Assessing the Federal Reserve's Index of Household Economic Well-Being: Obama to Trump to Covid to Biden

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1. INTRODUCTION: THE DIMENSIONS OF ECONOMIC WELL-BEING

Many journalists, economists, and commentators have often focused on "economic well-being" (EWB) as an important component of overall well-being for individuals, families, and households. They also have recognized the multidimensional nature of EWB as a concept, and have suggested specific and directly measurable indicators related to these dimensions.

Perhaps the two most obvious candidates for inclusion in a list of dimensions of EWB are monetary wealth and income. In addition, recent economic studies have suggested others such as debt and ability to meet "required" payments (perhaps defined as debt service payments and payments for housing). Other examples are ability to meet unexpected expenses, ability to maintain expense levels in the event of unexpected loss of income (e.g., as a result of job loss), insurance coverage (to protect from unexpected losses) and access to credit.

Survey researchers at the Federal Reserve have made contributions to our knowledge of levels and trends in EWB through their studies with data from the annual national Survey of Household Economic Decision-Making (SHED)conducted every year since 2013 under their auspices. The most recent (2023) version of the SHED includes sections on 20 different substantive areas that may impact the levels of household economic well-being reported by survey respondents. For each of these areas, the SHED questionnaire includes a category of specific and primarily factual questions.

For example, questions in the "Banking" category ask specifically if in the past 12 months the respondent (and/or their spouse or partner) took each of the following actions:

- Currently have a checking, savings or money market account?
- Purchase a money order from a place other than a bank in the past 12 months?
- Cash a check at a place other than a bank in the past 12 month?
- Take out a payday loan or payday advance in the past 12 months?
- Take out a pawn shop loan or an auto title loan in the past 12 months?

¹ The substantive areas are: living arrangements, caregiving, employment, housing, natural disasters, banking, credit applications, credit conditions, cryptocurrency, education, student loans, retirement and investment, income and consumption, inflation, emergency funds, food security, criminal justice, health and insurance, childhood background, and demographics. [1]

- Obtain a tax refund advance to receive your refund faster in the past 12 months?
- Pay an overdraft fee on a bank account in the past 12 months?

Besides these specific and primarily factual questions, the SHED also asks the respondent to choose a response to the following subjective question:

Overall, which one of the following best describes how well you are managing financially these days?

The four possible ordered responses are:

1-"Finding it difficult to get by",2-"Just getting by",3"Doing okay", or 4-"Living comfortably".

This survey strategy follows an approach also used in a number of other surveys dealing with respondents' well-being and satisfaction. For example, in the National Health Interview Survey (NHIS), which is the major federal source for tracking population health levels and trends, the survey asks about specific health problems and diagnoses but also includes an overall subjective rating of the respondent's health status with an ordered 5-category response ranging from "poor" to "excellent".[2]

2. THE FEDERAL RESERVE METHOD FOR CONSTRUCTING AN OVERALL EWB INDICATOR

The ordered categorical response chosen by each SHED respondent represents an overall subjective assessment. That assessment is presumably based on the specific facts of their individual situation. When comparing average responses of differing groups of individuals, however, or when tracking trends in average responses over time, it is often very useful to have a method for aggregating responses across individuals who vary in their chosen categories of the overall subjective assessment.

The analogy to the NHIS also applies here. In tracking overall population health status over time, five separate trends (one for "poor", one for "fair", etc.) showing the precent of respondents in each category separately will rarely be easily translatable into a summary statement about trends for the whole population of respondents. What is needed for this purpose is to apply weights to each of the 5 response categories and to then compute the weighted sum of the respondent percentages in each

of the category. The process for determining these weights is therefore important.

The Federal Reserve has also developed and applied a process for deriving category weights (of the EWB responses in the SHED) that is straightforward and simple.[3, pp. 5-10] A weight of 1 is assigned to each of the 2 highest categories ("doing okay" and "living comfortably"), a weight of 0 is assigned to each of the 2 lowest categories ("finding it difficult" and "just getting by"), and the resulting weighted sum is described as the percent of respondents who are "at least doing okay".

Looking critically at this process, however, raises several concerns. What are the bases for assuming that the two highest responses are actually equivalent? What are the bases for assuming the two lowest responses are actually equivalent? Perhaps most importantly, are these assumptions consistent with the patterns of responses in the SHED data to the large number of specific survey items that are related to the factual (i.e., non-subjective) experiences of the respondents?

3. A SIMPLE AND IMPROVED METHOD FOR OVERALL EWB INDICATOR CONSTRUCTION: THE EWB-REV

There is a relatively simple alternative for constructing an overall EWB indicator that avoids the Federal Reserve's two arbitrary assumptions just described and that also makes use of the substantial information in the SHED on factual experiences of the respondent. This alternative method, which uses an iterative application of OLS regression, has been previously described in detail by Rubinfeld[4] and appears to have been first applied by Rubinfeld in a 1973 study on municipal bond credit ratings.[5] Other applications to voting behavior[4], to labor-force participation [4], and to self-reported health status [6] have also previously appeared in the literature.

In the current paper, we apply this method to the objective and subjective responses in the SHED survey data. The method provides estimates of the subjective values, relative to the highest-valued subjective outcome (i.e., in the case of the SHED data, "living comfortably"). These estimated subjective values can then be used to construct an alternative overall economic well-being measure that can be compared across groups of individuals and/or over time.

The explanatory variables used in this iterative OLS regression analysis were selected from 9 of the 21 categories of questions relating to specific facts that were included in the 2023 SHED survey questionnaire:

Income and consumption
Retirement and investments
Banking
Health and Insurance
Emergency funds
Housing
Credit conditions
Student loans
Demographics.²

A total of 71 variables were selected from these 9 categories for inclusion in the regressions. Most of these are binary variables but 10 are categorical variables and 4 are continuous (with 2 entered in quadratic form). Counting the multiple coefficients for each categorical variable, each regression included estimates for slightly more than 100 coefficients. Specific variable definitions, organized by category, are provided in Appendix A. The steps in the iterative method are described in Table 1.

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² With the large number of measures already included, other categories mentioned in note 1 above were excluded given a residual concern for parsimony. Exclusion of particular categories were based on several rationales. First, while survey data in the 2023 SHED are available on 11,400 respondents, some categories (e.g., cryptocurrency, criminal justice) involved experiences with which very few respondents had any familiarity. Second, for some categories, such as food security, all questions and elicited responses that were arguably too subjective. Of course, with a larger data set and the same method of analysis used here, inclusion of these and other dimensions in would have be possible

TABLE 1: ITERATIVE OLS METHOD FOR ASSIGNING RELATIVE WEIGHTS AND CONSTRUCTING EWB-REV VALUES FROM SHED DATA

Step	Procedure
#	
1	Create a 0-1 dependent variable ("B24dum")set =1 for respondents
	with the best subjective EWB outcome ("living comfortably") and
	set=0 otherwise.
2.	Do OLS regression of B24dum on (a)explanatory variables based on
	the respondents' responses to a number of factual SHED survey
	items and (b)0-1 dummy variables B23dum and B22dum for cases
	with the relevant intermediate EWB outcomes ("doing okay" and
	"just getting by" respectively.
3.	Replace "0" values of B24dum to -1 x (estimated B23dum
	coefficient "B23hat" and -1 x (estimated B22dum coefficient
	"B22hat") from step 2 only for respondents "doing okay" and for
	respondents "just getting by" respectively (Denote this modified
	dependent variable as B24dum*.)
4.	Rerun the OLS regression from Step 2, but use B24dum* as the
	dependent variable and drop B22dum and B23dum from the set of
	explanatory variables in the regression.
5.	Use step 4 regression results to compute predicted values of
	B24dum* only for (a) "doing okay" respondents and (b) for those
	"just getting by". Then compare means of these 2 predicted
	values with values of B22hat and B23hat respectively from step
	2. IF for both B22hat and B23hat mean predicted values are very
	close (e.g., within 0.001), use these mean predicted values as
	weights assigned to "doing okay" and "just getting by"
	respectively. IF NOT repeat steps 4 and 5 over again and again
	until both mean predicted values change by less than 0.001.

We applied this procedure to all 11,400 respondents in the 2023 SHED data using a regression model with the variables from Appendix A and reported regression results in detail in Appendix B.³ Further comments about the regression results are also given in Appendix B.

The process converged after 4 iterations. The initial estimated dummy variable coefficients for the two intermediate responses, and the mean predicted values for each of these responses after each of the 4 iterations following the initial regressions are shown in Table 2.

³ Note that all regressions in the analysis used the population weights provided in the SHED data.

TABLE 2: ESTIMATED EWB-REV VALUES FOR INTERMEDIATKEY SHED RESPONSES FROM ITERATIVE OLS METHOD

		Value	Est Std. Err.	Std. Dev.	P-value
1	Initial Coefficient for "Just	-0.5825566	0.0058507		< 0.0005
	getting by"				
2	Mean Y-hat 1 for "Just				
	getting by"	0.5849422		0.1517908	
3	Mean Y-hat 2 for "Just				
	getting by"	0.5859481		0.1515989	
4	Mean Y-hat 3 for "Just				
	getting by"	0.5878206		0.151993	
5	Mean Y-hat 4 for "Just				
	getting by"	0.5885874		0.152245	
6	FED EWB value for "Just	0			
	getting by"				
7	Initial Coefficient for "Doing	-0.7646814	0.0044681		< 0.0005
	Okay"				
8	Mean Y-hat 1 for "Doing				
	Okay"	0.7685084		0.1297283	
9	Mean Y-hat 2 for "Doing				
	Okay"	0.7691176		0.1294449	
10	Mean Y-hat 3 for "Doing				
	Okay"	0.771163		0.1294057	
11	Mean Y-hat 4 for "Doing				
	Okay"	0.7721098		0.1294489	
12	FED EWB value for "Doing				
	Okay"	1.0			

Rows 1 and 7 show the estimated dummy variable coefficients for the 2 intermediate responses from the initial OLS regression. These coefficients are just -1 multiplied by the mean of the predicted values for cases reporting intermediate subjective responses. Rows 2-5 and 7-11 show the result from the 4 subsequent iterations. Results in rows 5 and 11 are the final estimated EWB-REV values for the cases with either of the 2 intermediate subjective overall economic well-being responses.

Comparison of these 2 estimates with the weights used by the Federal Reserve are also shown in Table 2 (Row 5 vs. Row 6 and Row 11 vs. Row 12). The differences are substantial. The most obvious difference is that in our EWB weight of 0.5885874 for the second lowest response ("Just getting by") was closer to the assigned weight

of 1 for "Living comfortably" than to the FED's assigned weight of 0 which it applies to both "Just getting by" and "Finding it difficult to get by".

Our weight for the second best response ("doing okay" was (0.7721098). This is about halfway between the weights for the second worst and best response; in contrast, the Fed's assigned a weight (1.0) equal to that for the best response ("Living comfortably").

The detailed results from the regressions used to derive our EWB weights are presented and discussed in Appendix B. The the regression results (in Appendices A and B) makes it clear that much of the information in the SHED based on factual experiences of the respondents influence their subejective wellbeing responses, and therefore our estimated subjective EWB weights, in a plausible direction.

4. Comparing EWB-REV vs. FED Weights for Tracking Changes in Overall Economic Well-Being

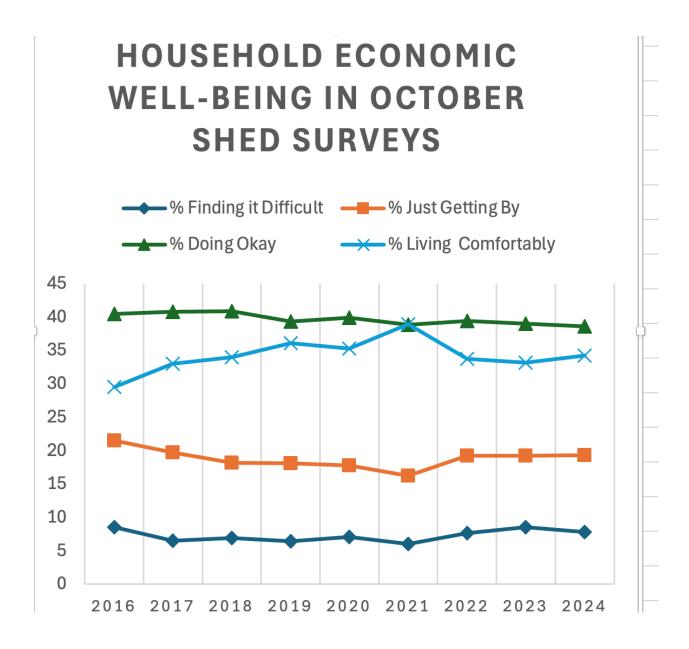
Do the differences between the FED's EWB weights and the regression-based EWB-REV weights matter in terms of tracking changes over time for all U.S. households? We examine the case of the 2017-2024 period which covers the first Trump term, the Covid pandemic, the macroeconomic policy response to Covid, and the last full year of Biden's presidency. Data on the percent of SHED respondents reporting each different level of subjective well-being are shown in Table 3.

Table 3:% Distribution of EWB Responses

Year	% Finding it Difficult	% Just Getting By	% Doing Okay	% Living Comfortably
2016	8.52	21.47	40.49	29.52
2017	6.49	19.7	40.78	33.04
2018	6.91	18.2	40.88	34.01
2019	6.45	18.1	39.31	36.08
2020	7.1	17.79	39.87	35.24
2021	6	16.27	38.81	38.92
2022	7.63	19.24	39.39	33.74
2023	8.57	19.25	39.02	33.17
2024	7.82	19.31	38.63	34.23

(Note that the reported percentages incorporate the population weights in the SHED in each year intended to reflect the entire distribution of U.S. households.) The differing trends in the

response categories can be seen more clearly in the following graph:



While the percentages for each of the four categories are fairly stable, there are several interesting changes over the time period. Going from the last Obama year (2016) to the first Biden year, there were declines in the lowest two categories and a clear increase in the percentage of respondents saying they were living comfortably. The most pronounced annual changes were from 2016 to 2017 with shifts away from the lowest categories and toward "living comfortably"), from 2020 to 2021 with similar changes, and fairly sharp reversals in these trends in 2021 to

2022. These changes coincided with the changes in presidential administrations, with the strong Covid relief effort in 2021, and with the cutbacks in Covid relief and the increase in price inflation from 2021 to 2022.

Applying the FED's EWB weights (0,0,1,1) and the regression-based EWB-REV weights (0,0.5885874,0.7721098,1), we obtain 2 the two resulting time paths of overall national economic well-being (Table 4).

Table 4: FED AND EWB-REV LEVELS AND OCTOBER YEAR-TO-YEAR % CHANGES IN NATIONAL ECONOMIC WELL-BEING FROM SHED DATA

Yr.	National EWB FED	National EWB- REV	FED % Δ Past 12 Mos.	EWB REV % Δ Past 12 Mos.
2016	70.010	73.420		
2017	73.820	76.122	5.442	3.680
2018	74.890	76.286	1.449	0.216
2019	75.390	77.085	0.668	1.047
2020	75.110	76.495	-0.371	-0.765
2021	77.730	78.462	3.488	2.571
2022	73.130	75.478	-5.918	-3.803
2023	72.190	74.628	-1.285	-1.126
2024	72.860	75.422	0.928	1.064

The similarity in time trends between the FED and EWB-REV figures is not surprising since all differences in the weighting scheme apply only to the two intermediate responses. Note also that with either weighting scheme the year-to-year percentage changes are generally small. It is surprising, however, that in the 3 years of the period when year-to-year changes were larger than 1.5%, the percentage changes (in **bold**) with the FED weighting scheme were 35% to 70% larger than the changes with the EWB-REV weighting. This suggests that the resulting EWB level is more volatile over time with the FED wieghts than it is with the EWB-REV weights.

5.Concluding Remarks

This note, drawing on earlier work of Rubinfeld and others, illustrates how the important ordinal subjective response data from the FED annual SHED survey can be combined with respondents' reports of a variety of relevant factual experiences, to produce a weighting scheme for estimating a cardinal index of household economic well-being applicable to

groups of individuals that does not rely on the FED's practice of reducing, by assumption, the four-level ordinal subjective responses the binary response of either "doing at least okay" or "not doing at least okay". Future applications of the EWB-REV weighting scheme to other comparisons of economic well-being between groups of households and over time will hopefully indicate its potential utility for arriving at judgements about policy impacts on household economic well-being.

APPENDIX A: EXPLANATORY VARIABLES IN THE ITERATIVE OLS REGRESSIONS

Variable names and definitions for the explanatory variables included in all OLS regressions are shown in Table Al below. All but 4 of these variables are categorical and the explanations for the category labels and definitions of the reference categories for all non-binary categorical variables are shown in the table. The 4 continuous variables are: (1) combined income of the respondent and spouse/partner expressed as percentile rankings (relrespinc); (2) monthly mortgage or rent payment levels (in \$000's) (housecostK); (3) household size (pphsize) and its square; and (4) respondent age in years and its square.

Note that binary dummy variables for the 2 intermediate level responses to the overall subjective EWB question, denoted by B22dum (for "just getting by)" and B23dum (for "doing okay"), are also included only in the initial OLS regression. The dependent variable in that initial regression is the binary dummy =1 only for the highest level subjective EWB response. That dummy variable is then converted to a 4-level categorical variable in the second and subsequent regressions.

Also note that in the initial regression, the coefficients for the intermediate level binary dummy variables are simply equal to -1 times the mean of the predicted values of the dependent variable for persons in each of the 2 intermediate level response categories. These mean predicted values are used as the initial estimates of the 2 intermediate-level weights whose final values are determined when the iterative OLS process converges (with a tolerance of 0.001).

TABLE A1: VARIABLE NAMES AND DEFINITIONS BY CATEGORY

Category: Income & Consumption

<u>Vble. Name</u> <u>Vble. Definition</u>

ppinc7	Household income 7 categories (<\$10,000 reference):\$10k-\$24,999;\$25k-\$49,999;\$50k-\$74,999;\$#75k-\$99,999;\$100k-\$149,999;\$150k+
120	spending vs. inc. in past mnth. 3 categories(inc. greater ref.):equal;spend greater.
I0_a	=1 if respondent [and/or your spouse or partner] received any income from wages, salaries, or self-employment income in past 12 mos.;=0 otherwise
I0_b	=1 if respondent [and/or your spouse or partner] received any income from Interest, dividends, or rental income in past 12 mos.;=0 otherwise
I0_c	=1 if respondent [and/or your spouse or partner] received any income from
	Social Security (including old age and DI) in past 12 mos.;=0 otherwise
I0_d	=1 if resp. [and/or your spouse or partner] received Supplemental Security
	Inc. (SSI),TANF or cash assist. from a welfare prog. in past 12 mos.;=0 otherwise
I0_e	=1 if respondent [and/or your spouse or partner] received any income from
	Unemployment Insurance in past 12 mos.;=0 otherwise
I0_f	=1 if respondent [and/or your spouse or partner] received any income from a pension in past 12 mos.;=0 otherwise
l41_a	=1 if in past 12 mos. respondent [and/or your spouse or partner] received
	any Earned Income Tax Credit (EITC);=0 otherwise
l41_b	=1 if in past 12 mos. respondent [and/or your spouse or partner] received
	Supplemental Nutrition Assistance (SNAP or food stamps);=0 otherwise
l41_c	=1 if in past 12 mos. respondent [and/or your spouse or partner] received
	Women, Infants, and Children (WIC) nutrition program benefits; =0 otherwise
l41_d	=1 if in past 12 mos. respondent [and/or your spouse or partner] received
	any housing assistance from government programs;=0 otherwise
	=1 if in past 12 mos. respondent [and/or your spouse or partner] received
	free or reduced price school lunches for your children;=0 otherwise
relrespinc	=total combined relative inc. of resp. and spouse/partner in past 12 mos.*

Category: Housing

GH1	Home ownership 4 categories (own with mortgage ref.):own home free and clear with no mortgage; rent; neither own nor pay rent.
housecostK	=approximate rent/mortgage payment each month for respondent (and/or your spouse or partner) in \$1,000's.**

Category: Retirement & Investments

Retired	=1 if retired;=0 otherwise
retplanok	Ret. savings on track 4 categories("no" ref.):1=yes;2=don't know;3=retired.)
K21_a	=1 if currently has retirement savings account (e.g., a 401(k) plan) ;=0 otherwise
K21_b	=1 if curently has pension with a defined benefit through an employer that will pay a monthly amount in retirement;=0 otherwise
K21_c	=1 if curently has stocks, bonds, ETFs, or mutual funds held outside a retirement account; =0 otherwise
K21_d	=1 if curently has savings account, money market account, or cert.of deposit (CD);=0 otherwise
K21_e	=1 if curently has cash value in a life insurance policy;=0 otherwise
K21_f	=1 if has business or real estate investment other than primary residence); =0 otherwise
K5A_anr	=1 if not retired & borrowed from ret. Account in past 12mos.;=0 otherwise
K5A_bnr	=1 if not retired & cashed out (permanently withdrawn) money from retirement account in past 12 mos.;=0 otherwise
K5A_cnr	=1 if not retired & reduced regular contributions to retirement accounts in past 12 mos.;=0 otherwise
ppfs0596	Approx. \$ amt. household savings+ investments, 8 categories:missing (ref),<50k,50k-<100k,100k-<250k,250k-<500k,500k-<1m,1m+,not sure

Category:Student Loans

	Sweet State of the
spousesl	=1 if curent student loan for spouse or partner;=0 no current loan (ref. cat.)
kidgkidsl	=1 if curent student loan for kids or grandkids;=0 no current loan (ref. cat.)
Sl1	=1 if has current student debt for own education; =0 if no current debt for own education (ref. category)
slbehind	=1 if behind on payments or in collections for any student loans from own education;=0 if no loans or not behind or in collections (ref. category).

Category: Demographics

pphsize Household size (number of persons); also number squared	
ppage	Respondent age in years, also age squared
ppgender	="female" if respondent is female;=0 otherwise

Category: Emergency Funds

EFUNDstatus	3 category: Emergency Funds 3 categories: =1 if has funds set aside to cover expenses for 3 mos. if illness, job loss, recession, or other emergencies?; 3= no set aside but could cover 3 months' expenses by borrowing, using savings or selling assets;=2 (ref) if no set aside and can't cover 3 mos. expenses.
EF3_a	=1 if respondent could pay for a \$400 emergency expense by putting on a credit card paying if off in full in next statement;=0 otherwise
EF3_b	=1 if respondent could pay for a \$400 emergency expense by putting on a credit card paying if off over time;=0 otherwise
EF3_c	=1 if respondent could pay for a \$400 emergency expense with money currently in checking/savings account or with cash;=0 otherwise
EF3_d	=1 if respondent could pay for a \$400 emergency expense by using money from a bank loan or line of credit;=0 otherwise
EF3_e	=1 if respondent could pay for a \$400 emergency expense by by borrowing from a friend or family member;=0 otherwise
EF3_f	=1 if respondent could pay a \$400 emergency expense by a payday loan, deposit advance, or overdraft;=0 otherwise
EF3_g	=1 if resp. could pay \$400 emerg. expense by selling something;=0 otherwise
EF3_h	=1 if resp. said unable to pay for the expense right now;=0 otherwise
EF7	5 categories:Based on current financial situation, what is largest emerg. exp. you could handle now using only your savings? (<\$100 reference category):\$100-\$499; \$500-\$999;\$1k-\$1,999;\$2k+
EF6C_rent	=1 if received but did not fully pay rent/mortgage bill in past month;=0 otherwise
EF6C_util	=1 if received but did not fully pay water/gas/electric bill in past month;=0 otherwise
EF6C_tel	=1 if received but did not fully pay phone/internet/cable bill in past month;=0 otherwise
EF6C_car	=1 if received but did not fully pay car payment bill in past month;=0 otherwise

Category: Credit Conditions

lastyr ccardpay	=0 if no credit card currently (ref.);'= 1 if always paid in full) in past 12 mos.;=2 if had unpaid balance only once;=3 if had unpaid balance some of the time; '=4 if had unpaid balance most or all of the time in past 12 months.
lastmnth ccardpay	=0 if have no credit card or used no card in past month (ref.);'=1 if paid at least the min. on all cards in past month;=2 if did not pay or paid < the min. payment on at least one card in past month.
EF5C	=1 if other than credit card bills, paid all bills in full last month;=0 otherwise

Category: Health and Insurance

-	Category:Health and Insurance
E1_a	=1 if during the past 12 mos., there was a time you needed a prescription medicine but went without because you couldn't afford it?=0 otherwise.
E1_b	=1 if during the past 12 mos., was there a time you needed a doctor or specialist but went without because you couldn't afford it?=0 otherwise.
E1_c	=1 if during the past 12 mos., was there a time you needed mental health care or counseling but went without because you couldn't afford it? =0 otherwise.
E1_d	=1 if during the past 12 mos., was there a time you needed dental care but went without because you couldn't afford it? =0 otherwise.
E1_e	=1 if during the past 12 mos., was there a time you needed follow-up care but went without because you couldn't afford it? =0 otherwise.
oopmmedx	Approximately how much did you pay out of pocket for unexpected major medical expenses in the past 12 months? 7 categories (\$0 is ref.):\$1-\$499=1;\$500-\$999=2;\$1k-\$1,999=3;\$2k-\$4,999=4;\$5k+=5;DK=6
E2B	=1 if respondent currently has medical debt;=0 otherwise
E4_a	=1 if respondent covered by employer/union health insurance;=0 otherwise
E4_b	=1 if respondent covered by TRICARE, VA, or other military or veteran's health care;=0 otherwise
E4_c	=1 if respondent covered by Medicare or Medicaid;=0 otherwise
E4_d	=1 if respondent covered by other purchased health insurance;=0 otherwise
E4_e	=1 if respondent covered by insurance purchased through a health insurance insurance exchange;=0 otherwise
E4_f	=1 if respondent covered by any other health insurance;=0 otherwise

BK1	=1 if resp./spouse/partner has no checking, savings or money market account;=0 otherwise
BK2_a	=1 if resp./spouse/partner purchased money order from place other than a bank in past 12 months;=0 otherwise
BK2_b	=1 if resp./spouse/partner cashed a check at a place other than a bank In the past 12 months;=0 otherwise
BK2_c	=1 if resp./spouse/partner took out a payday loan or payday advance in the past 12 mos.;=0 otherwise
BK2_d	=1 if resp./spouse/partner took out a pawn shop loan or an auto title loan. In the past 12 months;=0 otherwise
BK2_e	=1 if resp./spouse/partner obtained a tax refund advance in order to receive refund faster In the past 12 months;=0 otherwise
BK2_fnu	=1 if resp./spouse/partner paid an overdraft fee on a bank account In the past 12 months;=0 otherwise

APPENDIX B: REGRESSION RESULTS FOR INITIAL, SECOND AND FINAL (FOURTH) REGRESSION ITERATIONS INITIAL REGRESSION COMMAND CODE IN STATA

. reg B24dum B22dum B23dum i.ppinc7 i.I20 i.I0_a i.I0_b i.I0_c i.I0_d i.I0_e i.I0_f i.I41_a i.I41_b i.I41_c i.I41_d i.kiddylunch c.relrespinc##c.relrespinc i.GH1 housecostK i.retired i.retplanok i.K21_a i.K21_b i.K21_c i.K21_d i.K21_e i.K21_f i.K5A_anr i.K5A_bnr i.K5A_cnr i.ppfs0596 i.spousesl i.kidgkidsl i.SL1 i.slbehind c.pphhsize##c.pphhsize c.ppage##c.ppage i.ppgender i.EFUNDstatus i.EF3_a i.EF3_b i.EF3_c i.EF3_d i.EF3_e i.EF3_f i.EF3_g i.EF3_h i.EF7 i.EF6C_rent i.EF6C_to i.EF6

Number of obs =

11.400

ITERATION 1 - INITIAL REGRESSION RESULTS

df

Source I

Source	55	αī		umber of o		11,400	
Model	2092.133	105 19 9		(105, 1129		517.47 0.0000	
Residual			504626 R	rob > F -squared	=	0.8279	
	+			dj R-squar			
Total	2527.00424	11,399 .221		oot MSE			
	B24dum	Coefficient	Std. err	. t	P> t	[95% conf.	. interval]
	B22dum	5825566	.0058507	-99.57	0.000	594025	5710882
	B23dum	7646814	.0044681	-171.14	0.000	7734397	7559232
	ppinc7	1					
\$10,0	000 to \$24,999	.0130073	.0120703	1.08	0.281	0106526	.0366671
	000 to \$49,999		.0117042	2.37	0.018	.0047832	.0506677
\$50,0	000 to \$74,999	.0293523	.0124076	2.37	0.018	.0050313	.0536733
\$75 , (000 to \$99 , 999	.0342112	.0128938	2.65	0.008	.0089371	.0594853
\$100,00	00 to \$149 , 999	.039959	.0129617	3.08	0.002	.0145518	.0653663
\$15	50,000 or more	.0550651	.0134487	4.09	0.000	.0287034	.0814268
_	I20						
	as your income	•		-5.80	0.000	0356259	
More tha	an your income	1091848	.005698	-19.16	0.000	1203539	0980157
	IO a						
	Yes	0117566	.0050135	-2.34	0.019	021584	0019291
	I0 b						
	Yes	0018224	.0050749	-0.36	0.720	0117701	.0081254
	I0_c						
	Yes	·	.0070852	-0.60	0.552	0181073	.009669
	I0_d						
	Yes	.0046387	.0094243	0.49	0.623	0138346	.0231119
	I0_e						

Yes	0140647	.0122013	-1.15	0.249	0379813	.0098519
IO_f Yes	.0146178	.0071794	2.04	0.042	.000545	.0286906
I41_a Yes	.0051589	.0074704	0.69	0.490	0094844	.0198021
I41_b Yes	0395275	.007939	-4.98	0.000	0550892	0239658
I41_c Yes	0043052	.0126953	-0.34	0.735	0291902	.0205798
I41_d Yes	0050362	.0120988	-0.42	0.677	0287519	.0186796
1.kiddylunch	0009431	.0085319	-0.11	0.912	0176671	.0157809
relrespinc c.relrespinc		.0353581	1.27 1.41	0.203 0.158	0242736 0186088	.1143425
GH1 Own your home free and c) Pay rent Neither own nor pay rent	.000638 004665 0207715	.0065453 .0054002 .0090323	0.10 -0.86 -2.30	0.922 0.388 0.021	012192 0152502 0384763	.013468 .0059203 0030667
housecostK	0075676	.0024254	-3.12	0.002	0123217	0028134
1.retired	.0519378	.0077233	6.72	0.000	.0367988	.0670769
retplanok 1 2 3	.062445 .0303512	.0057437 .0060819 (omitted)	10.87 4.99	0.000	.0511863 .0184295	.0737036 .0422728
K21_a Yes K21 b	 0061952	.0051339	-1.21	0.228	0162585	.0038682
Yes K21_c	0081737	.0050704	-1.61	0.107	0181126	.0017653
Yes K21 d	.0036746	.0049871	0.74	0.461	0061011	.0134502
Yes K21 e	000042	.0047403	-0.01	0.993	0093338	.0092498
Yes K21 f	.0084915	.0045142	1.88	0.060	000357	.0173401
Yes	.0088637	.0064177	1.38	0.167	0037161	.0214435
1.K5A_anr	.0016398	.0092881	0.18	0.860	0165665	.0198461
1.K5A_bnr	.0079616	.010083	0.79	0.430	0118029	.027726

1.K5A_cnr	0278855	.0081385	-3.43	0.001	0438383	0119326
ppfs0596	 					
Under \$50,000	0056575	.0058016	-0.98	0.330	0170297	.0057147
\$50,000 - \$99,999	0040813	.0075096	-0.54	0.587	0188014	.0106388
\$100,000 - \$249,999		.007404	0.18	0.860	0132028	.0158236
\$250,000 - \$499,999	.0072592	.0085211	0.85	0.394	0094436	.0239621
\$500,000 - \$999,999	.0187697	.0091308	2.06	0.040	.0008717	.0366676
\$1,000,000 or more	.0414644	.0092679	4.47	0.000	.0232978	.059631
Not sure	.0306009	.0096416	3.17	0.002	.0117017	.0495
1.spousesl	 0073745	.0074371	-0.99	0.321	0219525	.0072035
1.kidgkidsl		.0090671	-0.19	0.850	0194854	.0160609
1.1149111401		•0030071	0.13	0.000	• 0 1 3 1 0 0 1	•0100003
SL1		0064306	0 05	0 245	0065201	010600
Yes	.0060789	.0064326	0.95	0.345	0065301	.018688
1.slbehind	0234428	.0142755	-1.64	0.101	0514252	.0045397
pphhsize	0075423	.0041326	-1.83	0.068	0156429	.0005583
c.pphhsize#c.pphhsize	.0005201	.0005017	1.04	0.300	0004633	.0015035
ppage		.0007408	-5.89	0.000	0058124	0029081
<pre>c.ppage#c.ppage</pre>	.0000341	7.69e-06	4.43	0.000	.000019	.0000491
ppgender	 					
Female	.0212128	.0038043	5.58	0.000	.0137557	.0286699
EFUNDstatus		0066657	2 65	0 000	0110060	0074007
1	.0243628	.0066657	3.65	0.000	.0112969	.0374287
3	.0332985	.006599	5.05	0.000	.0203632	.0462337
EF3_a						
Put it on my credit card	.0267347	.0053522	5.00	0.000	.0162434	.037226
EF3 b						
Put it on my credit card	0117696	.0062263	-1.89	0.059	0239743	.0004351
EF3_c						
With the money currently	.0334275	.004877	6.85	0.000	.0238678	.0429872
EF3 d						
Using money from a bank 1		.0116607	-1.93	0.054	0453584	.0003557
EF3 e		.0110007	1.55	0.001	.0133301	.0003337
		0072052	-4.24	0.000	045205	016656
By borrowing from a frien	•	.0073052	-4.24	0.000	045295	016656
EF3_f						
Using a payday loan, depo		.0159737	1.13	0.258	0132351	.0493873
EF3_g						
By selling something	0216717	.0081796	-2.65	0.008	0377051	0056383
EF3 h						
I wouldn't be able to pay	0788397	.0077468	-10.18	0.000	0940247	0636547
1 1 1 1 1 1 1						
	1					

EF7 \$100 to \$499 \$500 to \$999 \$1,000 to \$1,999 \$2,000 or more	.0676047 .0681082 .0797328 .1097259	.0074461 .0088595 .0094715 .0092226	9.08 7.69 8.42 11.90	0.000 0.000 0.000 0.000	.0530089 .0507421 .061167 .0916481	.0822004 .0854742 .0982986 .1278037
1.EF6C_rent 1.EF6C_util 1.EF6C_car 1.EF6C_tel	0119036	.0135867 .0112362 .0136954 .0124803	-5.52 -1.06 -0.48 -1.67	0.000 0.289 0.633 0.095	1016291 0339286 0333819 0453004	0483646 .0101213 .0203087 .0036267
lastyrccardpay 1 2 3 4	.0254937 .0364911 .0127026 0103074	.0098825 .0126341 .0107662 .0109476	2.58 2.89 1.18 -0.94	0.010 0.004 0.238 0.346	.0061223 .0117261 008401 0317665	.0448651 .0612562 .0338062 .0111518
lastmnthccardpay 1 2	0092023 0302838	.0083952	-1.10 -2.07	0.273 0.038	0256583 0589642	.0072537 0016034
EF5C Yes	.0051704	.0069106	0.75	0.454	0083756	.0187163
E1_a Yes E1_b	0039868	.008013	-0.50	0.619	0196937	.0117201
Yes E1_c Yes	0198904 0420242	.0076385	-2.60 -5.44	0.009	0348632 0571746	0049176 0268738
E1_d Yes	0364122	.0060598	-6.01	0.000	0482904	0245341
E1_e Yes	0114613	.0088166	-1.30	0.194	0287432	.0058207
oopmajmedexp 1 2 3 4 5	0033287 .0009551 0074927 007007 0168551 .0083226	.0092819 .0091033 .0087484 .0088916 .0115244 .0206117	-0.36 0.10 -0.86 -0.79 -1.46 0.40	0.720 0.916 0.392 0.431 0.144 0.686	0215228 016889 0246412 0244361 0394449 0320799	.0148653 .0187993 .0096557 .010422 .0057347 .0487252
E2B Yes	0249476	.0058158	-4.29	0.000	0363475	0135476
E4_a Yes	.0026059	.0053432	0.49	0.626	0078676	.0130795

E4_b						
Yes E4 c	.0163645	.0065213	2.51	0.012	.0035816	.0291475
Yes	0129322	.0061873	-2.09	0.037	0250604	0008039
E4_d Yes E4 e	008948	.0081973	-1.09	0.275	0250161	.00712
Yes	.0172945	.0097425	1.78	0.076	0018024	.0363915
E4_f Yes	.0315896	.0079054	4.00	0.000	.0160937	.0470856
BK1 Yes	0396257	.0091243	-4.34	0.000	0575108	0217405
BK2_a Yes BK2 b	.0142189	.0070473	2.02	0.044	.000405	.0280327
Yes	.0171529	.0080266	2.14	0.033	.0014193	.0328865
BK2_c Yes BK2 d	043098	.0123735	-3.48	0.000	0673522	0188438
Yes	.0083252	.0130221	0.64	0.523	0172003	.0338507
BK2_e Yes 1.BK2_fnu _cons	.0275577 0281881 .756412	.0165371 .0067967 .0247459	-4.15	0.096 0.000 0.000		

ITERATION 2 - STATA CODE FOR RECODING DEPENDENT VARIABLE

. gen B24dum1=B24dum

- . replace B24dum1=0.7646814 if B23dum==1
- . replace B24dum1=0.5825566 if B22dum==1

ITERATION 2 - STATA REGRESSION COMMAND

. reg B24duml i.ppinc7 i.I20 i.I0_a i.I0_b i.I0_c i.I0_d i.I0_e i.I0_f i.I41_a i.I41_b i.I41_c i.I41_d i.kiddylunch c.relrespinc#c.relrespinc i.GH1 housecostK i.retired i.retplanok i.K21_a i.K21_b i.K21_c i.K21_d i.K21_e i.K21_f i.K5A_anr i.K5A_bnr i.K5A_cnr i.ppfs0596 i.spousesl i.kidgkidsl i.SL1 i.slbehind c.pphhsize#c.pphhsize c.ppage##c.ppage i.ppgender i.EFUNDstatus i.EF3_a i.EF3_b i.EF3_c i.EF3_d i.EF3_e i.EF3_f i.EF3_g i.EF3_h i.EF7 i.EF6C_rent i.EF6C_tol i.EF6C_car i.EF6C_tol i.lastyrccardpay i.lastmnthccardpay i.EF5C i.E1_a i.E1_b i.E1_c i.E1_d i.E1_e i.oopmajmedexp i.E2B i.E4_a i.E4_b i.E4_c i.E4_d i.E4_e i.E4_f i.BK1 i.BK2_a i.BK2_b i.BK2_c i.BK2_d i.BK2_e i.BK2_fnu [aweight=weight pop]

ITERATION 2 - REGRESSION RESULTS

Source	SS	df	MS	Number of obs	=	11,400
 +-				F(103, 11296)	=	104.02
Model	412.486833	103	4.00472653	Prob > F	=	0.0000
Residual	434.871241	11,296	.038497808	R-squared	=	0.4868
 +-				Adj R-squared	=	0.4821
Total	847.358073	11,399	.074336176	Root MSE	=	.19621

B24dum1	Coefficient	Std. err.	t	P> t	[95% conf.	. interval]
ppinc7	+ 					
\$10,000 to \$24,999	.0130073	.0120689	1.08	0.281	0106498	.0366643
\$25,000 to \$49,999	.0277255	.0117018	2.37	0.018	.0047879	.050663
\$50,000 to \$74,999	.0293523	.0124064	2.37	0.018	.0050336	.053671
\$75,000 to \$99,999	.0342112	.0128894	2.65	0.008	.0089458	.0594766
\$100,000 to \$149,999	.039959	.0129582	3.08	0.002	.0145588	.0653593
\$150,000 or more	.0550651	.0134431	4.10	0.000	.0287142	.081416
120	[[
The same as your income	0266314	.0045586	-5.84	0.000	035567	0176959
More than your income	1091848	.0056825	-19.21	0.000	1203235	0980462
10.0						
IO_a Yes	 0117566	.0050097	-2.35	0.019	0215764	0019367
I0_b		0050500	0.06	0 510	0117601	0001000
Yes	0018224	.0050739	-0.36	0.719	0117681	.0081233
I0 c						
Yes	0042192	.0070844	-0.60	0.551	0181058	.0096675
IO d	 					
Yes	.0046387	.0094231	0.49	0.623	0138322	.0231095
195		.0031201	0.13	0.020	.0100022	.0201030
I0_e						
Yes	0140647	.0122	-1.15	0.249	0379789	.0098495
IO f	 					
Yes	.0146178	.0071735	2.04	0.042	.0005565	.0286791
I41_a Yes	 .0051589	.0074674	0.69	0.490	0094784	.0197962
162	.0031309	.00/40/4	0.09	0.490	.0094704	.0191902
I41_b						
Yes	0395275	.0079372	-4.98	0.000	0550858	0239692

I41_c Yes	0043052	.0126925	-0.34	0.734	0291848	.0205744	
I41_d Yes 1.kiddylunch relrespinc	0050362 0009431 .0450344	.0120975 .0085309 .0353345	-0.42 -0.11 1.27	0.677 0.912 0.203	0287493 0176652 0242274	.018677 .015779 .1142962	
c.relrespinc#c.relrespinc	.047881	.0338523	1.41	0.157	0184754	.1142374	
GH1 Own your home free and c) Pay rent Neither own nor pay rent	.000638 004665 0207715	.0065439 .0053977 .0090311	0.10 -0.86 -2.30	0.922 0.387 0.021	0121892 0152455 038474	.0134652 .0059156 003069	
housecostK 1.retired	0075676 .0519378	.0024249	-3.12 6.73	0.002	0123209 .0368116	0028143 .0670641	
retplanok 1 2 3	.062445 .0303512	.0057109 .0060662 (omitted)	10.93 5.00	0.000	.0512506	.0736393	
K21_a Yes	0061952	.0051333	-1.21	0.228	0162574	.0038671	
K21_b Yes	0081737	.0050696	-1.61	0.107	018111	.0017637	
K21_c Yes	.0036746	.0049856	0.74	0.461	0060982	.0134473	
K21_d Yes	000042	.0047393	-0.01	0.993	0093318	.0092478	
K21_e Yes	.0084915	.0045134	1.88	0.060	0003556	.0173387	
K21_f Yes 1.K5A_anr 1.K5A_bnr 1.K5A_cnr	.0088637 .0016398 .0079616 0278855	.0064161 .0092812 .0100817 .0081334	1.38 0.18 0.79 -3.43	0.167 0.860 0.430 0.001	0037128 016553 0118003 0438284	.0214403 .0198325 .0277234 0119426	
ppfs0596 Under \$50,000 \$50,000 - \$99,999 \$100,000 - \$249,999	0056575 0040813 .0013104	.0058007 .0075082 .0074033	-0.98 -0.54 0.18	0.329 0.587 0.860	017028 0187988 0132013	.0057129 .0106362 .0158222	

\$250,000 - \$499,999 \$500,000 - \$999,999 \$1,000,000 or more Not sure	.0072592 .0187697 .0414644 .0306009	.0085201 .0091246 .0092396 .0096393	0.85 2.06 4.49 3.17	0.394 0.040 0.000 0.002	0094417 .0008838 .0233531 .0117061	.0239602 .0366555 .0595757 .0494956
1.spousesl 1.kidgkidsl	0073745 0017123	.0074356	-0.99 -0.19	0.321 0.850	0219495 0194792	.0072005
SL1 Yes 1.slbehind pphhsize	.0060789	.0064314 .0142703 .0041305	0.95 -1.64 -1.83	0.345 0.100 0.068	0065277 0514151 0156387	.0186856 .0045296 .0005541
c.pphhsize#c.pphhsize	.0005201	.0005016	1.04	0.300	0004631	.0015032
ppage	0043602	.0007404	-5.89	0.000	0058115	002909
c.ppage#c.ppage	.0000341	7.68e-06	4.43	0.000	.000019	.0000491
ppgender Female	.0212128	.0037993	5.58	0.000	.0137656	.0286601
EFUNDstatus 1 3	.0243628	.0066567	3.66 5.05	0.000	.0113146	.037411
EF3_a Put it on my credit card	.0267347	.005346	5.00	0.000	.0162555	.0372139
EF3_b Put it on my credit card	0117696	.0062214	-1.89	0.059	0239647	.0004255
$${\rm EF3_c}$$ With the money currently	.0334275	.0048745	6.86	0.000	.0238726	.0429824
EF3_d Using money from a bank l	0225013	.0116596	-1.93	0.054	0453561	.0003535
EF3_e By borrowing from a frien	0309755	.0073037	-4.24	0.000	0452919	016659
EF3_f Using a payday loan, depo	.0180761	.0159705	1.13	0.258	0132289	.0493811
EF3_g By selling something	0216717	.0081752	-2.65	0.008	0376965	0056469
EF3_h						

I wouldn't be able to pay	0788397	.0077344	-10.19	0.000	0940005	0636789	
EF7							
\$100 to \$499	.0676047	.0074365	9.09	0.000	.0530279	.0821814	
\$500 to \$999	.0681082	.0088242	7.72	0.000	.0508111	.0854052	
\$1,000 to \$1,999	.0797328	.0094437	8.44	0.000	.0612214	.0982442	
\$2,000 or more	.1097259	.009192	11.94	0.000	.091708	.1277438	
, , , , , , , , , , , , , , , , , , , ,							
1.EF6C rent	0749968	.0135686	-5.53	0.000	1015937	0484	
1.EF6C util		.0112352	-1.06	0.289	0339266	.0101194	
1.EF6C car		.0136923	-0.48	0.633	0333759	.0203028	
1.EF6C tel	0208369	.0124758	-1.67	0.095	0452916	.0036179	
1.Eroc_cer	.0200309	.0124730	1.07	0.095	.0452910	.0030179	
lastyrccardpay							
1	.0254937	.0098806	2.58	0.010	.0061261	.0448614	
2		.0126321	2.89	0.004	.0117301	.0612522	
3	.0127026	.0107609	1.18	0.238	0083906	.0337958	
4	0103074	.0107003	-0.94	0.346	0317609	.0111461	
4	0103074	.0109447	-0.94	0.540	0317009	.0111401	
lastmnthccardpay							
	000000	0002024	1 10	0 272	0256520	0070400	
1	0092023	.0083924	-1.10	0.273	0256528	.0072482	
2	0302838	.0146252	-2.07	0.038	0589517	0016159	
_							
EF5C							
Yes	.0051704	.0069095	0.75	0.454	0083735	.0187142	
I							
E1_a							
Yes	0039868	.0080104	-0.50	0.619	0196886	.011715	
I							
E1 b							
Yes	0198904	.0076362	-2.60	0.009	0348586	0049221	
j							
E1 c							
Yes	0420242	.0077276	-5.44	0.000	0571717	0268767	
100	• 0 12 02 12	•0077270	0.11	0.000	•00/1/1/	•0200707	
E1 d							
Yes	0364122	.0060566	-6.01	0.000	0482842	0245402	
165	.0301122	.0000000	0.01	0.000	.0102012	.0210102	
E1 e							
Yes	0114613	.0088082	-1.30	0.193	0287269	.0058044	
169	.0114013	.0000002	1.50	0.193	.0207209	.0050044	
a anna ima da un							
oopmajmedexp	0022207	0000001	0.26	0 700	0015104	0140610	
1	0033287	.0092801	-0.36	0.720	0215194	.0148619	
2	.0009551	.009096	0.11	0.916	0168745	.0187848	
3	0074927	.0087452	-0.86	0.392	0246349	.0096495	
4	007007	.0088907	-0.79	0.431	0244343	.0104203	
5	0168551	.0115221	-1.46	0.144	0394404	.0057301	
6	.0083226	.020608	0.40	0.686	0320725	.0487178	

E2B Yes	•	.0058141	-4.29	0.000	0363443	0135509
E4_a Yes	 .0026059	.0053421	0.49	0.626	0078656	.0130774
E4_b Yes		.0065196	2.51	0.012	.003585	.0291441
E4_c Yes		.0061867	-2.09	0.037	0250591	0008052
E4_d Yes		.0081948	-1.09	0.275	0250113	.0071152
E4_e Yes		.0097406	1.78	0.076	0017988	.0363878
E4_f Yes		.0079032	4.00	0.000	.016098	.0470812
BK1 Yes	0396257	.0091232	-4.34	0.000	0575087	0217426
BK2_a Yes	 .0142189	.0070448	2.02	0.044	.0004099	.0280279
BK2_b Yes	 .0171529	.0080243	2.14	0.033	.0014239	.0328819
BK2_c Yes	043098	.0123711	-3.48	0.000	0673476	0188485
BK2_d Yes	.0083252	.0130205	0.64	0.523	0171973	.0338477
BK2_e Yes 1.BK2_fnu _cons	.0275577	.0165346 .0067921 .0246388	1.67 -4.15 30.70	0.096 0.000 0.000	0048529 0415018 .7081156	.0599684 0148743 .8047083

ITERATION 2 - RECODING DEPENDENT VARIABLE (B24dum1) VALUES FOR INTERMEDIATE RESPONSES

*****USE PREDICT COMMAND TO OBTAIN MEAN VALUES OF PREDICTIONS FOR INTERMEDIATE CASES
(B23dum==1 for "Doing okay and B22dum==1 for "Just getting by").

. predict B23hatnu1 if B23dum==1
(option xb assumed; fitted values)
. predict B22hatnu1 if B22dum==1
(option xb assumed; fitted values)

. sum B22hatnu1 B23hatnu1

Variable	Obs	Mean	Std. dev.	Min	Max
B22hatnu1	2,137	.5849422	.1517908	.0882174	1.002591
B23hatnu1	4,450	.7685084	.1297283	.1601671	1.067576

*****RECODE DEPENDENT VARIABLE VALUES FOR INTERMEDIATE CASES WITH MEANS OF PREDICTED VALUES

- . replace B24dum1=.765084 if B23dum==1
- . replace B24dum1=.5849422 if B22dum==1

. ta	ab	B24	1dur	n1

Cum.	Percent	Freq.	B24dum1
8.24 26.98 66.02 100.00	8.24 18.75 39.04 33.98	939 2,137 4,450 3,874	0 .5849422 .765084
	100.00	+ 11,400	Total

ITERATION 3 - REGRESSION RESULTS

Source	SS	df	MS	Number of obs F(103, 11296)	=	11,400 103.62
	410.898561 434.878027		3.98930641 .038498409	Prob > F R-squared	=	0.0000
+- Total	845.776588	11,399	.074197437	Adj R-squared Root MSE	=	0.4811 .19621

B24d	um1	Coefficient	Std. err.	t	P> t	[95% conf.	interval]
ppi	nc7						
\$10,000 to \$24,9	99	.0130363	.0120689	1.08	0.280	0106209	.0366936
\$25,000 to \$49,9	99	.0278002	.0117019	2.38	0.018	.0048625	.0507379
\$50,000 to \$74,9	99	.0293362	.0124065	2.36	0.018	.0050173	.0536551
\$75,000 to \$99,9	99	.0340951	.0128895	2.65	0.008	.0088295	.0593607
\$100,000 to \$149,9	99	.0398558	.0129583	3.08	0.002	.0144554	.0652563
\$150,000 or mo	re	.0549327	.0134432	4.09	0.000	.0285816	.0812837

 I20						
The same as your income	0264539	.0045586	-5.80	0.000	0353895	0175182
More than your income	1090225	.0056825	-19.19	0.000	1201612	0978838
IO a						
Yes	0117537	.0050097	-2.35	0.019	0215737	0019338
 d 01						
Yes	0018333	.0050739	-0.36	0.718	011779	.0081125
j						
IO_c		0070045	_0 50	0.553	_ 0100000	0006930
Yes	0042029	.0070845	-0.59	0.553	0180896	.0096839
I0_d						
Yes	.0046721	.0094232	0.50	0.620	0137989	.0231432
I0_e						
Yes		.0122001	-1.15	0.250	0379618	.0098669
TO 6						
IO_f Yes	.0145509	.0071735	2.03	0.043	.0004895	.0286123
	.0110000	100.1.00	2.00	3.010	.0001090	,0200120
I41_a		0074654	0.60	0 400	000511	0107600
Yes	.0051258	.0074674	0.69	0.492	0095116	.0197633
I41_b						
Yes	0395384	.0079373	-4.98	0.000	0550968	0239799
I41 c						
Yes	0042234	.0126926	-0.33	0.739	0291032	.0206564
i						
I41_d Yes	0050021	.0120976	-0.41	0.679	0287155	.0187112
res 1.kiddylunch		.0120976	-0.41	0.679	0287155	.0157783
relrespinc		.0353348	1.27	0.202	0242195	.1143052
TOTTESPINE	.0150125	.0333310	±• ∠ /	0.202	.0212193	•1110002
c.relrespinc#c.relrespinc	.0476539	.0338526	1.41	0.159	0187031	.1140108
GH1						
Own your home free and c)	.0006142	.006544	0.09	0.925	0122131	.0134416
Pay rent	0046053	.0053978	-0.85	0.923	0151859	.0059753
Neither own nor pay rent	0207904	.0090311	-2.30	0.021	038493	0030878
Mercher own her pay rene	.0207504	.0000011	2.50	0.021	.000190	.0030070
housecostK	0075587	.002425	-3.12	0.002	0123121	0028054
1.retired	.051825	.0077168	6.72	0.000	.0366986	.0669513
retplanok						
Techtanov I						

1 2 3	.062221 .0302111 0	.0057109 .0060663 (omitted)	10.90	0.000	.0510265	.0734154
K21_a Yes	 0061803	.0051334	-1.20	0.229	0162426	.0038821
K21_b Yes	 0081668 	.0050697	-1.61	0.107	0181042	.0017707
K21_c Yes	 .0036416 	.0049857	0.73	0.465	0061312	.0134143
K21_d Yes	 000011 	.0047393	-0.00	0.998	0093009	.0092789
K21_e Yes	 .0084702 	.0045135	1.88	0.061	000377	.0173174
K21_f Yes 1.K5A_anr 1.K5A_bnr 1.K5A_cnr	.0088358	.0064161 .0092813 .0100818 .0081335	1.38 0.19 0.79 -3.42	0.169 0.848 0.431 0.001	0037409 0164151 0118278 0437434	.0214125 .0199707 .0276962 0118573
ppfs0596 Under \$50,000 \$50,000 - \$99,999 \$100,000 - \$249,999 \$250,000 - \$499,999 \$500,000 - \$999,999 \$1,000,000 or more Not sure	 0056352 0041095 .0013229 .0072677 .0187433 .0414118 .0305797	.0058008 .0075083 .0074034 .0085202 .0091247 .0092397 .0096394	-0.97 -0.55 0.18 0.85 2.05 4.48 3.17	0.331 0.584 0.858 0.394 0.040 0.000	0170057 0188271 013189 0094333 .0008573 .0233003 .0116848	.0057353 .0106081 .0158347 .0239688 .0366292 .0595232 .0494746
1.spousesl 1.kidgkidsl	0074082 0016467	.0074356	-1.00 -0.18	0.319 0.856	0219833 0194138	.007167
SL1 Yes 1.slbehind pphhsize	.0060546	.0064315 .0142705 .0041305	0.94 -1.65 -1.82	0.347 0.099 0.069	0065522 051525 0155982	.0186613 .0044202 .0005948
c.pphhsize#c.pphhsize	.0005173	.0005016	1.03	0.302	0004658	.0015005
ppage	0043525	.0007404	-5.88	0.000	0058038	0029012
c.ppage#c.ppage	.000034	7.68e-06	4.42	0.000	.0000189	.000049

ppgender Female	.021134	.0037993	5.56	0.000	.0136867	.0285813	
EFUNDstatus 1 3	.0242813 .0331621	.0066567	3.65 5.03	0.000	.011233	.0373296	
EF3_a Put it on my credit card	 .026666	.0053461	4.99	0.000	.0161867	.0371452	
EF3_b Put it on my credit card	0116724	.0062215	-1.88	0.061	0238676	.0005228	
EF3_c With the money currently	 .0333761 	.0048745	6.85	0.000	.0238212	.0429311	
EF3_d Using money from a bank 1	 0225246 	.0116597	-1.93	0.053	0453795	.0003304	
EF3_e By borrowing from a frien	 031014 	.0073037	-4.25	0.000	0453305	0166974	
Using a payday loan, depo	 .0179972 	.0159706	1.13	0.260	013308	.0493024	
EF3_g By selling something	021614 	.0081753	-2.64	0.008	037639	0055891	
EF3_h I wouldn't be able to pay	0788825	.0077345	-10.20	0.000	0940433	0637216	
\$100 to \$499 \$500 to \$999 \$1,000 to \$1,999 \$2,000 or more	.0675516 .0679596 .0795779 .1094265	.0074365 .0088243 .0094438 .0091921	9.08 7.70 8.43 11.90	0.000 0.000 0.000 0.000	.0529747 .0506624 .0610664 .0914084	.0821285 .0852568 .0980894 .1274445	
1.EF6C_rent 1.EF6C_util 1.EF6C_car 1.EF6C_tel	0119027 0066257	.0135687 .0112353 .0136924 .0124759	-5.55 -1.06 -0.48 -1.68	0.000 0.289 0.628 0.093	1018477 0339259 0334652 0453945	0486535 .0101204 .0202139 .0035153	
lastyrccardpay 1 2 3 4	.0254456 .0365025 .0127322 010265	.0098806 .0126322 .0107609 .0109448	2.58 2.89 1.18 -0.94	0.010 0.004 0.237 0.348	.0060778 .0117413 0083611 0317187	.0448133 .0612637 .0338256 .0111887	

E1_a	0186937
	0116491
E1_b Yes 0198257 .0076362 -2.60 0.0090347940	0048573
E1_c Yes 0419785 .0077277 -5.43 0.00005712620	0268309
E1_d Yes 0363379 .0060567 -6.00 0.000048210	0244658
E1_e Yes 0115947 .0088083 -1.32 0.1880288605 .0	0056711
oopmajmedexp	
1 0033828 .0092802 -0.36 0.7150215736	.014808
	0189293
3 0074589 .0087453 -0.85 0.3940246012 .0	0096834
	0104311
5 0168539 .0115221 -1.46 0.1440394394 .0	0057315
6 .0084276 .0206081 0.41 0.6830319679 .0	0488231
E2B	
Yes 0249108 .0058142 -4.28 0.00003630760	0135141
E4_a Yes .0026318 .0053422 0.49 0.6220078398 .0	0131034
E4_b Yes .0163811 .0065197 2.51 0.012 .0036014 .0	0291607
E4_c Yes 0129444 .0061867 -2.09 0.03602507140	0008173
E4_d Yes 0088796 .0081949 -1.08 0.279024943 .0	0071838
E4_e Yes .0173525 .0097407 1.78 0.075001741 .0	0364459

E4_f Yes	.0316099	.0079032	4.00	0.000	.0161182	.0471017
BK1 Yes	 0396514	.0091233	-4.35	0.000	0575346	0217682
BK2_a Yes	.0141921	.0070448	2.01	0.044	.000383	.0280012
BK2_b Yes	.0171371	.0080243	2.14	0.033	.001408	.0328662
BK2_c Yes	 0431234 	.0123712	-3.49	0.000	0673731	0188737
BK2_d Yes	 .0083679 	.0130206	0.64	0.520	0171548	.0338906
BK2_e Yes 1.BK2_fnu _cons	 .0275719 0281878 .7572061	.0165347 .0067922 .024639	1.67 -4.15 30.73	0.095 0.000 0.000	004839 0415016 .7089093	.0599828 0148739 .8055028

ITERATION 4 REGRESSION RESULTS

Source |

(Using dependent variable recodes for intermediate responses based on Iteration 3 regression results.)

df MS Number of obs = 11,400

Model Residual Total	411.424311 434.952441 846.376751	11,296 .	.99441078 038504997 	F(103, 112) Prob > F R-squared Adj R-squa: Root MSE	= =	103.74 0.0000 0.4861 0.4814 .19623	
	B24dum1	Coeffici		err. t		[95% conf	. interval]
\$25,0 \$50,0 \$75,0 \$100,00	ppinc7 000 to \$24,999 000 to \$49,999 000 to \$74,999 000 to \$99,999 00 to \$149,999 00,000 or more	. 0129 .02774 .02936 .03427 .03982 .05480	37 .01 48 .0117 21 .0124 29 .0128 03 .0129	207 1.07 029 2.37 076 2.37 906 2.66 594 3.07	0.284 0.018 0.018	0107223 .0048052 .0050412 .0090051 .0144177 .0284474	.0365963 .0506845 .0536831 .0595407 .0652229 .0811541

I20 The same as your income More than your income	 0260931 1092999	.004559	-5.72 -19.23	0.000	0350295 1204396	0171567 0981603
IO_a Yes	 0114905	.0050102	-2.29	0.022	0213113	0016697
I0_b Yes	 0019351	.0050744	-0.38	0.703	0118818	.0080115
IO_c Yes	0042383 	.0070851	-0.60	0.550	0181263	.0096496
IO_d Yes	 .004691	.009424	0.50	0.619	0137816	.0231636
IO_e Yes	 0141194	.0122012	-1.16	0.247	0380359	.009797
IO_f Yes	 .0142598	.0071742	1.99	0.047	.0001972	.0283224
I41_a Yes	 .0048959	.0074681	0.66	0.512	0097428	.0195346
I41_b Yes	 0397133	.007938	-5.00	0.000	0552731	0241535
I41_c Yes	0043452	.0126937	-0.34	0.732	0292271	.0205367
I41_d Yes 1.kiddylunch relrespinc	0050229 0010274	.0120986 .0085317 .0353378	-0.42 -0.12 1.32	0.678 0.904 0.186	0287382 017751 0225085	.0186925 .0156963 .1160281
c.relrespinc#c.relrespinc	.0448441	.0338555	1.32	0.185	0215185	.1112067
GH1 Own your home free and c) Pay rent Neither own nor pay rent housecostK 1.retired	.0004977 0046692 0208844 0075261	.0065445 .0053982 .0090319 .0024252	0.08 -0.86 -2.31 -3.10 6.70	0.939 0.387 0.021 0.002 0.000	0123307 0152507 0385885 0122799 .0365502	.0133262 .0059123 0031803 0027724 .0668055
retplanok	.0618957	.0057114	10.84	0.000	.0507003	.0730911

2 3	.0306556	.0060668 (omitted)	5.05	0.000	.0187636	.0425476	
K21_a Yes	 0061867	.0051338	-1.21	0.228	0162498	.0038765	
K21_b Yes	 0080857	.0050701	-1.59	0.111	018024	.0018526	
K21_c Yes	.0035592	.0049861	0.71	0.475	0062144	.0133328	
K21_d Yes	 0000339	.0047397	-0.01	0.994	0093246	.0092568	
K21_e Yes	 .0085039	.0045139	1.88	0.060	000344	.0173519	
K21_f Yes 1.K5A_anr 1.K5A_bnr 1.K5A_cnr	.0087096 .0016201	.0064167 .0092821 .0100826 .0081342	1.36 0.17 0.78 -3.39	0.175 0.861 0.436 0.001	0038682 0165744 011915 043533	.0212873 .0198145 .0276124 0116442	
ppfs0596 Under \$50,000 \$50,000 - \$99,999 \$100,000 - \$249,999 \$250,000 - \$499,999 \$500,000 - \$999,999 \$1,000,000 or more Not sure	0055885 0042069 .0012921 .0071819 .0183146 .0404304 .0303651	.0058013 .0075089 .007404 .0085209 .0091255 .0092405	-0.96 -0.56 0.17 0.84 2.01 4.38 3.15	0.335 0.575 0.861 0.399 0.045 0.000	01696 0189257 013221 0095206 .0004271 .0223173 .0114686	.005783 .0105119 .0158052 .0238844 .0362021 .0585434 .0492616	
1.spousesl 1.kidgkidsl	0072789 0014792	.0074363	-0.98 -0.16	0.328 0.870	0218553 0192479	.0072975	
SL1 Yes 1.slbehind pphhsize	.0061646	.006432 .0142717 .0041308	0.96 -1.67 -1.80	0.338 0.095 0.073	0064433 0517778 0155143	.0187724 .0041722 .0006801	
c.pphhsize#c.pphhsize	.0005107	.0005016	1.02	0.309	0004725	.001494	
ppage	0043356	.0007405	-5.86	0.000	005787	0028842	
c.ppage#c.ppage	.0000339	7.68e-06	4.41	0.000	.0000188	.0000489	
ppgender	 						

Female	.0211348	.0037996	5.56	0.000	.0136869	.0285827
EFUNDstatus 1 3	.0247067	.0066573	3.71 5.06	0.000	.0116573	.0377562
EF3_a Put it on my credit card	.0269588	.0053465	5.04	0.000	.0164787	.037439
EF3_b Put it on my credit card	 0117494 	.006222	-1.89	0.059	0239457	.0004468
$$\rm EF3_c$$ With the money currently	 .0334957 	.004875	6.87	0.000	.02394	.0430515
EF3_d Using money from a bank 1	 0225169 	.0116607	-1.93	0.054	0453739	.00034
EF3_e By borrowing from a frien	 0310982 	.0073044	-4.26	0.000	045416	0167804
EF3_f Using a payday loan, depo	 .0178203 	.015972	1.12	0.265	0134876	.0491282
EF3_g By selling something	 0219278 	.008176	-2.68	0.007	0379541	0059015
EF3_h I wouldn't be able to pay	 0794559 	.0077351	-10.27	0.000	0946181	0642937
\$100 to \$499 \$500 to \$999 \$1,000 to \$1,999 \$2,000 or more 1.EF6C_rent 1.EF6C_ttil 1.EF6C_ttel lastyrccardpay	0119191	.0074372 .0088251 .0094446 .0091928 .0135699 .0112363 .0136936 .012477	9.15 7.82 8.53 11.95 -5.57 -1.06 -0.49 -1.69	0.000 0.000 0.000 0.000 0.000 0.289 0.625 0.090	.0534878 .051718 .0620037 .0918136 1021802 0339441 033534 0455808	.082644 .0863153 .0990299 .1278528 0489816 .010106 .0201497 .0033333
2 3	.0367146	.0126332 .0107619	2.91 1.22	0.004 0.222	.0119512 0079471	.0614779 .0342432
4 lastmnthccardpay	010028 	.0109457	-0.92	0.360	0314835	.0114276

1 2	0092907 0308413	.0083931	-1.11 -2.11	0.268 0.035	0257427 0595119	.0071613 0021707
EF5C Yes	.0052514	.0069101	0.76	0.447	0082937	.0187964
E1_a Yes	0039191	.0080112	-0.49	0.625	0196223	.0117842
E1_b Yes	 0197981 	.0076369	-2.59	0.010	0347677	0048284
E1_c Yes	 0419809 	.0077284	-5.43	0.000	0571298	0268319
E1_d Yes	 0363888 	.0060572	-6.01	0.000	0482619	0245157
E1_e Yes	 0117963 	.008809	-1.34	0.181	0290635	.005471
3 4 5 6	0033292 .001088 0072025 0069547 0171031 .0085748	.009281 .0090968 .0087461 .0088915 .0115231 .0206099	-0.36 0.12 -0.82 -0.78 -1.48 0.42	0.720 0.905 0.410 0.434 0.138 0.677	0215216 0167434 0243463 0243836 0396904 0318241	.0148631 .0189193 .0099413 .0104742 .0054843 .0489738
E2B Yes	0250334	.0058147	-4.31	0.000	0364311	0136356
E4_a Yes	.0026878	.0053426	0.50	0.615	0077847	.0131603
E4_b Yes	 .0165399 	.0065202	2.54	0.011	.0037592	.0293207
E4_c Yes	 0129073	.0061873	-2.09	0.037	0250354	0007792
E4_d Yes	0088464	.0081956	-1.08	0.280	0249111	.0072183
E4_e Yes	.0172928	.0097415	1.78	0.076	0018022	.0363879
E4_f	İ					

Yes	.0318119	.0079039	4.02	0.000	.0163188	.0473049
BK1 Yes	 0396883	.0091241	-4.35	0.000	057573	0218036
BK2_a Yes	.0144163	.0070454	2.05	0.041	.0006061	.0282266
BK2_b Yes	 .0169159 	.008025	2.11	0.035	.0011854	.0326463
BK2_c Yes	0433529	.0123723	-3.50	0.000	0676047	0191011
BK2_d Yes	.0083524 	.0130217	0.64	0.521	0171725	.0338772
BK2_e Yes 1.BK2_fnu _cons	 .0272994 0285221 .7586885	.0165361 .0067927 .0246411	1.65 -4.20 30.79	0.099 0.000 0.000	0051142 0418371 .7103876	.0597131 0152071 .8069893

******CHECKING FOR CONVERGENCE AFTER ITERATION 4

predict B23hatnu4 if B23dum==1
predict B22hatnu4 if B22dum==1

. sum B23hatnu* B22hatnu*

Variable	Obs	Mean	Std. dev.	Min	Max
B23hatnu1 B23hatnu2 B23hatnu3 B23hatnu4 B22hatnu1	4,450 4,450 4,450 4,450 4,450 2,137	.7685084 .7691176 .771163 .7721098 .5849422	.1297283 .1294449 .1294057 .1294489 .1517908	.1601671 .1608215 .1612347 .1613074 .0882174	1.067576 1.067373 1.067664 1.067864 1.002591
B22hatnu2 B22hatnu3 B22hatnu4	2,137 2,137 2,137	.5859481 .5878206 .5885874	.1515989 .151993 .152245	.0893403 .0897391 .0896995	1.002676 1.004185 1.004965

***COMPUTING CHANGE FROM ITERATION 3 TO ITERATION 4 IN ESTIMTED ITERATION WEIGHTS

[.] di .5885874 -.5878206

^{.0007668}

[.] di .7721098 - .771163

^{.0009468}