



Original article

The short-term impacts of Earned Income Tax Credit disbursement on health

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Abstract

Background: There are conflicting findings regarding long- and short-term effects of income on health. Whereas higher average income is associated with better health, there is evidence that health behaviours worsen in the short-term following income receipt. Prior studies revealing such negative short-term effects of income receipt focus on specific subpopulations and examine a limited set of health outcomes.

Methods: The United States Earned Income Tax Credit (EITC) is an income supplement tied to work, and is the largest poverty reduction programme in the USA. We utilize the fact that EITC recipients typically receive large cash transfers in the months of February, March and April, in order to examine associated changes in health outcomes that can fluctuate on a monthly basis. We examine associations with 30 outcomes in the categories of diet, food security, health behaviours, cardiovascular biomarkers, metabolic biomarkers and infection and immunity among 6925 individuals from the U.S. National Health and Nutrition Survey. Our research design approximates a natural experiment, since whether individuals were sampled during treatment or non-treatment months is independent of social, demographic and health characteristics that do not vary with time.

Results: There are both beneficial and detrimental short-term impacts of income receipt. Although there are detrimental impacts on metabolic factors among women, most other impacts are beneficial, including those for food security, smoking and trying to lose weight.

Conclusions: The short-term impacts of EITC income receipt are not universally health promoting, but on balance there are more health benefits than detriments.

Key words: Income, public policy, biomarkers, health behaviours, socioeconomic, tobacco

Key Messages

- Although higher average income is generally associated with better health, the majority of evidence suggests that health related behaviours worsen in the short term following income receipt.
- We utilize the fact that in-work tax credit recipients in the USA receive large cash transfers in the months of February, March and April, as a natural experiment.
- In the months when cash transfers occurred, about one-third of the health-related outcomes examined differ as compared with those of eligible individuals in control months.
- Eligible individuals in the disbursement months had less food insecurity, less smoking or exposure to smoke and were more likely to be trying to lose weight.
- Compared with prior work that examined specific subpopulations of individuals or only substance use outcomes, our findings show on balance beneficial impacts of income receipt on health-related outcomes in individuals in households with dependent children.

Background

Existing research on income and health reveals an apparent puzzle. Having a higher level of income has long been shown to be associated with better health outcomes and behaviours.^{1,2} In contrast, income receipt, conceptually distinct from income itself, has been shown to increase some unhealthy behaviours and mortality. Focusing on income receipt and using multiple identification strategies, studies have documented higher rates of substance use, drug-related hospitalizations, and mortality in the days and weeks following disbursement of government transfers and wage payments from the military.^{3–7} These findings raise concerns about potential unintended negative short-term health consequences of income transfers. However, existing studies have focused on a small range of health outcomes, emphasizing negative health behaviours (e.g. substance use) while leaving other relevant health outcomes untested. Behavioural and psychological responses to income are complex and it is reasonable that income receipt effects may vary for different health outcomes. In addition, most prior studies have been limited to specific subpopulations, such as individuals in the military or older individuals,⁶ and these results may not generalize to broader populations.

To present a more comprehensive examination of short-term health consequences of income receipt, we analyse data from the U.S. National Health and Nutrition Examination Survey (NHANES) and examine 30 health-related outcomes that fluctuate over short periods of time (i.e. less than 1 month) in the categories of diet, food security, health behaviours, cardiovascular biomarkers, metabolic biomarkers, and infection and immunity. Our analytical strategy takes a natural experiment approach. It utilizes the fact that in the USA, recipients of the Earned

Income Tax Credit (EITC) typically receive lump sum payments in February, March and April, following the processing of federal and state income taxes. Over 80% of individuals who qualify for the EITC actually receive the credit.⁸ Because the NHANES study surveys individuals throughout the year, we can measure a ‘treatment’ group of EITC-eligible individuals who were surveyed in February, March or April and a ‘comparison’ group of EITC-eligible individuals who were surveyed in May through January. The benefit for inference is that the month in which an individual is surveyed is random with respect to individual characteristics. To account for temporal trends in health outcomes unrelated to income receipt, we can further compare EITC-ineligible individuals who were surveyed in February, March or April vs May through January. This analytical strategy, a Difference-in-Difference approach, overcomes many of the common confounding problems that arise in analyses of income receipt and health, since many of the unmeasured individual-level characteristics that may jointly determine income receipt and health are unlikely to have seasonal variations that differ between EITC-eligible individuals and non-EITC-eligible individuals.

The EITC is an in-work tax credit. Since the EITC is the largest anti-poverty programme in the USA,⁸ this analysis also helps illuminate the health consequences of a major piece of US social welfare policy, with potential relevance to other in-work tax refund policies in other industrialized countries.⁹ There is a developing body of work that has examined the effects of the EITC in the USA, and of similar types of in-work tax credits in other countries, on health and development outcomes.^{10–13} Our work builds on this prior work by focusing specifically on the effects of income receipt, rather than on income more generally. This study, focusing on short-term fluctuations in health, captures

health consequences of a time-delimited cash transfer that occurs in particular months.¹⁴

Methods

Data and analytical sample

Data for our analysis were from the U.S. Third National Health and Nutrition Examination Survey (NHANES III), which was designed to be representative of the non-institutionalized population of the USA.¹⁵ These data were collected as a cross-sectional study from 1988–94. Since the EITC is much more generous for individuals with dependents, we restricted our sample to respondents between the ages of 21 and 50 years ($n = 10\,022$) because this age group is most likely to have children in the household who are aged 18 years or under. There were less than 3% of participants missing any specific outcome: we excluded 672 individuals who were missing data on any of the outcomes examined. We also excluded from our analysis respondents living in the Mid-Western and North-Eastern United States because these regions did not have year-round data collection, due to snowfall in winter. The remaining population of individuals from the Western and Southern United States numbered 6925. Of these 6925 individuals, 1689 (24%) were measured in the 'treatment' months of February through April, 2225 (32%) qualified for some EITC and 870 (13%) qualified for at least \$1000 (in year-2000 dollars) of EITC. There were 167 women who qualified for at least \$1000 of EITC in the treatment months, and 392 women who qualified for any EITC in the treatment months. There were 111 men who qualified for at least \$1000 of EITC in the treatment months, and 276 men who qualified for any EITC in the treatment months.

The Earned Income Tax Credit exposure

Estimated eligibility and predicted amounts of federal EITC credit were calculated for each household, based on the following characteristics: (i) earned income; (ii) number of dependent children; and (iii) year. Earned income was self-reported for the household. Number of dependent children was not directly reported, but was calculated based on the total number of related persons living in a household, subtracting two if the respondent was cohabitating/married, and subtracting one if the household was single-headed. We then predicted EITC amounts using the National Bureau of Economic Research's TAXSIM programme,¹⁶ which calculates exact qualifying values of the EITC using federal and state income tax code. This form of exposure results in an intent-to-treat interpretation of

model coefficients. Because moderate-to-large income transfers are more likely to be consequential for health, we defined an EITC-eligible household as one eligible for \$1000 (in year-2000 dollars) or more in EITC benefits. The \$1000 cut-point was chosen as a relatively large credit, but not so high a level that there would only be a small population receiving the credit. As a secondary analysis, we examined the population eligible for any EITC credit. Our hypothesis was that there will be greater magnitudes of association with outcomes for those who are eligible for greater amounts of EITC dollars.

Health outcomes

The choice of outcomes was done a priori, based on the following criteria: (i) factor available in NHANES III; (ii) factor shown in the literature to be associated with health; and (iii) factor that can vary within a 1-month period based on environmental changes. We examined an overall diet measure using the Healthy Eating Index (ranging from 1 to 100) created by the U.S. Centers for Disease Control, as well as specific measures of intake of vegetables, fruit, meat, dairy, sodium, saturated fat and variety of foods (each ranging from 1 to 10).^{17–19} Food insecurity was assessed with two validated questions.²⁰ The first question captured how many days in the past month a family had no food or was unable to buy food, and the second assessed whether this was because of insufficient money, food stamps or vouchers. We dichotomized this count as 0 days or at least 1 day. For the behaviours we examined: self-report of current smoking (yes or no); self-report of trying to lose weight (yes or no); cotinine (a continuous biomarker assessing exposure to second-hand smoke and individual smoking); self-reported marijuana smoking (any days in the past month); percent of kCal from alcohol (derived from self-report of alcohol consumption in the past 3 days) and a self-report of the number of days in the past month on which the participant walked a mile (dichotomized as less than 4, or 4 or more).¹⁵ For cardiovascular biomarkers we examined: systolic blood pressure; diastolic blood pressure; and forced expiratory volume.²¹ For metabolic biomarkers we examined: high-density lipoprotein (HDL) cholesterol; low-density lipoprotein (LDL) cholesterol; triglycerides; haemoglobin (Hb) A1c and plasma glucose.²¹ For infection and immunity we examined: C-reactive protein; fibrinogen; lymphocytes; self-report of respiratory illness in the past 4 weeks; and self-report of cold or flu episodes in the past 4 weeks.²¹ To facilitate interpretation and comparison with previous studies focusing on negative health behaviours, all the outcomes are coded so that higher values indicate a negative (i.e. worsening) health indicator/behaviour.

Statistical model

The difference-in-difference approach was specified as:

$$Y_i = \alpha + \beta_1 \text{ Treatment Months} + \beta_2 \text{ EITC qualified} + \beta_3 \text{ Treatment Months} \times \text{EITC qualified} + \beta_4 \text{ age} + \beta_5 \text{ wave of data} + \varepsilon_i$$

Treatment months indicates that the respondents' measurements were taken in February–April. EITC qualified indicates qualification for $\geq \$1000$ of EITC. β_3 reflects differences in health between the treatment and the comparison group. All analyses are done stratified by gender, due to literature suggesting different health-related

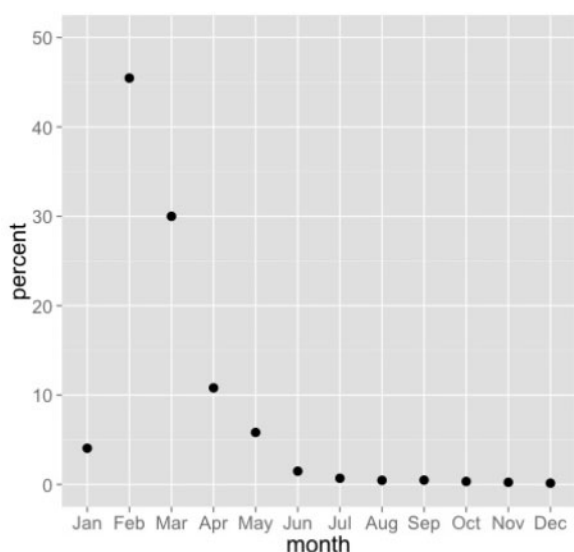


Figure 1. Distribution of Earned Income Tax Credit Disbursement by month. Authors' calculation from Monthly U.S. Treasury Receipt Data.

associations with income by gender^{22,23} and differences in household structure of male and female EITC recipients.²⁴ To correct for the sample design of the NHANES,²⁵ all the results are weighted and adjusted for clustering using the survey package in R.²⁶ When presenting results, continuous outcomes are expressed in terms of Z-scores to facilitate comparisons across coefficients. Using the survey package in R to implement a generalized linear approach, we modelled dichotomous outcomes with a binomial model with the logit link function, and continuous outcomes with a Gaussian error distribution with the identity link function.

Prior work suggests that standard asymptotic assumptions of error distribution for difference-in-difference models with small numbers of treatment groups may underestimate the true variance.²⁷ To address this, we perform additional robustness checks by creating nine placebo treatment groups based on all possible adjacent 3-month combinations between May and January (i.e. non-treatment months), and fit models with each of these 3-month groups as a placebo treatment. We will have increased confidence in our primary analysis if the coefficients from each of the nine placebo treatment groups are closer to the null than the coefficient from our actual treatment group.

Results

Figure 1 presents data on how EITC is distributed by month in the US population, showing February, March and April as the months when most households receive their credit.

Table 1 presents the racial/ethnic and educational distribution of the population who did not qualify for a minimum of \$1000 in EITC as compared with those who did in the treatment months (February–April) and the control

Table 1. Percentages (and standard deviation) of not eligible for \$1000 or more EITC vs qualifiers by treatment and control month, stratified by gender, NHANES III

	Women		Men	
	Not eligible (n = 3012)	EITC eligible Feb–Apr (n = 167)	Not eligible (n = 2967)	EITC eligible Feb–Apr (n = 111)
Race				
White	72 (0.44)	42 (0.46)	71 (0.45)	42 (0.42)
Black	12 (0.32)	28 (0.45)	12 (0.32)	26 (0.38)
Mexican American	7.0 (0.26)	13 (0.48)	8.5 (0.28)	21 (0.50)
Other	9.2 (0.29)	17 (0.23)	8.6 (0.28)	12 (0.48)
Education				
<High school	16 (0.36)	41 (0.50)	19 (0.40)	52 (0.48)
High school diploma	36 (0.48)	41 (0.46)	32 (0.47)	31 (0.40)
Some college	48 (0.50)	17 (0.37)	49 (0.50)	17 (0.36)

months (May–January), by gender. Among both men and women, non-Whites and respondents with lower education were overrepresented among those qualifying for at least \$1000 in EITC. This differential distribution motivates the need for quasi-experimental approaches to examine impacts of EITC disbursement. The identical distribution of characteristics between those surveyed in the treatment as compared with the control months suggests random distributions of these characteristics throughout the year.

Supplementary Figure 1 (available as Supplementary data at *IJE* online) shows the distribution of annual income for individuals qualifying for at least \$1000 of EITC in the middle, those qualifying for less than \$1000 but more than zero on the left and those who do not qualify for any EITC on the right. Because the EITC is targeted at low-wage workers, the income distribution for qualifiers is much more restricted than for non-qualifiers. Table 2 presents means (continuous measures), percentages (dichotomized measures) and standard deviations for the 30 health

Table 2. Means (or proportions) and standard deviation of risk factors, by gender, NHANES III ($n = 6925$)

	Women						Men					
	Not eligible		EITC eligible				Not eligible		EITC eligible			
	$(n = 3012)$		Feb–Apr $(n = 167)$		May–Jan $(n = 392)$		$(n = 2967)$		Feb–Apr $(n = 111)$		May–Jan $(n = 276)$	
Diet												
Overall diet	62.7	12.2	59.5	10.5	60.7	11.9	60.9	12.0	59.7	11.9	58.6	11.1
Vegetables	5.98	3.45	5.21	3.55	5.20	3.44	5.97	3.38	5.71	3.37	5.39	3.57
Fruit	3.35	3.78	2.35	3.45	2.70	3.52	2.64	3.37	3.17	3.74	2.00	3.27
Meat	6.76	3.09	7.24	2.79	6.78	3.04	7.97	2.73	8.26	2.73	8.23	2.69
Dairy	6.13	3.67	5.66	3.74	5.52	3.84	6.76	3.61	6.47	3.83	5.62	4.02
Sodium	6.62	3.68	6.58	3.79	7.05	3.64	4.01	4.03	4.39	4.19	4.00	4.11
Saturated fat	6.19	3.96	5.92	4.12	6.44	3.92	6.31	3.87	6.49	3.82	6.82	3.72
Diet variety	7.54	3.29	6.50	3.60	6.67	3.59	7.98	2.93	7.53	3.23	7.26	3.35
Food security												
Not enough food ^a	0.01	0.11	0.15	0.36	0.09	0.29	0.02	0.13	0.07	0.26	0.10	0.29
No money for food ^a	0.01	0.11	0.16	0.37	0.10	0.30	0.02	0.13	0.08	0.27	0.10	0.31
Health behaviours												
Smoking ^a	0.25	0.43	0.38	0.49	0.33	0.47	0.37	0.48	0.40	0.40	0.55	0.50
Not try lose weight ^a	0.57	0.50	0.66	0.47	0.63	0.48	0.76	0.43	0.70	0.46	0.80	0.40
Cotinine	54.9	113	69.3	121	90.9	153	105	165	90.7	117	153	166
Marijuana ^a	0.06	0.23	0.06	0.24	0.09	0.28	0.14	0.35	0.10	0.30	0.21	0.41
Alcohol	2.24	6.04	1.11	4.05	1.58	5.99	3.73	7.29	3.09	7.70	4.26	10.2
Not walk mile/week ^a	0.69	0.46	0.78	0.42	0.73	0.45	0.68	0.47	0.68	0.47	0.62	0.49
Cardiovascular												
Systolic pressure	71.9	9.52	69.6	10.6	71.4	9.69	77.1	9.49	75.9	10.9	75.8	10.1
Diastolic pressure	113	12.7	110	14.4	111	13.1	120	11.2	120	12.4	120	11.8
Forced Expiratory volume	3653	623	3554	577	3458	613	5096	822	4752	860	4852	789
Pulse rate	76.3	11.8	78.0	11.8	76.9	11.2	72.8	11.3	76.7	13.0	73.2	12.2
Metabolic												
HDL cholesterol	1.43	0.39	1.39	0.34	1.31	0.38	1.18	0.35	1.20	0.36	1.18	0.36
LDL cholesterol	3.00	0.83	3.09	0.87	3.02	1.02	3.36	0.90	3.14	0.95	3.15	0.98
Triglycerides	5.00	0.93	5.12	1.20	4.96	1.14	5.13	1.04	5.21	1.14	4.97	1.07
Haemoglobin A1c	5.11	0.67	5.34	0.99	5.26	1.02	5.29	0.70	5.50	0.84	5.32	0.67
Glucose	91.2	19.2	95.5	31.7	92.9	31.5	97.0	23.1	98.8	31.3	94.5	16.6
Infection & immunity												
C-reactive protein	0.42	0.53	0.48	0.56	0.50	0.63	0.31	0.40	0.38	0.47	0.37	0.38
Lymphocytes	31.9	8.14	32.2	7.77	32.8	8.14	32.5	7.74	32.8	9.55	34.1	8.79
Illness ^a	0.22	0.42	0.23	0.42	0.23	0.42	0.16	0.37	0.26	0.44	0.26	0.44
Respiratory infection ^a	0.08	0.28	0.10	0.30	0.07	0.26	0.03	0.18	0.11	0.31	0.06	0.24
Cold ^a	0.35	0.48	0.35	0.48	0.36	0.48	0.29	0.45	0.34	0.47	0.33	0.47

^aVariables that are dichotomous. Means (or percentages for dichotomous variables) and standard deviations are calculated using study sampling weights.

outcomes examined among those who did not qualify for a minimum of \$1000 in EITC as compared with those who did, in the treatment months (February–April) and the control months (May–January), by gender.

As a specification test, Table 3 presents odds ratios, coefficients and 95% confidence intervals for the treatment months and EITC qualification interaction term (i.e. β_3 from our model) for demographic and anthropometric factors that could not vary over time. The purpose of these models was to test whether there were any significant differences in the sample by month that could confound our estimates. All 95% confidence intervals in this table included the null, which is consistent with our prior belief that there would be no meaningful demographic differences across EITC-eligible individuals who were surveyed in February through April compared with other months.

Table 4 presents primary findings for our 30 health outcomes, stratified by gender. As in Table 3, these results reflect the treatment months and EITC qualification interaction term (i.e. β_3 from our model). Continuous outcomes are Z-scored, thus coefficients can be interpreted in terms of the difference in standard deviations of the outcome associated with EITC treatment. Associations with dichotomous outcomes are expressed as odds ratios. Because all the outcomes are coded with higher values indicating worse health, values greater than the null (0 for continuous measures, 1 for dichotomous measures) indicate that EITC treatment is associated with worse health-related outcomes. Among the 60 relationships examined, 21 had 95% confidence intervals that did not include the null, substantially more than the three that would be expected by chance. Of the 13 factors where 95%

confidence intervals did not include the null for women, nine were in a direction associated with better health (meat, not enough food, not enough money for food, smoking, not trying to lose weight, marijuana use, HDL cholesterol, lymphocytes and colds), and 4 were in the direction associated with worse health (sodium, pulse rate, LDL cholesterol and triglycerides). Of the eight factors where 95% confidence intervals did not include the null for men, seven were in a direction associated with better health (fruit, dairy, not enough food, not enough money for food, not trying to lose weight, cotinine and marijuana) and one was not (saturated fat). In general, effect sizes were small, with impacts on smoking, trying to lose weight and food insecurity around 2–3%, and differences in metabolic factors and diet around one-third of a standard deviation.

As a sensitivity analysis, Supplementary Table 1 (available as Supplementary data at *IJE* online) presents a similar analysis but for individuals qualifying for an EITC of any size. Among the 60 relationships examined, 13 had confidence intervals that did not cross the null. Among women, all six confidence intervals were associated with better health (meat consumption, smoking, not trying to lose weight, marijuana use, diastolic blood pressure and colds). Among men, six were associated with better health (overall diet, fruit, sodium, not trying to lose weight, cotinine and marijuana) and one was not (Haemoglobin A1c).

Figures 2 and 3 show results of the permutation tests of placebo treatments for women and men, respectively. The light-coloured dot at 2 on the x-axis shows the coefficient for the true treatment months shown in Table 4, and the black dots indicate the placebo treatment permutation

Table 3. Impact of EITC disbursement month and qualifying for at least \$1000 in EITC on negative control dependent outcomes from difference-in-difference models, by gender, NHANES III ($n = 6925$)

Dependent variable	Women ($n = 3571$)			Men ($n = 3354$)		
	Coef/OR	Lower 95% CI	Upper 95% CI	Coef/OR	Lower 95% CI	Upper 95% CI
White	0.927	0.504	1.702	1.434	0.607	3.385
Black	0.972	0.615	1.536	0.919	0.453	1.864
Mexican American	1.081	0.762	1.535	1.290	0.797	2.088
<High school	1.678	0.877	3.209	0.962	0.485	1.907
High school diploma	0.649	0.369	1.140	1.326	0.620	2.836
Some college	1.013	0.621	1.653	0.742	0.361	1.521
Annual income	1692	–1786	5170	828	–1727	3383
Height	0.019	–0.006	0.043	0.034	–0.007	0.075
Currently working	1.001	0.999	1.002	1.002	1.000	1.003

Models include age, treatment month, EITC qualification, and treatment month*EITC qualification interaction term. Only the coefficient for interaction term is shown in the table, as an odds ratio for categorical variables and as a coefficient for continuous variables (annual income and height). Annual income is in year 2000 American dollars. Height is in centimetres.

Coef, coefficient; OR, odds ratio.

Table 4. Impact of EITC disbursement month and qualifying for at least \$1000 in EITC in difference-in-difference models, by gender, NHANES III ($n = 6925$)

	Women ($n = 3571$)			Men ($n = 3354$)		
	Coef/OR	Lower 95% CI	Upper 95% CI	Coef/OR	Lower 95% CI	Upper 95% CI
Diet						
Overall diet	0.150	-0.056	0.355	0.004	-0.154	0.162
Vegetables	0.071	-0.141	0.284	-0.098	-0.465	0.269
Fruit	0.066	-0.103	0.235	-0.396	-0.631	-0.162
Meat	-0.382	-0.615	-0.150	-0.043	-0.263	0.177
Dairy	0.114	-0.140	0.369	-0.326	-0.633	-0.019
Sodium	0.330	0.120	0.540	-0.262	-0.632	0.108
Saturated fat	-0.020	-0.342	0.303	0.393	0.116	0.670
Diet variety	0.149	-0.207	0.506	-0.107	-0.379	0.166
Food security						
Not enough food ^a	0.968	0.952	0.985	0.973	0.954	0.991
No money for food ^a	0.970	0.954	0.987	0.972	0.954	0.991
Health behaviours						
Smoking ^a	0.982	0.974	0.989	1.003	0.995	1.011
Not try lose weight ^a	0.977	0.964	0.990	0.974	0.964	0.984
Cotinine	-0.045	-0.318	0.228	-0.314	-0.531	-0.097
Marijuana ^a	0.997	0.995	0.998	0.994	0.992	0.996
Alcohol	-0.004	-0.213	0.206	0.009	-0.417	0.435
Not walk mile/week ^a	0.998	0.996	1.001	1.002	0.999	1.006
Cardiovascular						
Systolic blood pressure	-0.078	-0.232	0.076	0.207	-0.033	0.447
Diastolic blood pressure	-0.125	-0.273	0.023	0.171	-0.004	0.347
Forced expiratory volume	-0.082	-0.247	0.083	0.045	-0.254	0.343
Pulse rate	0.343	0.074	0.611	0.174	-0.070	0.418
Metabolic						
HDL cholesterol	-0.386	-0.681	-0.092	0.120	-0.153	0.394
LDL cholesterol	0.238	0.038	0.437	0.152	-0.275	0.578
Triglycerides	0.218	0.020	0.417	-0.053	-0.300	0.194
Haemoglobin A1c	0.186	-0.005	0.377	0.069	-0.106	0.244
Glucose	0.089	-0.046	0.225	0.095	-0.044	0.234
Infection & immunity						
C-reactive protein	0.026	-0.087	0.138	0.003	-0.174	0.180
Lymphocytes	-0.342	-0.506	-0.178	-0.135	-0.551	0.281
Illness ^a	0.999	0.997	1.002	1.000	0.998	1.001
Respiratory infection ^a	1.001	1.000	1.002	1.000	0.999	1.001
Cold ^a	0.990	0.983	0.997	0.995	0.990	1.001

Coefficients are from interaction term of disbursement month X EITC qualified for at least \$1000. All continuous outcomes are Z-scored. Positive coefficients indicate association with worse health. Odds ratios above 1 indicate association with worse health. Coefficients and 95% confidence intervals are emboldened when the 95% confidence interval does not include the level of the coefficient indicating no association.

^aCoefficients for dichotomous outcomes are presented as odds ratios.

Coef, coefficient; OR, odds ratio.

analysis for the nine other groups, with starting month as indicated on the x-axis (i.e. 4 indicates April, May, June; 5 indicates May, June, July). When the coefficient for the true treatment group is further from the null than all nine of the placebo treatments, this indicates support for a true difference during the treatment months. This was for most factors consistent with the findings presented in Table 4 with the exception of marijuana use for men and

women, LDL cholesterol for women and fruit consumption for men.

Discussion

The majority of outcomes we examined were not affected by income receipt among EITC recipients. Among factors that were affected, the short-term effects of EITC receipt

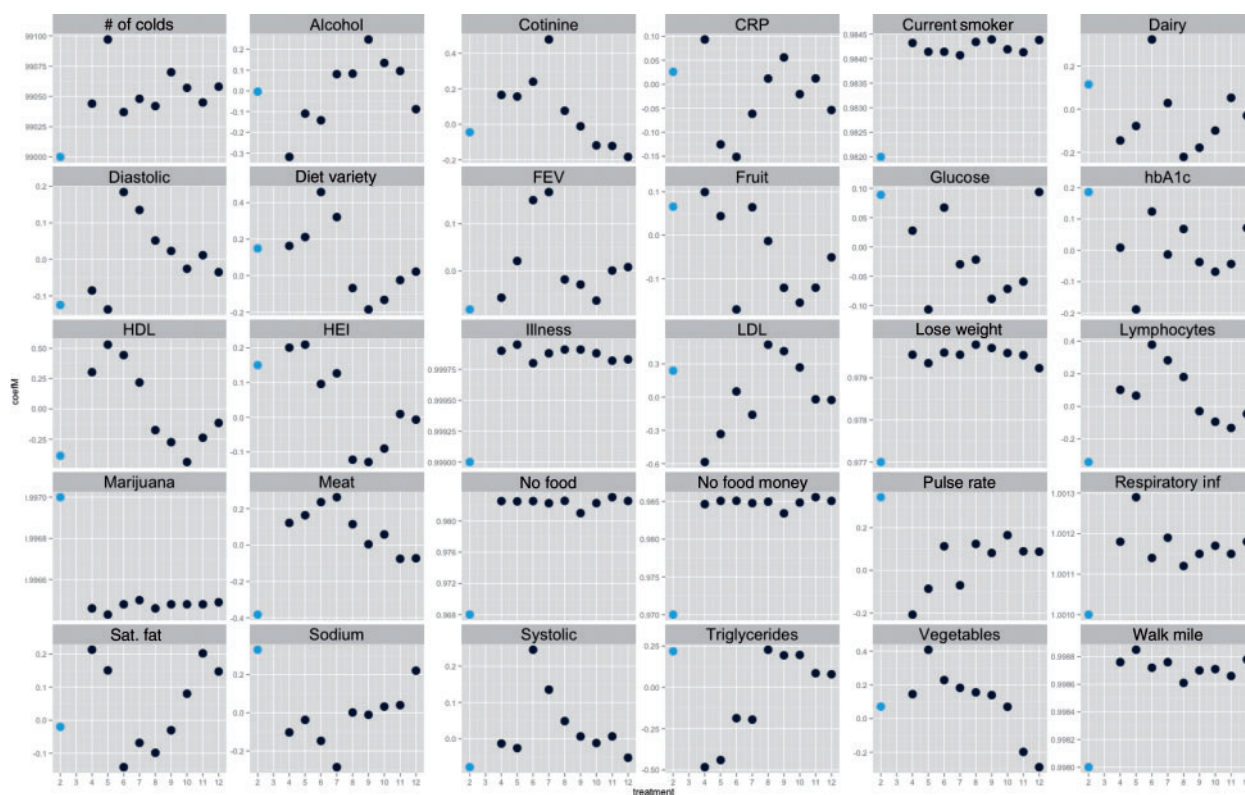


Figure 2. Permutation tests of placebo treatments for difference-in-difference models of EITC receipt and health-related outcomes, women. Light dots indicate point estimate of actual exposure. Black dots indicate point estimates of placebo permutation exposures. Y-axis for each plot differs depending on scale of outcome. X-axis indicates first month of 3-month group of exposure (i.e. 2 indicates true treatment months of Feb, Mar, Apr; 4 indicates placebo treatment months of Apr, May, Jun). FEV, Forced Expiratory Volume; HEI, Healthy Eating Index Overall Diet.

were on balance more health-promoting than detrimental. Indeed, many outcomes that are key determinants of health (e.g. food security, smoking/exposure to smoke) were affected in a health-promoting direction. Focusing on results for respondents who were eligible for at least \$1000 of EITC, nine of the 13 factors where the 95% confidence interval did not include the null for women, and seven of the eight factors where the 95% confidence interval did not include the null for men, are in a direction associated with better health. Results related to diet were mixed. On the positive side, EITC income receipt appeared associated with a healthier level of meat consumption for women and a healthier level of dairy consumption for men. On the negative side, EITC income receipt appeared to increase sodium for women and saturated fat for men. Given prior studies documenting increased substance use in response to income receipt, it is also noteworthy that we find that EITC income lowers smoking-related outcomes (i.e. self-reported smoking for women and cotinine levels for men) and we further find no evidence that EITC income increases alcohol consumption. When repeating our analysis with a different treatment group, those eligible for an EITC of any size, results for many of the outcomes (most

notably, efforts at weight loss, indicators of smoking, and some gender-specific results related to meat consumption) remained consistent, although they were typically more modest in magnitude, which is unsurprising given that the mean qualified credit amount was lower.

There is some evidence that social spending and cash assistance programmes may have important public health consequences.²⁸ However, individuals' psychological and behavioural responses to income receipt are likely to be complex and potentially variable. Consequently, there are ongoing debates as to whether income transfers tend, on average, to be more health promoting or health hindering. Examining the EITC, which is the largest anti-poverty programme in the USA, and considering the wide variety of health-related outcomes that were selected a priori to avoid selective publication bias of only significant findings,²⁹ the above results shed important light on this issue. Contextualizing these results in terms of the existing literature, we note that our results roughly align with existing studies of EITC income. Prior studies have examined how changes in EITC policies predict specific health outcomes, such as infant health,^{30–32} biomarkers,³³ obesity³⁴ and self-assessed health.³³ Most of these studies suggest that

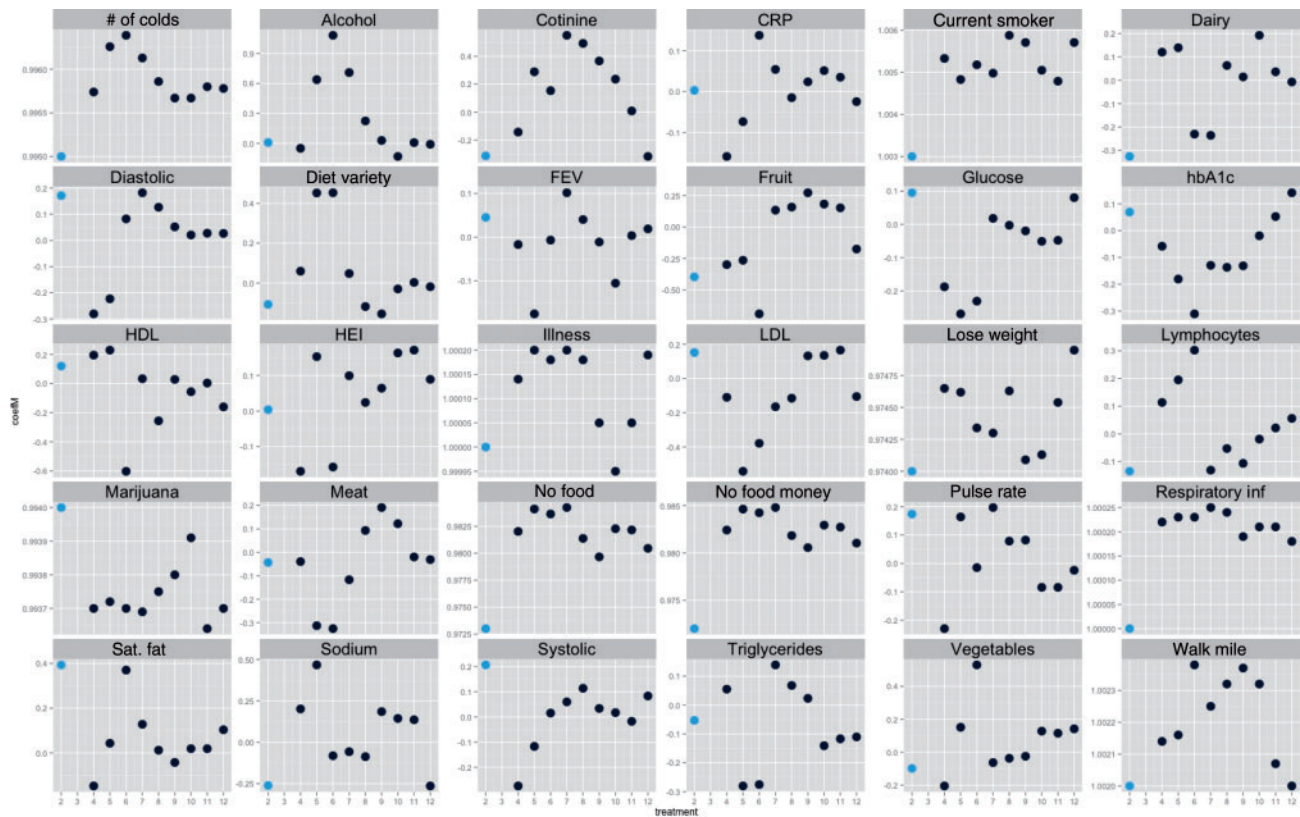


Figure 3. Permutation tests of placebo treatments for difference-in-difference models of EITC receipt and health-related outcomes, men. Light dots indicate point estimate of actual exposure. Black dots indicate point estimates of placebo permutation exposures. Y = axis for each plot differs depending on scale of outcome. X-axis indicates first month of 3-month group of exposure (i.e. 2 indicates true treatment months of Feb, Mar, Apr; 4 indicates placebo treatment months of Apr, May, Jun). FEV, Forced Expiratory Volume; HEI, Healthy Eating Index Overall Diet.

EITCs are health promoting,^{30,32,33} but there are exceptions which reveal deleterious health effects of the EITC.^{31,34} Our analysis, which reveals mixed effects but on balance, more health-promoting effects, conforms to this general pattern in the literature. It is important to highlight, though, that our study stands apart from these prior studies because, whereas most of these studies focus on annual trends and capture effects of both income transfers and employment incentives of the EITC, our study of seasonal fluctuations isolates effects of income transfers from the EITC.

The results challenge prior literature that suggests that substance use and abuse increase in the short term following income receipt.⁴⁻⁶ One possible explanation is that our study has less power to detect effects because we do not know the exact date and amount of income received, thus likely biasing our effect estimates to the null. Consistent with this explanation, our estimates of association are smaller in magnitude than other income disbursement studies.⁷ Other potential explanations for differences in findings are that unlike prior studies of income receipt among military and older individuals,⁶ our sample included very few single men. Since EITC qualification

requires dependent children, and single-parent male-headed households are rare in the USA, most men in our sample are married and cohabiting with children.

A few issues and caveats to our results should be considered. First, this was an intent-to-treat analysis (ITT). Using the NHANES, we could not know when or whether a particular respondent received the EITC. Rather, we compared health outcomes across seasonal periods and subgroups that align with the structure of the EITC. Compared with estimates based on direct measurement of the EITC, this ITT strategy may yield downwardly biased results because of measurement error, but it also avoids a likely source of upward bias if healthier individuals are more likely to actually receive the EITC. No nationally representative US health datasets collect information on individuals' EITC receipt, in part because many people who use tax preparation services are not actually aware of what portion of their tax refund reflected the EITC. Given these issues, and the fact that most EITC-eligible individuals (80–86%⁸) receive the credit, an ITT analysis is likely a preferable strategy for modelling EITC effects. There also is likely some small misclassification of findings because the dependency status of children living with their parents

is unknown. Because of lack of year-round data collection in the Mid-Western and North-Eastern United States, generalizability of our results to these regions is precluded. We also note that our sample size of treated individuals was small (167 women and 111 men), thus our findings should be replicated to determine the stability of our results.

Given our identification strategy, the most likely sources of confounding bias are related to seasonal variations. There are noted seasonal fluctuations in health behaviours and biomarkers,^{35,36} but for these variations to upwardly bias our estimates, EITC-qualifiers would need to have more positive seasonal variation in outcomes than non-qualifiers. For example, individuals in the EITC-qualifying range may be more likely to be working in the new year or more likely to maintain health-promoting behaviours for several months following new-year resolutions.³⁷

Our analysis could not speak to the mechanisms that link EITC income to positive or negative health-related outcomes. It is reasonable to suspect that such mechanisms are related to a combination of material factors, consumption patterns and psychological mechanisms. For example for low-wage workers, receiving at least \$1000 of EITC income may reduce psychological stress related to finances, making it easier to improve health behaviours like reducing smoking. Future researchers involved in more focused primary data collection may be able to shed light on these important issues. Additionally, given the structure of the NHANES, we cannot know how long the short-term effects documented in this study might be sustained over time.

Supplementary Data

Supplementary data are available at *IJE* online.

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