |
| Residential property value |                               |                                   |                                     |               |
| Category 1 ($26,855-204k) | 110                     | 51 (46.3)                         | 47 (42.7)                           | −0.61 (5.26) |
| Category 2 ($205k-268k) | 108                     | 44 (40.7)                         | 37 (34.2)                           | 0.36 (3.85)  |
| Category 3 ($269k-378k) | 111                     | 36 (32.4)                         | 37 (33.3)                           | −1.02 (4.79) |
| Category 4 ($379k-914k) | 109                     | 14 (12.8)                         | 12 (11.0)                           | −0.56 (3.99) |
| P-value  |                                   | <0.0001                           | <0.0001                             | 0.525        |
The observed SES gradient for prevalent obesity in the US was confirmed in the SOS II study, using both the self-reported measures of SES (income and education) and an objective measure (residential property values at the tax parcel level). Consistent with other studies, residential property values predicted prevalent obesity better than self-reported household incomes (26), an often problematic variable in health surveys (27,28).

The present data were also consistent with some past interpretations of cross-sectional federal surveys (3,17). In those surveys, prevalent obesity was linked to lower SES; yet patterns of weight gain were reported for all SES groups (17,29). In the SOS II, SES variables had no impact on the direction or the magnitude of short-term weight change, measures over only 1 year. Residential property values and education, powerful predictors of prevalent obesity, did not predict the direction or the magnitude of 1-year weight change.

Consistent with other federal reports (30), no increase in obesity prevalence was observed. To the contrary, there was a very minor weight loss of 0.46% body weight. However, that seeming stability of population mean weight masked significant gains or losses at the individual level. Participants in this purely observational (i.e., non-intervention) study were equally split among those who gained, lost, or maintained body weight. In past studies (31), weight gain of 0.5-2 kg per year was classified as moderate and a weight gain of >2 kg per year was classified as large. In the present cohort, one in four participants lost >2 kg, whereas one in five gained >2 kg.

**Discussion**

The observed SES gradient for prevalent obesity in the US was confirmed in the SOS II study, using both the self-reported measures of SES (income and education) and an objective measure (residential property values at the tax parcel level). Consistent with other studies, residential property values predicted prevalent obesity better than self-reported household incomes (26), an often problematic variable in health surveys (27,28).

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Even obesity prevention studies have observed weight loss and weight gain within the same cohort. In one study (32), 15% of the cohort (n = 823) lost 5% percent body weight while 9.3% gained >5% body weight. The authors suggested that short-term weight change (1 year) was a phenomenon that deserved more research attention.

**Table 3** Multivariate logistic regression to examine the association between prevalent obesity at 1-year follow-up and key SES indicators

|                          | Model 1 |           |          | Model 2 |           |          | Model 3 |           |          |
|--------------------------|---------|-----------|----------|---------|-----------|----------|---------|-----------|----------|
|                          | OR      | 95% CI    | P-value  | OR      | 95% CI    | P-value  | OR      | 95% CI    | P-value  |
| **Annual household income ($)** |         |           |          |         |           |          |         |           |          |
| <35,000                  | 1       | 1         | 1        | 1       | 1         | 1        | 1       | 1         | 1        |
| 35,000–<50,000           | 0.93    | 0.44, 1.96| 0.868    | 1.17    | 0.55, 2.51| 0.670    | 1.36    | 0.60, 3.05| 0.448    |
| 50,000–<100,000          | 0.62    | 0.35, 1.12| 0.117    | 0.81    | 0.43, 1.50| 0.509    | 1.04    | 0.54, 2.02| 0.895    |
| ≥100,000                 | 0.48    | 0.26, 0.87| 0.017    | 0.76    | 0.39, 1.46| 0.412    | 1.35    | 0.64, 2.81| 0.422    |
| **Education**            |         |           |          |         |           |          |         |           |          |
| High school or less      | 1       | 1         | 1        | 1       | 1         | 1        | 1       | 1         | 1        |
| Some college             | 0.95    | 0.47, 1.92| 0.907    | 0.97    | 0.49, 1.97| 0.968    | 0.91    | 0.44, 1.87| 0.799    |
| College 4 years          | 0.37    | 0.18, 0.74| 0.005    | 0.39    | 0.19, 0.79| 0.010    | 0.42    | 0.20, 0.88| 0.022    |
| College 5 years or more  | 0.30    | 0.14, 0.62| 0.002    | 0.34    | 0.15, 0.73| 0.006    | 0.37    | 0.16, 0.82| 0.015    |
| **Residential property value** |       |           |          |         |           |          |         |           |          |
| Category 1 ($27k–204k)   | 1       | 1         | 1        | 1       | 1         | 1        | 1       | 1         | 1        |
| Category 2 ($205k–268k)  | 0.69    | 0.39, 1.22| 0.21     | 0.74    | 0.40, 1.37| 0.350    |         |           |          |
| Category 3 ($269k–378k)  | 0.66    | 0.37, 1.15| 0.146    | 0.75    | 0.40, 1.42| 0.388    |         |           |          |
| Category 4 ($379k–914k)  | 0.15    | 0.07, 0.32| 0.000    | 0.18    | 0.08, 0.42| 0.000    |         |           |          |

Model 1: Adjusted for age + gender + race/ethnicity.
Model 2: Model 1 + mutually adjusted for income and education.
Model 3: Model 2 + adjusted for residential property values.

**Table 4** Multivariate linear regression to examine the association between 1-year weight change and key SES indicators

|                          | Model 1 |           |          | Model 2 |           |          | Model 3 |           |          |
|--------------------------|---------|-----------|----------|---------|-----------|----------|---------|-----------|----------|
|                          | β       | 95% CI    | P-value  | β       | 95% CI    | P-value  | β       | 95% CI    | P-value  |
| **Annual household income ($)** |         |           |          |         |           |          |         |           |          |
| <35,000                  | 0       |           |          | 0       |           |          | 0       |           |          |
| 35,000–<50,000           | 0.63    | −0.99, 2.26| 0.444    | 0.70    | −0.90, 2.31| 0.383    | 0.66    | −0.96, 2.29| 0.424    |
| 50,000–<100,000          | 0.34    | −1.11, 1.80| 0.643    | 0.44    | −1.00, 1.90| 0.546    | 0.48    | −1.00, 1.98| 0.519    |
| ≥100,000                 | 0.90    | −0.50, 2.31| 0.207    | 1.08    | −0.32, 2.48| 0.133    | 1.36    | −0.14, 2.87| 0.077    |
| **Education**            |         |           |          |         |           |          |         |           |          |
| High school or less      | 0       |           |          | 0       |           |          | 0       |           |          |
| Some college             | −0.09   | −2.13, 1.94| 0.926    | −0.16   | −2.15, 1.83| 0.874    | −0.28   | −2.28, 1.70| 0.777    |
| College 4 years          | −0.07   | −2.00, 1.85| 0.935    | −0.31   | −2.15, 1.52| 0.733    | −0.28   | −2.10, 1.54| 0.762    |
| College 5 years or more  | −0.14   | −2.09, 1.79| 0.881    | −0.52   | −2.38, 1.34| 0.583    | −0.46   | −2.31, 1.37| 0.617    |
| **Residential property value** |       |           |          |         |           |          |         |           |          |
| Category 1 ($27k–204k)   | 0       |           |          | 0       |           |          | 0       |           |          |
| Category 2 ($205k–268k)  | 0.94    | −0.30, 2.19| 0.138    | 0.73    | −0.51, 1.98| 0.248    |         |           |          |
| Category 3 ($269k–378k)  | −0.45   | −1.82, 0.90| 0.510    | −0.78   | −2.18, 0.62| 0.275    |         |           |          |
| Category 4 ($379k–914k)  | 0.06    | −1.22, 1.35| 0.924    | −0.53   | −1.92, 0.86| 0.452    |         |           |          |

Model 1: Adjusted for age + gender + race/ethnicity.
Model 2: Model 1 + mutually adjusted for income and education.
Model 3: Model 2 + adjusted for residential property values.
The present use of residential property values, an objective measure of SES, has implications for future studies on obesity and the built environment (33). In this study, tax parcel property values served as an individual SES metric, complementing self-reported education and incomes (23,27,28). However, neighborhood property values, a related measure defined as the mean residential property value within certain distances from respondent’s residence, can reflect the quality of the surrounding built environment (24). In other studies, rather than using residential property values at the tax parcel level, we have used average residential property values for all dwellings within a 400 m or an 800 m buffer of the study respondent’s residence. In one Seattle-based study (24), multiple perceived aspects of “obesogenic” environments were linked to neighborhood residential property values in a predictable manner. Whereas perceived traffic, crime, and the reported presence of fast foods and convenience stores were linked to lower property values within 800 m, the presence of parks was linked to higher property values (23). In turn, lower neighborhood residential property values were associated with higher prevalent obesity among women (21,26).

Multiple studies have tried to link rising obesity rates with selected features of the built environment, with mixed results (21). Even though the data were almost entirely cross-sectional, the implication was that living in an “obesogenic” neighborhood was causally linked to weight gain. Whether objective features of the built environment, as captured by neighborhood property values can predict weight gain remains to be seen.

The present study had several limitations. The chief limitation was that duration of exposure was only 1 year; consequently, our conclusions only apply to short-term weight change. Ideally, weight changes need to be studied over a much longer period. There may have been selection bias in study enrollment; however, the direction of weight change does not show that the study sample was predisposed to weight loss. The SES variables were obtained only at baseline—we do not have data on how changes in SES would affect body weight change.

The present data have some public policy implications. First, low individual SES did not predispose to weight gain and was no barrier to weight loss, at least in the short term. These findings offer significant new hope for community-level obesity prevention and control. While long-term obesity may be strongly influenced by SES (23,24), weight dynamics may be more closely linked to individual attitudes, dietary behaviors, diet quality, and diet cost. More studies are needed to determine what dietary and behavioral mechanisms can potentiate or override the observed SES constraints on body weight.

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