

# Public Transit, Obesity, and Medical Costs: Assessing the Magnitudes

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## Abstract

*Objective.* This paper assesses the potential benefits of increased walking and reduced obesity associated with taking public transit in terms of dollars of medical costs saved and disability avoided.

*Methods.* I conduct new analysis of a nationally representative U.S. transportation survey to gauge the net increase in walking associated with public transit usage. I translate minutes spent walking into energy expenditures and reductions in obesity prevalence, estimating the present value of costs and disability that may be avoided.

*Results.* Taking public transit is associated with walking 8.3 more minutes per day on average, or an additional 25.7–39.0 kcal. Hill et al. (2003) estimate an increase in net expenditure of 100 kcal/day can stop the increase in obesity in 90% of the population. Additional walking associated with public transit could save \$5,500 per person in present value by reducing obesity-related medical costs. Savings in quality-adjusted life years could be even higher.

*Conclusions.* While no silver bullet, walking associated with public transit can have a substantial impact on obesity, costs, and well-being. Further research is warranted on the net impact of transit usage on all behaviors, including caloric intake and other types of exercise, and on whether policies can promote transit usage at acceptable cost.

Keywords: obesity; walking; exercise; transportation; health expenditures

# 1 Introduction

A topic of much recent interest is the degree to which public transportation may increase exercise through walking. Other things equal, an increase in exercise could then improve health outcomes by lowering obesity, which many view as a looming but potentially manageable threat to public health (Hill et al., 2003; Olshansky et al., 2005; Preston, 2005). The “New Urbanism” movement of the 1990s, which continues today, calls for development of denser, grid-based neighborhoods in order to increase walking, bicycling, and use of transit (Cervero and Radisch, 1996). More recently, the entire January 2007 issue of *Environment and Behavior* was devoted to examining how the built environment relates to diet and exercise, and thus to obesity. A joint study by the Transportation Research Board and the Institute of Medicine (2005) surveyed the state of knowledge regarding the built environment and physical activity. While the study showed that the built environment, including access to public transit, can help or hinder the choice to engage in physical activity, it emphasized how the lack of good data and inadequate study design have significantly hampered inference.

Since residents typically select their communities, much remains unclear about the causal influence of environment on activity (Handy and Mokhtarian, 2005; Ogilvie et al., 2006). Longitudinal panel studies of relocated families and their behavior and outcomes, like the nascent RESIDE project in Perth, Australia, are designed to untangle this issue. Without a clear sense of how urban form and the availability of mass transit can actually produce more exercise, researchers are limited in advocating specific policy interventions. Still, it is worth assessing the potential magnitudes of the influences

of public transit on health, in order to gauge the plausible scope for policy and motivate further research.

The amount of additional physical activity associated with public transportation appears potentially significant. Besser and Dannenberg (2005) report that half of the roughly 3% of adults in the 2001 U.S. National Household Travel Survey (NHTS) who walked to and from public transit spent 19 minutes or more in total walking time, and almost a third exceeded 30 minutes. Wener and Evans (2007) find that the average New York City train commuter walked about 9,500 steps per day, roughly 2,000 or 30% more steps than the average car commuter. Several papers associate form of transit with obesity directly. Frank et al. (2004) report that obesity around Atlanta, as measured by body mass index (BMI), is associated positively with time spent in cars and negatively with mixed land-use and with walking. Gordon-Larsen et al. (2005) reveal that non-overweight young adults in the Add Health survey were more likely to engage in active transportation like walking or bicycling, possibly in addition to taking public transit. Rundle et al. (2007) find BMI to be inversely associated with the density of bus stops, subway stops, and population around New York City.

But is the *additional* walking associated with mass transit large enough to reduce obesity and associated health care costs? If yes, by how much? In this paper, I address this question by modeling daily time spent walking based on characteristics including transit use, and I then translate those differences into extra net energy expenditure and reductions in obesity.

## 2 Methods

### 2.1 Estimating additional walking associated with public transit

The quantity of interest is the additional amount of physical activity associated with taking public transit as opposed to driving. Wener and Evans (2007) measure this directly by asking a sample of car and train commuters around New York City to wear pedometers, and then comparing total steps for each group. No comparable study exists at the national level, but the 2001 National Household Travel Survey (NHTS) contains similar data that is nationally representative.

Part of the 2001 NHTS included a daily travel diary in which household respondents were asked to self-report all trips, their purposes, starting and ending times, and the means of transportation during an assigned travel day. Individuals or their proxies were asked to fill out the travel diaries on their travel day, and then to relay that information during follow-up telephone interviews. The goal was to obtain travel information for each and every member of the household, but when members were unavailable, their trips went unmeasured. Walking trips undertaken for any reason, whether part of daily commutes, chores or errands, recreation or exercise, were part of the universe of daily trips recorded, as were bicycle trips. No other forms of physical activity were directly measured, but trips were also classified by purpose and could include trips to the gym, to exercise, or to play sports. All legs of all trips on the travel day were categorized by means of transportation. I define a public transit user as anyone who reports using it for any reason

during the travel day.

Walking time in the NHTS is a limited measure of total physical activity in at least three respects. First, and most obviously, physical activity other than bicycling and walking goes unmeasured. Second, since the survey covers only one travel day per individual, it cannot measure behavior on other days that may be related. If a transit user walks more on Monday through Friday, she may choose to walk less on Saturday and Sunday because she is worn out. Third, self-reported walking time may not be a good objective measure of walking. As reviewed by Tudor-Locke and Myers (2001), the literature examining objective and subjective measures of walking typically reveals that individuals under-report total walking. To address the first two limitations, I explore the available information on other exercise in the NHTS, and I also examine how excluding weekend trips affects the results. The third limitation will bias my estimate of additional walking if under-reporting is correlated with public transit use. This seems unlikely but is untestable. I can only compare my estimates to those of Wener and Evans (2007), who collect objective measures of extra walking using pedometers.

I construct total reported daily walking in minutes for each individual in the NHTS, and then I estimate the following model for individual  $i$ :

$$\text{walktime}_i = \alpha_i + \beta \text{pubtrans}_i + \vec{B} \cdot \vec{X} + \epsilon_i, \quad (1)$$

where  $\alpha_i$  is a fixed effect based on geography,  $\text{pubtrans}$  is a dummy variable indicating public transit use,  $\vec{X}$  is a vector of socioeconomic and demographic controls, and  $\epsilon_i$  is a white-noise error. I interpret  $\beta$  as the additional walking associated with transit use.

Estimation is complicated by several concerns. Since public transit use is a choice, endogeneity may render estimates of  $\beta$  biased and inconsistent. With neither a controlled experiment nor good instrumental variables, little can be done other than to acknowledge this problem and work toward improving future study design. A more tractable problem is the fact that walktime exhibits severe response pooling, with 85 percent of NHTS respondents reporting no walking at all. (Table 1 reports characteristics of the weighted sample.) I therefore estimate equation (1) using the Tobit, a standard model for dealing with truncated data.

The version of the dataset I downloaded from the Inter-university Consortium for Political and Social Research contains a total of 140,915 individuals with at least partial travel diaries. Roughly 50,000 completed the entire survey and thus also have a sample weight. Sample weights were calculated based on the characteristics of all sampled households and Census data, where controls included geographic, socioeconomic, and demographic variables. For comparability to Besser and Dannenberg (2005), I define covariates similarly and restrict my analysis to respondents 18 years old and over. Of the 105,942 individuals in the adult subsample, 39,782 filled out the entire survey and have a sample weight, and 28,771 records contain all covariates.

## **2.2 Estimating changes in obesity based on walking**

I convert my estimate of  $\beta$ , minutes of additional walking, into reductions in obesity prevalence in three steps. First, I translate minutes of walking into kilocalories (kcal) of energy expended using the basal metabolic rates (BMR)

reported by Morabia and Costanza (2004): slow walking expends 3.1 kcal per minute, moderate walking 3.9 kcal/min, and fast walking 4.7 kcal/min. Then I convert additional kcal expended into reductions in stored energy using the efficiency factor of 50% cited by Hill et al. (2003). Finally, I match reductions in stored energy with percentiles of the empirical distribution of excess energy stored reported by Hill et al., who examine recent waves of the National Health and Nutrition Examination Survey (NHANES). The result is a percentage that represents the share of Americans for whom weight gain would be eliminated by the given amount of extra walking. This can also be interpreted as the percentage reduction in the percentage increase in obesity prevalence for the average American.

### **2.3 Forecasting obesity prevalence**

My baseline forecast of obesity prevalence is a simple extrapolation of past trends in U.S. obesity rates since 1960. These statistics are reported by Flegal et al. (2002) and Ogden et al. (2006), who examine data from the 1960–1962 National Health Examination Survey (NHES) and subsequent NHANES waves. They are depicted in the left side of Figure 1, which plots historical and projected adult obesity prevalence. I produce alternative forecasts of obesity prevalence by multiplying the baseline annual increase in obesity by the percentage for whom walking eliminates weight gain, which I estimate as described above.



## 2.4 Estimating medical cost savings and other benefits

Obesity is costly along several dimensions. Chronic illnesses such as diabetes and musculoskeletal disorder associated with obesity (Must et al., 1999) cost additional dollars of medical expenditure. These and other diseases also reduce the quality of life (Cutler and Richardson, 1997), and they can also shorten life (Olshansky et al., 2005).

I forecast total medical cost savings per person by first projecting obesity prevalence for each remaining year of the average U.S. citizen's life using the extrapolation technique and the intervention estimates described above. The Census Bureau estimates the median age in the U.S. to be about 36, and a person of that age can expect to live 46 more years (Bell and Miller, 2005). Next, I obtain estimates of additional medical costs associated with obesity, which are assessed by a number of researchers, including Allison et al. (1999), Sturm (2002), Finkelstein et al. (2003), and Lakdawalla et al. (2005). It is convenient to combine the estimates of Sturm, who examines additional costs under age 65, with those of Lakdawalla et al., who explore costs over 70. I translate each of their estimates into 2007 dollars by assuming an annual rate of real growth in per capita medical spending of 3%, per Lakdawalla et al., plus an additional 2.7% per year in general price inflation, for total annual growth of 5.7% in the nominal amounts. Sturm's estimate becomes \$650 per year for those under 65, while that of Lakdawalla et al. becomes \$46,000 for ages 70 and over. I assume that additional spending between age 65 and 70 is \$650.

In my next steps, I follow the lead of Lakdawalla et al., who explain their assumptions about future cost growth and discounting in greater detail.

First, I project additional real spending in each future year by assuming that real per capita costs grow 3% per year. In each year of the projection, I apply the forecast obesity prevalence rate to the additional per capita spending associated with obesity. Then I compute the present discounted value of future amounts using a rate of time discounting also equal to 3% per year. Finally, I allocate spending between public and private funding sources by assuming all spending at ages under 65 is privately funded, while 92.3% of spending at 65 and over is public.

In addition to increased medical costs, obesity also threatens the quality of health and well-being, most notably later in life, and I measure these costs as well. Lakdawalla et al. and Reynolds et al. (2005) both argue that obese elderly can expect to live roughly the same number of remaining years as the non-obese, but that their quality of life will be eroded through obesity-related disability. Lakdawalla et al. expect obese 70 year olds to enjoy 4 years of disability-free life, or 2.8 fewer than the non-obese. Reynolds et al. estimate the obese will live more like 8 years free of disability, but still about 2 years less than the non-obese. I assume 2.5 fewer years of disability-free life for obese elderly, and I ignore impacts earlier in life, about which less is known. The value of a life year spent in disability, or a quality-adjusted life-year (QALY) weight, is about 80% of a life year in perfect health (Cutler and Richardson, 1997), and I use that factor along with estimates of the value of a life year (Viscusi and Aldy, 2003) to gauge the welfare costs of obesity-related disability.

## 3 Results

### 3.1 Additional walking through transit

Table 1 describes the characteristics of the weighted NHTS sample of adults, where the observations are person-days. Typical respondents are nearly 50 years old, roughly split between men and women, predominantly white, and typically hold only high school degrees. Average household income in the dataset is roughly twice per capita income because the data file frequently includes both adults in a typical household. Eighty percent of respondents own their own home. The average population density among these respondents, at just under 4,000 people per square mile, is relatively high and comparable to levels around San Diego, Sacramento, and Portland, OR. But average rail access is low at only 12.3%, which is also roughly the percentage in the Pacific division, and the average number of household vehicles is 2.2.

Total walking time during the travel day averages just 5.1 minutes, with only 14.6% reporting any walking at all. Time spent bicycling is even less common. Only about 2.7% reported any use of public transit, with most using buses as opposed to trains, although some used both. An almost equally large share, 1.9%, reported walking as their sole means of transit on their travel day, but these were primarily recreational walkers rather than commuters. Although they reported little walking, which as discussed above may be an underestimate of actual walking, 12.3% of respondents reported a trip whose purpose was to facilitate engaging in physical exercise, like driving to the gym or to a softball game.

Table 2 presents the results of estimating equation (1), the model of

daily walking time, using several alternative specifications. The first column presents simple ordinary least squares (OLS) estimates, which do not correct for the response heaping at zero minutes of walking time. The other four columns employ the Tobit estimation technique, which corrects for the 85% change of reporting zero walking. In each set of Tobit results, I report the marginal effects on the observed (truncated) variable rather than the underlying behavioral parameters, which are roughly equal to the reported coefficients divided by the probability of positive walking time, 0.15. The latter would be only appropriate if zero walking times actually represented truncated negatives, which seems implausible.

Results are fairly robust across specifications, with public transit use significant at the 1% level in each and associated with between 8 and 10 additional minutes of walking per day. The Tobit specification produces a slightly smaller point estimate than OLS, and adding Census region fixed effects in column 3 lowers it a little more. Respondents in New England, the mid-Atlantic states, Mountain, and Pacific regions reported the most walking, while residents of states in East and West South Central regions, which are located along the Gulf, walked the least. Specifying state fixed effects yielded similar results.

When I include an indicator variable for train use, shown in the fourth column, a significant difference between train and bus commuters emerges. Users of public transit who do not use trains walk only an additional 6 minutes compared with non-users, while those who do walk another 4.5 minutes more, for a total of 10.5 extra minutes per day.

In the fifth column, I remove the train dummy and re-estimate the fixed

effects model after excluding observations in which no other mode of transportation other than walking was reported. Public transit obviously cannot increase walking among those who report no other form than walking, so it is useful to estimate  $\beta$  without them. As shown, dropping exclusive walkers increases the point estimate somewhat, to 10.4.

In the sixth and final column, I drop weekend observations and find little change relative to column 3. My preferred estimate is the 8.3 additional minutes that appears in the third column.

### **3.2 Other physical activity and transit**

The NHTS provides only two other measures of physical activity, bicycling time and trips taken to the gym, to exercise, or to play sports. Bicycling is extremely rare in the NHTS, as shown in Table 1, and its association with public transit is small and of borderline significance. In a Tobit regression of bicycling time on the same covariates and region fixed effects listed in Table 2, I find transit use to be associated with just 0.2 fewer minutes of bicycling per day, significant at the 10% level.

Transit use is negatively associated with taking trips to go to the gym, to exercise, or to play sports. I ran a probit model on the probability of reporting any such trip on the covariates in Table 2, and I found transit use was associated with a reduction in the probability of trips for exercise of 2.7%, significant at the 5% level (estimates not shown). Since only 12.3% of the sample reported any such trips, per Table 1, this seems relatively large.

### **3.3 Reductions in obesity**

Walking expends 3.1, 3.9, or 4.7 kcal/min depending on whether the walking is slow, moderate, or brisk (Morabia and Costanza, 2004), so the 8.3 minutes of additional walking associated with transit use could represent 25.7, 32.4, or 39.0 additional kcal expended each day. At a 50% efficiency rate, those numbers translate into 12.9, 16.2, and 19.5 fewer kcal stored per day. The distribution of excess energy stored reported by Hill et al. (2003) reveals that these levels of additional expenditure could eliminate weight gain in approximately 43%, 50%, or 60% of the population.

### **3.4 Obesity prevalence scenarios**

An OLS regression line through the historical obesity prevalence data in Figure 1 has a slope equal to roughly 0.5% per year, significant at the 1% level. The baseline forecast, with future obesity prevalence increasing 0.5% each year, is shown at top-right in the figure. With 8.3 additional minutes of slow walking per day, 43% of the 0.5% annual increase is offset, leaving 0.29% per year. Similarly, moderate walking reduces growth to 0.25% while brisk walking leaves 0.2%. These three scenarios are depicted in the right-hand side of Figure 1 beneath the baseline projection.

### **3.5 Medical cost and QALY savings**

Table 3 compares the present value of future medical costs in the baseline scenario to five alternatives: the three I have presented thus far; one based on the Wener and Evans (2007) estimate of 2,000 extra steps; and for compari-

son, a scenario with 100 additional kcal/day expended, per the suggestion of Hill et al. (2003). At a rate of 0.04 kcal/step (Hill et al., 2003; Tudor-Locke and Bassett, Jr., 2004), 2,000 steps translates into 80 kcal, or 20.5 minutes of moderate walking.

At baseline, I project that obesity will generate an extra \$34,200 in health costs per person, \$24,400 of which will be borne by Medicare and other public sources. An additional 8.3 minutes of daily walking could save \$4,800, \$5,500, or \$6,600 in present value per person depending on the intensity, and about 80% of the savings is public money. Walking the additional 2,000 steps estimated by Wener and Evans could save \$9,500 per person, and expending an extra 100 kcal/day, as suggested by Hill et al. (2003) could save \$9,900. Reducing obesity below current levels would save even more.

Table 4 presents estimates of QALYs lost and their value in each of the future scenarios, assuming obesity reduces disability-free life by 2.5 years. At baseline, I forecast a 54% chance of obesity over age 70 for the average 36 year old, which is associated with a loss of 0.27 QALY. If a life year spent in perfect health is worth \$200,000 (Viscusi and Aldy, 2003) and grows over time at roughly the real discount rate (Costa and Kahn, 2004), that represents a cost of \$54,000, or about 160% of the additional medical cost shown in Table 3. Walking 8.3 extra minutes could reduce obesity prevalence in old age to between 42 and 45%, representing a savings around \$10,000 or almost double the medical cost savings shown for that scenario in Table 3.

## 4 Discussion

The objective of this paper was to explore the potential benefits of shifting an average U.S. citizen from driving to using public transit. Savings in avoided medical expenses through increased physical activity and decreased obesity appear to be relatively large, around a present value of \$5,500 per person. By comparison, a recent estimate suggests that quitting smoking could save 7,600 and 12,200 € in present value per person depending on sex (Rasmussen et al., 2005), and a comparable dollar figure should be similar. Results hinge on an unknown that is difficult to estimate with great confidence: the additional physical activity associated with public transit. My best guess is that transit use involves 8.3 additional minutes of daily walking, but at least three caveats apply.

First, I cannot correct for the endogeneity of transit choice in cross sectional data without good instrumental variables, and the true causal relationship between transit use and walking could be very different. It is unclear which way this may bias results; transit users could be people who enjoy living in urban areas and walking, or they could also be essentially forced to use transit and might offset the required energy expenditures by exercising less. A related issue that I have avoided addressing in this paper is the complicated relationship between obesity and public transit use. We want to consider public transit as an intervention to combat obesity, but this could get complicated if obesity affects the choice of transportation. Are people obese because they do not choose public transit, or do they not choose public transit because they are obese and it is difficult? It seems unlikely that anyone rationally chooses to be obese (Cutler et al., 2003), so anti-obesity



policies in general are unlikely to trigger rational behavior that directly counteracts the intent, as might be an issue with anti-smoking policies if addiction is rational. But obesity is a state out of which it is costly to climb, or in this case walk, so the design of interventions should acknowledge the costs and probably seek to reduce them. These issues are related to the question of neighborhood selection in environmental studies, and much work remains to be done untangling them.

Second, and somewhat related, I have poor measures of physical activity other than walking, but the limited measures I have suggest some offsetting behavior. Bicycling is so rare that its tiny negative association with transit use is of limited importance. But what are we to make of less frequent trips to the gym among transit users? McDonald and Moffitt (1980) suggest a decomposition of the Tobit coefficient that is worth exploring. The 8.3 additional minutes of walking associated with taking public transit can be decomposed into portions attributable to the associated change in minutes of walking given one is already walking, which is about 8.8, times the share of the sample already walking, 0.15, plus the expected amount of walking if one walks at all, about 33.2 minutes in this sample, times the change in the probability of walking associated with public transit use, which a probit regression reveals to be about 0.21. The larger of these two additive pieces is clearly the latter, which is to say that getting people to walk at all appears in these data to be a key contribution of public transit. So while there may be a cost in the probability of going to the gym, on the order of 2.7%, the 21% increase in the probability of walking seems large by comparison. Further, even if there is offsetting behavior, it could be rational and not a cause

for concern. People who use transit could be getting the optimal amount of exercise without trips to the gym, while drivers may simply not be reporting their exercise. While the low reporting of trips to exercise in the NHTS casts some doubt on this hypothesis, the data are not rich enough to tell.

Third, I cannot measure net energy expenditure, which is a function of caloric intake in addition to physical activity. It could be that public transit indeed causes more physical activity, but individuals offset the potential health effects by eating more. I have no way of testing this hypothesis in the data we currently have.

A more complete look at changes in total physical activity and other behavior associated with transit awaits further study. But it is noteworthy that Wener and Evans (2007) reveal a considerably larger effect than mine, some 2,000 additional steps or 20.5 minutes. They examine entire weeks of readings on pedometer wearers who were instructed to remove them only at bedtime. Presumably only those activities not involving footsteps went unmeasured. My estimate of 8.3 minutes is also considerably less than the 23–24 minutes of transit-related walking reported by Besser and Dannenberg (2005). They examine the self-reports of 3,312 NHTS respondents who walked the entire way to and from public transit, meaning they do not measure individuals who use mixed modes in getting to or from transit, who probably walk less and whom I measure as transit users. Besser and Dannenberg also do not measure other walking during the day, which could be lower among transit users. Regarding the difference between my estimate and the results of Wener and Evans, mine could understate the true impact of transit use on walking if walking were significantly under-reported by transit users in the

NHTS. It could also be the case that the sampled New York train commuters walked more than train commuters elsewhere, perhaps because 79% of them were college graduates.

A somewhat less important component of my analysis is the underlying forecast of obesity prevalence. While estimates of savings relative to baseline depend critically on the intervention, they are more or less independent of the choice of the baseline. I have chosen a simple linear extrapolation of the historical trend, which has been a 0.5% linear rate of increase since the inception of the NHES/NHANES survey in 1960. This is half the 1% rate suggested by Hill et al. (2003), who base it on statistics from the 1988–1994 and 1999–2000 waves of the NHANES that are rounded to the nearest whole percent. Since 1976, a linear trend is closer to 0.7%, but that ignores two previous data points. To be sure, extrapolating past linear trends forward is a risky enterprise, especially when the variable in question is a prevalence rate that must lie between zero and one and could be even more limited. But there is currently no evidence of an upper or lower bound on obesity prevalence, which varies considerably over time and space. According to the Global InfoBase of the World Health Organization, obesity prevalence is approaching 80% in parts of Polynesia. My baseline forecast results in a prevalence of about 55% by 2050, which seems plausible given international trends.

Reduced obesity through walking is only one of a number of potential gains from increased use of public transit. Air pollution is a health hazard that is reduced when public transit usage increases. Users of public transit face much lower rates of injury and death, and increased use should reduce

the hazards faced by motorists and pedestrians. Since public transit is more affordable, its availability can improve access to care and outcomes for vulnerable groups such as low-income pregnant women (Evans and Lien, 2005). Substituting public transit for car commuting probably reduces stress (Wener et al., 2006). These separate pathways through which transit may influence health warrant further investigation, but I leave a broader treatment to future efforts.

I also leave for future efforts several questions that are very tricky to answer. We do not know how to entice an individual to switch from car commuting to public transit, how to extend public transportation into underserved areas, and how high the marginal costs of these activities are relative to these potential marginal benefits I have outlined. Non-users of transit presumably face costs that are high enough to dissuade use. If required to use public transit, or even if prompted to do so through higher taxes and user costs, individuals may face substantial welfare costs that could easily exceed the benefits I have estimated. Dardis and Keane (1995) have assessed welfare costs of quitting smoking among rationally addicted smokers; we need a similar analysis of transportation choices.

## 5 Conclusion

Use of public transit is associated with more walking, by about 8.3 extra minutes per day. This is not enough walking to halt the spread of obesity, but it could substantially reduce it. The present value of medical expenditure savings per person could be \$5,500, while the value of reduced disability could

be even greater.

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**Table 1: Sample characteristics of adults in the 2001 National Household Travel Survey**

<b>Variable</b>	<b>Average</b>
Age	48.2
Male	47.7%
White	83.4%
African American	5.7%
Other race	10.8%
Hispanic	6.3%
Less than high school degree	8.0%
High school graduate	55.9%
College graduate	27.2%
Graduate or professional degree	8.8%
Household income	\$65,473
Homeownership	80.2%
Census division of residence:	
New England	5.5%
Middle Atlantic	13.0%
East North Central	18.4%
West North Central	9.1%
South Atlantic	18.1%
East South Central	5.8%
West South Central	8.7%
Mountain	7.5%
Pacific	14.0%
Population per square mile, block group	3,933
Lives in an MSA with rail	12.3%
Number of household vehicles	2.2
Total daily walking time, minutes	5.1
Total daily biking time, minutes	0.4
Any walking	14.6%
Any biking	1.0%
Used public transit	2.7%
Used bus	2.0%
Used rail	0.9%
Only reported walking	1.9%
Only reported walking, to work	0.4%
Reported trips to gym/exercise/play sports	12.3%
Sample size	28,771

**Notes:** The sample is adults age 18 and over, and sample weights are used to adjust for representativeness and incomplete survey response. A user of public transit is defined as anyone who reported using public transit at any time during his or her travel day. The "Other" racial group includes those not listed and all those self-identifying as mixed-race. "Hispanic" derives from a separate ethnicity question. "College degree" includes the bachelor but not the associate. "Advanced degree" includes any graduate or professional degree. Household income is measured over the previous 12 months.

**Table 2: Marginal effects of characteristics on total daily walking time in the 2001 National Household Transport Survey**

	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	<b>6</b>
	<b>OLS</b>	<b>Tobit</b>	<b>Tobit</b>	<b>Tobit</b>	<b>Tobit</b>	<b>Tobit</b>
Use public transit	9.50 *** (1.21)	8.74 *** (0.89)	8.26 *** (0.87)	5.86 *** (0.85)	10.36 *** (0.94)	8.23 *** (0.92)
Use train	--	--	--	4.50 *** (1.31)	--	--
Age	0.02 *** (0.01)	0.02 *** (0.01)	0.02 *** (0.01)	0.02 *** (0.01)	0.02 *** (0.00)	0.03 *** (0.01)
Male	-0.47 ** (0.21)	-0.52 *** (0.17)	-0.51 *** (0.17)	-0.53 *** (0.17)	-0.44 *** (0.15)	-0.54 *** (0.19)
African American	-0.85 (0.49)	-0.83 (0.31)	-0.57 * (0.32)	-0.58 * (0.32)	-0.92 *** (0.29)	-0.28 (0.37)
Other race	-0.36 (0.53)	-0.53 (0.38)	-0.52 (0.38)	-0.52 (0.38)	-0.45 (0.36)	-0.42 (0.43)
Hispanic	0.43 (0.69)	0.43 (0.54)	0.45 (0.55)	0.46 (0.55)	0.20 (0.50)	0.71 (0.63)
Less than high school degree	-1.12 *** (0.36)	-0.81 *** (0.29)	-0.75 *** (0.29)	-0.73 ** (0.29)	-0.97 *** (0.27)	-0.72 ** (0.33)
College graduate	1.00 *** (0.29)	1.09 *** (0.24)	1.11 *** (0.24)	1.11 *** (0.24)	1.04 *** (0.22)	0.80 *** (0.26)
Graduate or professional degree	1.59 *** (0.48)	1.40 *** (0.37)	1.38 *** (0.37)	1.36 *** (0.37)	1.20 *** (0.34)	1.47 *** (0.42)
Log of Household Income	-0.17 (0.15)	-0.09 (0.11)	-0.14 (0.11)	-0.16 (0.11)	0.09 (0.10)	-0.07 (0.13)
Log population per square mile, block group	0.42 *** (0.06)	0.50 *** (0.05)	0.42 *** (0.05)	0.42 *** (0.05)	0.34 *** (0.05)	0.43 *** (0.06)
Own home	-0.62 ** (0.30)	-0.81 *** (0.24)	-0.72 *** (0.24)	-0.72 *** (0.24)	-0.15 (0.22)	-0.72 *** (0.27)
Number of household vehicles	-0.78 *** (0.10)	-0.86 *** (0.10)	-0.86 *** (0.10)	-0.86 *** (0.10)	-0.57 *** (0.09)	-0.83 *** (0.11)
N	28,771	28,771	28,771	28,771	28,217	21,326
Census region fixed effects?	no	no	yes	yes	yes	yes
Include those who only reported walking?	yes	yes	yes	yes	no	yes
Include weekends?	yes	yes	yes	yes	yes	no

**Notes:** See notes to Table 1. Each column reports marginal effects from estimation of a model of daily walking time as shown in equation (1) in the text, with standard errors in parentheses. Column 1 reports ordinary least squares estimates, while columns 2-6 report Tobit estimates under various alternative specifications, where the reported marginal effects of each Tobit are the partial derivatives of the expected value of the observed (truncated) walking time variable. Asterisks denote statistical significance, with three at the 1% level, two at 5%, and one at the 10% level.

**Table 3: The effect of additional walking on the present value of additional health care spending associated with obesity, 2007 dollars**

	Additional energy expenditure per day	Annual growth in obesity prevalence	Present value of additional health spending per person	Present value of savings relative to baseline	Present value of additional public health spending per person	Present value of public savings relative to baseline
Baseline	0 kcal	0.5%	\$34,200	--	\$24,400	--
8.3 additional minutes of walking: slowly (3.1 kcal/min)	25.7 kcal	0.29%	\$29,400	\$4,800	\$20,600	\$3,800
moderately (3.9 kcal/min)	32.4 kcal	0.25%	\$28,700	\$5,500	\$19,900	\$4,500
briskly (4.7 kcal/min)	39.0 kcal	0.20%	\$27,600	\$6,600	\$19,000	\$5,400
2,000 additional steps of walking, or 20.5 minutes at 3.9 kcal/min	80 kcal	0.07%	\$24,700	\$9,500	\$16,700	\$7,700
100 kcal per day	100 kcal	0.05%	\$24,300	\$9,900	\$16,400	\$8,000

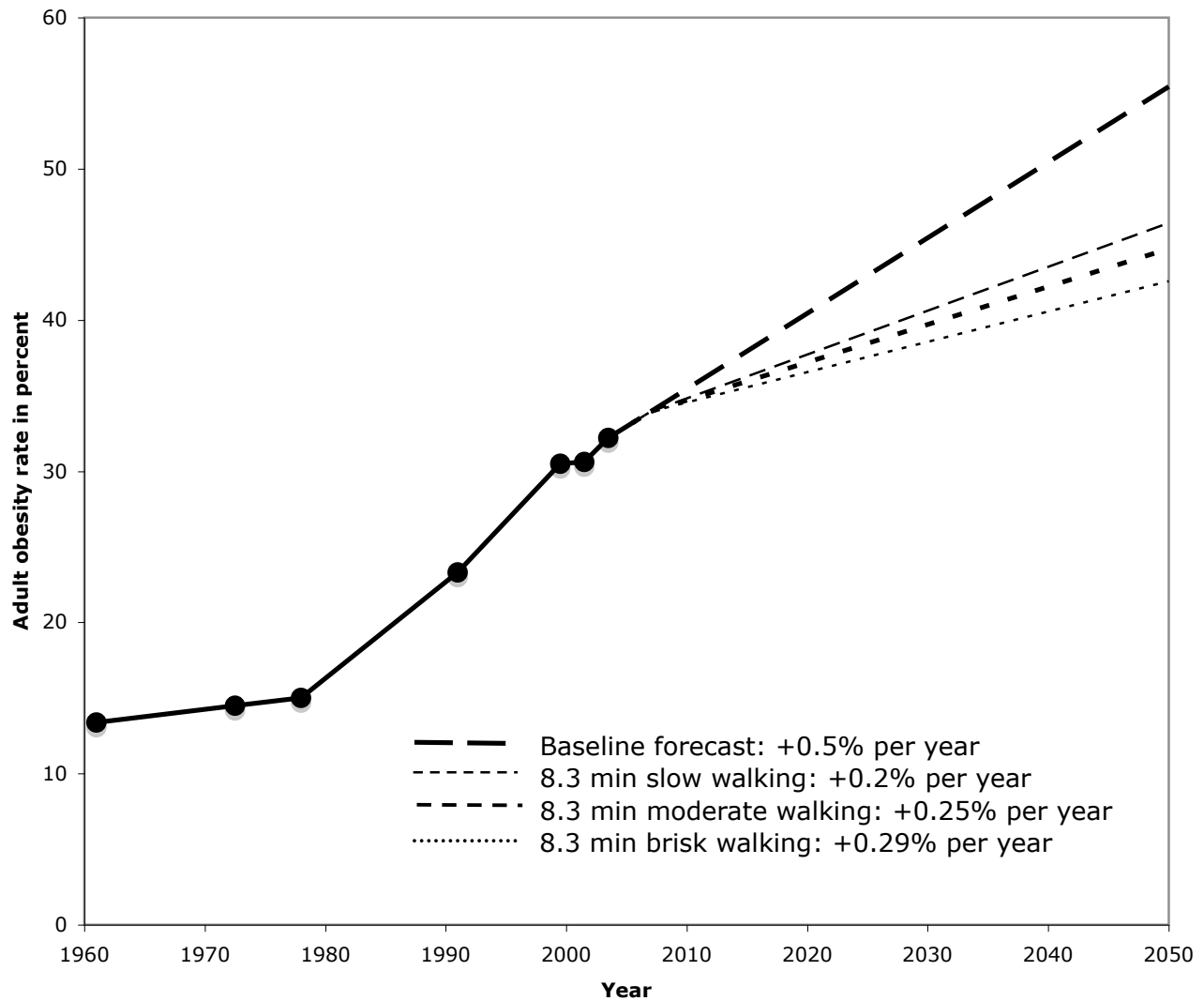
**Notes:** Savings are presented per person, not per obese person. Forecasts assume a 36 year old in 2007 who lives another 46 years. The obesity prevalence rate is assumed to be 34% in 2007. From age 36 to 69, obese individuals incur \$650 extra each year in medical expenditures, inferred from Sturm (2002) per the text. Individuals 70 and over incur \$46,000 extra in present value, from Lakdawalla et al. (2005) as described in the text, 92.3% of which is paid by Medicare. That share of spending between age 65 and 70 is also categorized as public. All real medical spending grows at 3% per year. The real discount rate is also 3% per year.

**Table 4: The effect of additional walking due to public transit on health status and its value by rate of future increase in obesity**

	Average obesity prevalence over age 70	Expected QALYs lost due to obesity	Expected present value of QALYs lost due to obesity	Savings relative to baseline
Baseline	54%	0.27	\$54,000	--
8.3 additional minutes of walking: slowly (3.1 kcal/min)	45%	0.23	\$45,400	\$8,600
moderately (3.9 kcal/min)	44%	0.22	\$44,000	\$10,000
briskly (4.7 kcal/min)	42%	0.21	\$42,000	\$12,000
2,000 additional steps of walking, or 20.5 minutes at 3.9 kcal/min	37%	0.18	\$36,800	\$17,200
100 kcal per day	36%	0.18	\$36,000	\$18,000

**Notes:** See notes to Table 3. The assumed QALY weight of a life-year spent disabled is 0.8. The present value of a life-year in perfect health is assumed to be \$200,000.

**Figure 1. Adult obesity rates in the U.S., historical and projected**



**Notes:** Data are the rate of obesity, defined as body mass index (BMI) 30 or greater, for the adult population measured in the National Health Examination Survey (NHES) and National Health and Nutrition Examination Survey (NHANES). Obesity rates are calculated by Flegal et al. (2002) and Ogden et al. (2006), appear as circles in the figure, and are plotted at the midpoint of the NHANES examination period. The dashed line represents a simple forward extrapolation from the most recent data point of the average annual increase in the rate, 0.5%, which is estimated by ordinary least squares regression of all historical data on time.