

**The Impact of the 1990s Economic Boom on Less-Educated Workers in  
Rural America:**

**Did the Rising Tide Lift All Boats?**

by

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## **ABSTRACT**

We use national longitudinal survey data (NLSY79) to investigate the impact of local labor market conditions on the employment and earnings of rural non-college-educated workers. Results suggest that local economic conditions in the late 1990s did have a positive impact on wages, and the effect is larger for workers with no more than a high school degree compared to their college-educated counterparts. We find little evidence of a difference between rural and urban impacts, suggesting that the 1990s boom helped both rural and urban less-educated workers. These results suggest that an expanding economy continues to be a powerful anti-poverty force.

## **Introduction**

The expression “a rising tide lifts all boats” sums up the belief that economic growth raises income for everyone. Indeed, in the 1960s economic growth in the United States was a powerful force for the reduction of poverty. In more recent decades, economists have questioned whether the growing economy has continued to have the same impact on poverty. In particular, in the 1980s researchers found that the relationship between growth and poverty reduction was significantly weakened.

Even though the longest economic expansion in the U.S. has now officially ended with a recession beginning in March 2001, the period of sustained growth in the 1990s provides an unprecedented opportunity to re-examine the impact of local economic conditions on disadvantaged workers. Studies to-date suggest that in metropolitan areas of the U.S., the 1990s boom did help disadvantaged workers by increasing wages and decreasing unemployment. However, few studies have examined the impact of the 1990s expansion on worker outcomes in rural areas.

This research builds upon two main strands of the literature. First, we draw on research that takes advantage of differences in local labor market conditions to examine the impact of overall economic conditions on individual outcomes, particularly for low-income or economically disadvantaged groups. As discussed below, most studies have found that, in metropolitan areas, local economic conditions have a larger impact on less skilled or more disadvantaged workers than on more skilled or older workers. In addition, our research draws upon the literature related to the “wage curve” of Blanchflower and Oswald, who find a negative relation between wages and unemployment rates across countries, regions, and time that is remarkably consistent.

In this paper we find that local economic conditions in the late 1990s did have a positive impact on wages, and that the impact is larger for workers with no more than a high school degree, compared to their more highly educated counterparts. We find little evidence of a difference between rural and urban impacts, suggesting that the 1990s boom helped both rural and urban less-educated workers. These results suggest that an expanding economy continues to be a powerful anti-poverty force. Given the end of the 1990s expansion in recent months, however, policymakers may once again be concerned about the wages and economic outlook for non-college educated workers.

### **Conceptual Framework and Relevant Literature**

Local labor market conditions may impact a worker's earnings and/or labor supply decision by affecting average wages or the likelihood of finding a job.<sup>1</sup> For example, in a job search model, better economic conditions in a labor market are likely to impact the distribution of wage offers. For an individual job seeker, higher employment growth in an area is likely to lead to an increase in the frequency of job offers, raising the probability of employment. It may also improve wage offers, increasing earnings, all else equal (Hoynes, 2000). In a job-queuing model, better local economic conditions may increase wages and employment of disadvantaged workers by both reducing unemployment and increasing upward mobility into higher wage jobs (Bartik 1996).

There are a growing number of studies that take advantage of differences in local labor market conditions to examine the impact of overall conditions on individual outcomes, particularly for low-income or economically disadvantaged groups. Hoynes

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<sup>1</sup> Labor supply is frequently modeled as a joint household decision in the case of married adults. Unfortunately our data do not allow us to model the labor supply decisions of husbands and wives jointly. Tokle and Huffman provide an example of estimation of a joint labor supply model.

(1988) examines the impact of business cycles for different subgroups based on race, gender and education by relying on variations in economic conditions across MSAs (metropolitan statistical areas). She finds that wages of less skilled workers are affected more by economic conditions than those of more skilled workers. Freeman and Rodgers focus on the 1990s expansion and find that the impact of favorable local economic conditions has been greatest for younger men (under age 25) and for African American men. Bartik (1996) and Bound and Holzer also find that employment growth leads to wage increases for younger, less experienced workers in urban areas.

In addition, our research is related to the literature on the “wage curve.” Blanchflower and Oswald find a negative empirical relation between wages and unemployment rates across countries, regions, and time that is remarkably consistent. This inverse relationship between wages and unemployment contrasts with a Harris-Todaro model of regional economies, in which areas with higher unemployment rates have higher wages (a compensating differential, in effect). Blanchflower and Oswald suggest that while this model may hold in the long run, at a point in time the cross-sectional relationship between wages and unemployment is negative. In support of their empirical findings, they present a number of labor contracting and wage efficiency models that could produce such a relationship.

Two recent papers address the “rising tide” question more directly using aggregate data. Hines, Hoynes and Krueger use aggregate measures of labor market outcomes at the MSA-level to estimate the impact of changes in unemployment over the business cycle. They find that employment, wages and hours worked for low-skilled workers increase during expansions and decrease during recessions, though the impact on

wages is fairly small. Freeman (2001) examines the effect on state poverty rates of changes in the unemployment rate and average earnings. He concludes that expansions with low unemployment rates (4-5%) and rising real wages will reduce poverty, though he notes that many people will remain poor due to barriers to labor force participation.

## **Data and Methods**

In this study our primary source of data is the 1979 National Longitudinal Survey of Youth (NLSY79). The Bureau of Labor Statistics began surveying a group of about 12,000 youth aged 14 to 22 in 1979 and has interviewed them annually since then (biannually since 1994). While the sample has undergone some revisions, the retention rate in 1998 for those who remain eligible was 84%. The NLSY79 includes extensive data on demographic and family characteristics, and work history and earnings. Under special agreement with the Bureau of Labor Statistics, we obtained the NLSY79 geocode data, which provides more detailed information on the location of respondents (e.g., county of residence).<sup>2</sup> This information allows us to compare findings using different definitions of “rural.”

One of our main objectives in this study is to investigate the effect of using different geographical units and definitions of “rural” compared with other studies. Recent studies use various definitions of local labor markets. Several use metropolitan statistical areas (MSAs) (Bound and Holzer; Hoynes, 1999; Bartik, 1991, 1996; Freeman and Rogers; Cain and Finnie), while others use state-level data (Tokle and Huffman; Freeman). In analyses of local labor market conditions and welfare spells, Hoynes (2000)

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<sup>2</sup> Use of the confidential geocode data is subject to special agreement with the Bureau of Labor Statistics. Researchers wishing to use these data must apply to BLS directly.

uses counties, and Fitzgerald uses both counties and Labor Market Areas as defined by the USDA Economic Research Service to define local labor markets.

In this study, we use commuting zones as defined by Tolbert and Sizer as the relevant labor market for each individual. The Tolbert and Sizer commuting zones are counties grouped together based on actual commuting patterns found in Census data. The commuting zones typically include several counties and can cross state boundaries. As shown in table 1, Tolbert and Sizer classify commuting zones based on the size of the largest population center. For this study, we define “rural” labor markets as non-metropolitan commuting zones.<sup>3</sup>

The commuting zone approach provides a more realistic approximation of the labor market opportunities faced by an individual. States are generally too large and counties too small to reflect a local labor market. Also, in order to examine the effect in rural areas, areas outside of MSA’s must be included.

One of the advantages of using the NLSY79 data is that it is a panel data set, tracking the same individuals over time, so we can control for unobserved individual effects. In addition, by using the geocode data, we can more accurately identify the type of labor market where the individual resides. However, there are some disadvantages to using the NLSY79 data for this type of study. The first drawback is the limited age range of respondents. By 1998, the respondents were between 33 and 41 years of age. Thus while the respondents are in their prime labor market years, we are unable to examine the impacts of local economic conditions on younger or older workers. Freeman and Rodgers, for example, find significant differences in the impacts on younger (under age

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<sup>3</sup> We also used the NLSY definitions of *smsa* (e.g., living in a metropolitan county) and *rural* (living in a county that is 0-49% urbanized) to compare results. While the samples are not identical, the results are qualitatively similar regardless which definition is used.

25) workers and others. A second drawback of the NLSY79 data for this study is that the number of respondents in the key category of interest, non-college educated workers in rural areas, is fairly small (less than 300).

### **Local Economic Conditions and Local Labor Markets**

A number of different variables have been used to measure labor market conditions: unemployment rates (Freeman and Rodgers; Fitzgerald); predicted employment growth, which is a proxy for labor demand calculated by weighting national sectoral growth rates by local industry sectoral shares (Bound and Holzer); changes in the “wage premium” implied by regional industry mix (Bartik, 1996); and employment growth (Bartik, 1991, 1996). We use two alternative measures: total employment growth (percent change in total employment in the county) and area unemployment rate.

Figure 1 shows how average annual job growth has varied across the different types of commuting zones between 1993 and 1998. While total employment tended to increase faster in the metropolitan commuting zones in 1996 and 1998 than in the non-metropolitan areas, the same is not true earlier. In 1993 and 1994, jobs grew faster in two of the three non-metropolitan categories compared with most of the metropolitan ones. Only in the smallest, most rural commuting zones has job growth consistently been lower than elsewhere.

Figure 2 shows the trends in unemployment rates across commuting zones. Unemployment rates were slightly higher in 1993 and 1994 in the non-metropolitan commuting zones than the metro areas. Unemployment rates trended downwards in the mid to late 1990s in all categories and were similar by 1998 across commuting zone



types. Within commuting zone types, unemployment rates and employment growth rates varied considerably, though on average all improved during the 1990s expansion.

### **Model and Estimation**

In order to estimate the impact of local labor market conditions on wage and employment outcomes, we estimate a reduced form model of the following basic form:

$$Y_i = \beta'X_i + \gamma'LM_i + e_i$$

where  $Y_i$  = the employment outcome for individual  $i$ ;  $X_i$  = a vector of human capital and demographic variables,  $LM_i$  = measures of local labor market conditions in the county, and  $e_i$  is a random error term. This approach is similar to that used by Bartik (1996), Bound and Holzer, and Freeman and Rodgers. We estimate models for three outcome measures: hourly wage, weekly wage, and the probability of employment.<sup>4</sup> Each of the wage equations is estimated using the Heckman sample selection correction technique.<sup>5</sup> All standard errors are estimated using the Huber-White robust method.

Control variables in each model are fairly standard for wage employment equations.<sup>6</sup> Means and standard deviations for all variables are shown in table 2. Socio-demographic variables included are the individual's age, gender, highest grade completed, marital status, race/ethnicity, total work experience (hours) and experience

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<sup>4</sup> We also estimated a multinomial logit model where the outcomes are employed, unemployed and out of the labor force. The results are similar to the model presented here in that factors that increase the probability of employment decrease the probability of unemployment.

<sup>5</sup> We face the common difficulties in specification of the selection equation. We include age, education, race, gender, marital status, number of children, education of parent, work experience and experience squared, south, and the local economic condition variable in the selection equation. Full results are shown in the appendix.

<sup>6</sup> We ran separate estimations for men and women and found that the coefficients on the local labor economic condition variables were quite similar for the wage equations, suggesting that pooling the data does not impact the key results. Not surprisingly, perhaps, the estimated equations on employment were quite different for men and women. In future work we will estimate these models separately for men and women.

squared, length of time in current job (tenure) and tenure squared. In addition, dummy variables are included for union status, major industry and occupational categories, south census region, and rural (defined based on commuting zone category).

Local labor market conditions are measured in two ways: local unemployment rate and change in local employment. We test whether the impact of local labor market conditions is different in rural versus urban areas by including an interaction term between the rural dummy and either the unemployment rate or the employment growth rate.

## **Results**

In this section we first examine the impact of local economic conditions on less educated versus more educated workers, and compare these results to other studies. For the purposes of this study we define “less educated” to include those workers with no more than a high school degree. We then examine the impact in rural versus urban areas to answer the question whether the “rising tide” is indeed helping less educated workers in rural areas. We estimate the models two ways, first using cross-sectional data from 1998 and secondly as a panel data set from 1993-1998 with a fixed effects model. The key results are summarized in tables 3 and 4.<sup>7</sup>

Looking first at the cross-sectional results, it is clear that better local economic conditions in 1998 improved wage outcomes for workers with a high school education or less. For these non-college educated workers, a one percent reduction in the unemployment rate is associated with about a 0.1 percent increase in hourly or weekly wages (both unemployment rate and wage variables are in natural log form) (see table

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<sup>7</sup> Complete results for the cross-sectional models are shown in the appendix, and the fixed effects results are available upon request from the authors.

3A). This estimate is very similar to the standard “wage curve” (Blanchflower and Oswald). Employment growth also exerts a significant impact on wages, raising wages by 2-3 percent for less educated workers (see table 4A). The impact of local economic conditions is weaker for workers with more than a high school education. While the estimated effects on hourly and weekly wages are significant for employment growth, they are somewhat smaller than for less educated workers. The impact of unemployment rate changes is not statistically significant at the 5 percent level for workers with more than a high school degree.

Using the 1998 data, we find that local economic conditions did not have a statistically significant impact on the likelihood of being employed for either group of workers. The estimated coefficients for the unemployment rate is negative (although not significant). Somewhat surprisingly, the estimated coefficients for employment growth are also negative (though not significant).

One of the concerns with cross-sectional estimates of wages is that unobserved characteristics of an individual may bias the results. Therefore we next used the 1993-98 NLSY79 data to estimate a fixed effects model.<sup>8</sup> Again we find that the local unemployment rate has a significant inverse relationship with hourly and weekly wages for workers with less than a high school education (see table 3B). The estimated coefficient is about half the size of the cross-sectional estimate. Unlike the 1998 cross-section result, the estimated impact of unemployment rate on the probability of employment is statistically significant using the 1993-98 panel. The estimated impacts of

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<sup>8</sup> We also estimated a random effects model but found that the Hausman test rejected the random effects assumption.

unemployment rate on wages or employment for workers with more than a high school education are not statistically significant in the fixed effects model.

The fixed effects model results using employment growth as the measure of local economic conditions differ considerably from the cross sectional results. In the fixed effects model, the estimated coefficients on employment growth are not significant for either education group, for either weekly or hourly wages (see table 4B). Employment growth does have a significant positive effect on the likelihood of employment for less educated workers, however.

We used two methods to test whether the impact of local labor market conditions differed in rural versus urban areas: we estimated separate models for rural and urban residents, and we estimated models using the pooled urban and rural data including an interaction term between rural residence and the local economic variable. In both cases, the results were similar. We found no statistically significant differences between rural and urban areas in terms of the impact of either unemployment rate or employment growth on wages or employment status. Finding no effect may be due to small sample sizes, or may in fact indicate that the labor market impacts do not differ on average in rural areas.

## **Conclusions**

The sustained economic expansion of the 1990s in the United States appears to have helped “to lift all boats,” by improving the wages and employment of non-college educated workers. Better local labor market conditions have a stronger impact on outcomes for less educated workers than for those with more than a high school

education. Using the NLSY79 data, this study confirms findings from studies using alternative data sets about the impacts of local labor market conditions. In addition, unlike other studies, we investigate whether this impact holds true in rural labor markets as well as metropolitan areas. We find that the evidence suggests that, in general, the impact of local labor market conditions on weekly wages is similar for rural and urban workers. The impact on employment status is less clear and suggests the need for further work with better data for rural areas. In addition, improved labor market conditions will primarily help those who are able to participate in the labor market. Poor families facing barriers to labor force participation or unable to work full-time year round struggle despite a growing economy.

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**Table 1**  
**Categorization of Commuting Zones and Sample Size**

<u>Category</u>	<u>Type of Commuting Zone</u>	<u>1998 sample size</u>	<u>Percent of sample</u>
1	Non-metro, small towns	108	1.3
2	Non-metro, small urban center	431	5.3
3	Non-metro, large urban center	327	4.0
4	Metro, small	974	11.9
5	Metro, medium	2315	28.2
6	Metro, major	4044	49.3
	TOTAL	8199	

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Note: Commuting Zone designations from Tolbert Charles M. and Molly Sizer, 1996, "U.S. Commuting Zones and Labor Market Areas: a 1990 Update." U.S. Department of Agriculture Economic Research Service Staff Paper No. 9614. September. Categorization is based on the size of the largest population center in the commuting zone.



**Table 2**  
**Variable Definitions and Descriptive Statistics**

	<b>Abbreviated Name</b>	<b>Variable Description</b>	<b>Mean</b>	<b>Std. Dev.</b>
<b>1998 Outcomes</b>				
Hourly Wage	hrwg	If employed in 1998, hourly wage for the respondent's main job.	15.3506	16.5954
Weekly Wage	wkwg	If employed in 1998, weekly wage for the respondent's main job.	621.4267	508.8708
Employed	empl	Indicates that the worker is employed at the time of the 1998 interview.	0.7991	0.4007
Out of the Labor Force	olf	Indicates that the worker is out of the labor force at the time of the 1998 interview.	0.1505	0.3576
Unemployed	unmp	Indicates that the worker is unemployed at the time of the 1998 interview.	0.0394	0.1946
<b>Personal Characteristics</b>				
Age	age79	Age of the respondent in 1998.	36.8982	2.3056
Years of Education	hgcslf	Highest grade completed by the respondent as of 1998.	13.0464	2.4628
Hispanic	hispanic	Indicates that the respondent is Hispanic.	0.1578	0.3646
Black	black	Indicates that the respondent is black.	0.2502	0.4331
Female	female	Indicates that the respondent is female.	0.4953	0.5000
Married	marrd	Indicates that the respondent is married at the time of the 1998 interview.	0.5570	0.4968
Years of Education for Parent	hgcparrt	Highest grade completed by the respondent's head parent.	10.8108	3.8788
Number of Children	children	Number of children the respondent has at the time of the 1998 interview.	2.0284	1.5465
South	so	Indicates that the respondent resides in the South at the time of the 1998 interview.	0.4057	0.4910
Rural	newrural	Indicates that the respondent resides in a rural area at the time of the 1998 interview. Rural is defined as one of the following: Non-Metro Small Town, Non-Metro Small Urban, or Non-Metro Larger Urban. Urban then defined as one of the following: Small Metro, Medium Metro, or Major Metro.	0.1056	0.3074
<b>Experience and Tenure</b>				
Total Work Experience (hours)	tothr	Hours of work experience the respondent has accumulated as of the 1998 interview.	23520.3900	14406.8700
Job Tenure (weeks)	tenure	If employed in 1998, weeks of tenure the respondent has with his/her main employer.	290.9466	275.5144

**Table 2, continued**

<b>Job Characteristics</b>				
Union Status	union	If employed in 1998, union status for the respondent's main job.	0.0510	0.2200
Industry Category		If employed in 1998, industry category for the respondent's main job.		
	indbc1	Non-Service	0.3452	0.4755
	indbc2	Professional Service	0.3037	0.4599
	indbc3	Retail Service	0.1262	0.3322
	indbc4	Commercial Service	0.2248	0.4175
Occupation Category		If employed in 1998, occupation category for the respondent's main job.		
	occbc1	Trade	0.2386	0.4263
	occbc2	Lower-Skilled Business	0.1579	0.3647
	occbc3	Higher-Skilled Business	0.1956	0.3967
	occbc4	Lower-Skilled Service	0.2013	0.4010
	occbc5	Professional Service	0.2066	0.4049
<b>Local Economic Conditions</b>				
Employment Growth	total	Employment growth from 1997 to 1998 associated with the county the respondent resides in at the time of the 1998 interview.	0.0252	0.0184
Unemployment Rate	lnurate	Natural log of the March 1998 unemployment rate associated with the county the respondent resides in at the time of the 1998 interview.	5.1548	0.0277

Descriptive statistics are for 1998.

**Table 3**  
**Summary of Results: Impact of Local Unemployment Rate**  
**on Wage and Employment Outcomes**

	High School or Less		More Than High School	
	Ln(Unemployment Rate)	Rural Interaction Term	Ln(Unemployment Rate)	Rural Interaction Term
<b>Table 3A: Cross-Section 1998</b>				
Ln(Hourly Wage)	-0.1130 ** (0.0235)	0.0846 (0.0543)	-0.0442 (0.0237)	0.0446 (0.0951)
Ln(Weekly Wage)	-0.1182 ** (0.0269)	0.0196 (0.0817)	-0.0399 (0.0312)	-0.0655 (0.1241)
Employment	-.0435 (0.1111)	-0.5318 (0.3658)	-0.0979 (0.1398)	0.5710 (0.5726)
<b>Table 3B: Panel Data 1993-1998</b>				
Ln(Hourly Wage)	-0.0666 ** (0.0205)	0.0523 (0.0609)	-0.0014 (0.0181)	-0.0919 (0.0714)
Ln(Weekly Wage)	-0.0461 * (0.0230)	0.0357 (0.0683)	0.0203 (-0.0214)	-0.1004 (0.0844)
Employment	-0.0236 * (0.0119)	-0.0666 (0.0394)	0.0047 (0.0110)	-0.0144 (0.0453)

**Table 4**  
**Summary of Results: Impact of Local Employment Growth**  
**on Wage and Employment Outcomes**

	High School or Less		More Than High School	
	Annual Job Growth	Rural Interaction Term	Annual Job Growth	Rural Interaction Term
<b>Table 4A: Cross-Section 1998</b>				
Ln(Hourly Wage)	2.8541 ** (0.5480)	-0.7258 (1.5992)	2.1005 ** (0.6738)	-3.1187 (2.0221)
Ln(Weekly Wage)	3.0854 ** (0.5952)	2.0482 (1.9716)	1.9949 * (0.8338)	-1.7194 (2.5876)
Employment	-4.8521 (2.5765)	7.3376 (6.3915)	-4.1335 (3.8682)	-9.7994 (11.1990)
<b>Table 4B: Panel Data 1993-1998</b>				
Ln(Hourly Wage)	-0.3302 (0.2862)	-0.5457 (0.6482)	0.1350 (0.2927)	-0.2254 (0.7389)
Ln(Weekly Wage)	-0.2690 (0.3203)	-1.0486 (0.7254)	0.2468 (0.3464)	0.8859 (0.8747)
Employment	0.3792 * (0.1680)	0.0451 (0.3774)	0.0666 (0.1757)	-0.3471 (0.4271)

Figure 1

Annual Job Growth - By Type of Commuting Zone

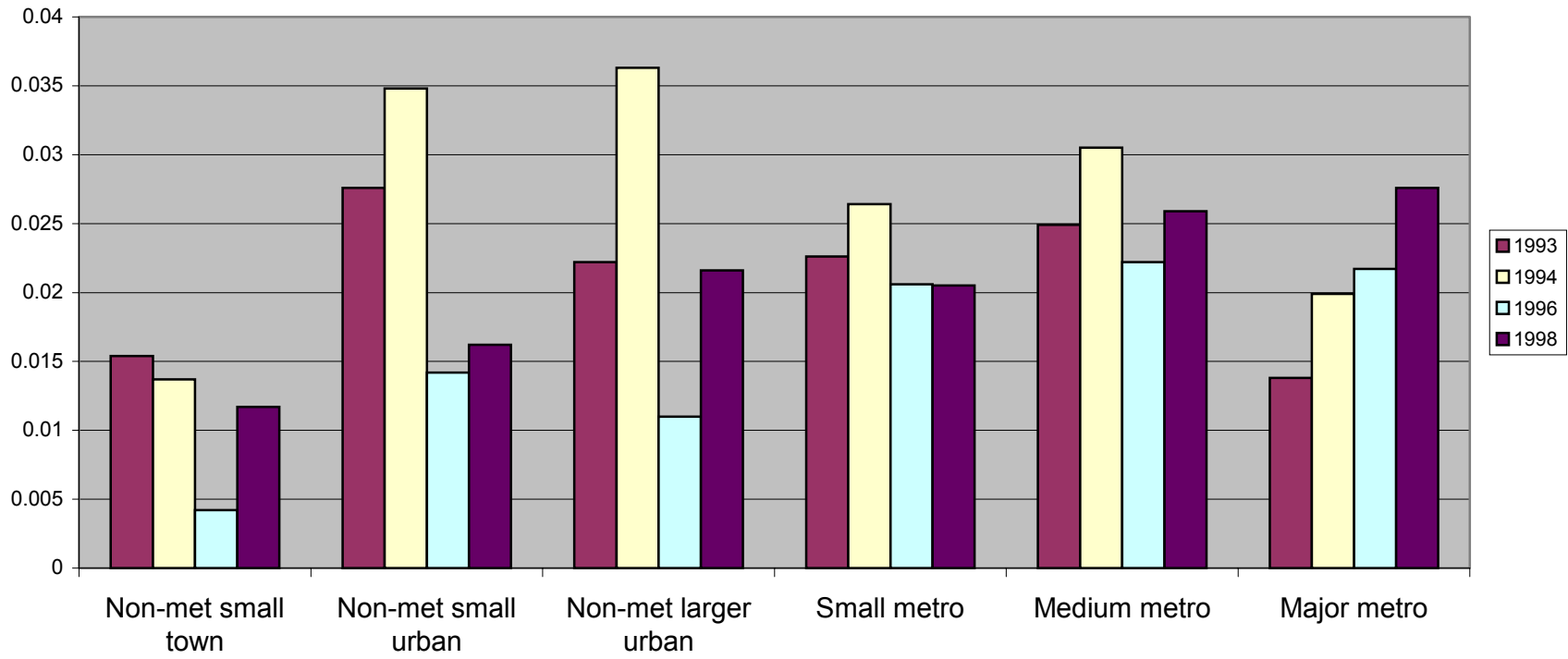
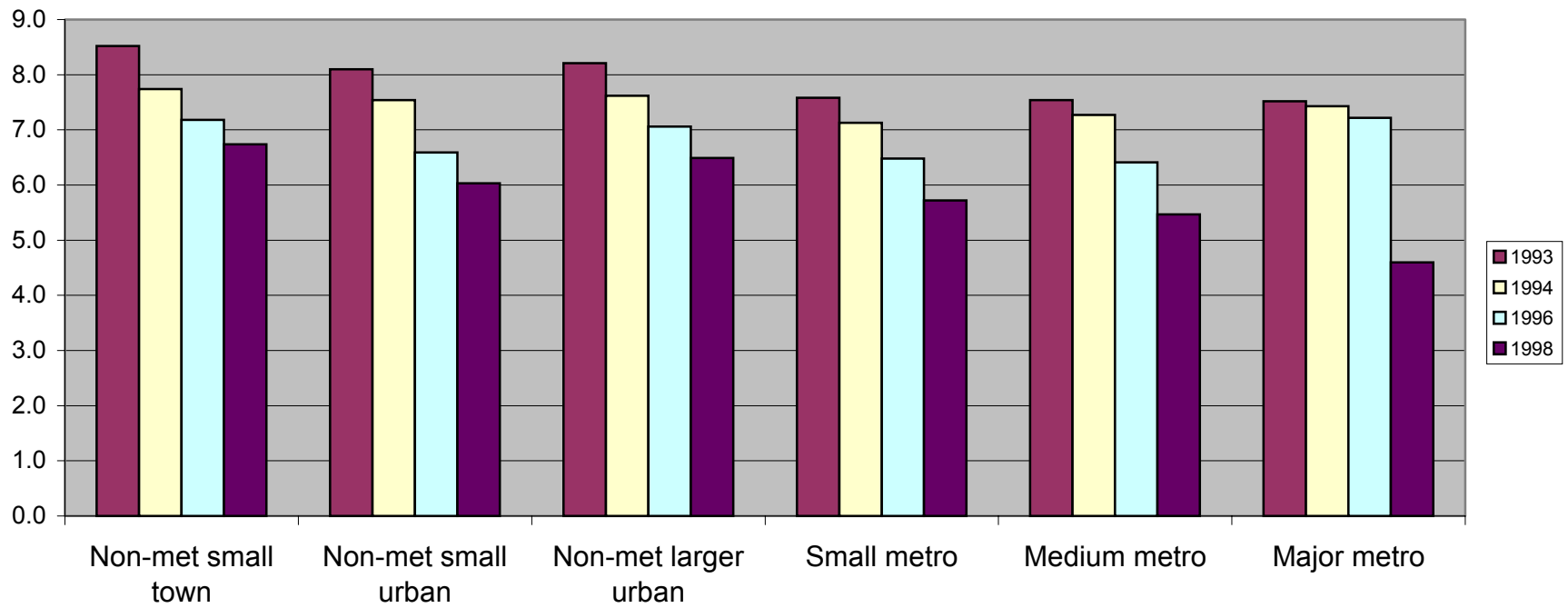


Figure 2

**Average Unemployment Rate By Type  
of Commuting Zone**



**Table A.1: 1998 Hourly Wage Estimation With Local Unemployment Rate – High School Education or Less**

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Heckman selection model                               Number of obs   =    3546
(regression model with sample selection)             Censored obs   =    1005
                                                    Uncensored obs =    2541

                                                    Wald chi2(21)  =    434.64
Log likelihood = -3265.757                          Prob > chi2    =    0.0000
    
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
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lnhrwg						
age79	-.0032116	.0041414	-0.775	0.438	-.0113286	.0049055
hgcslf	.0399347	.0087568	4.560	0.000	.0227718	.0570977
hispanic	.0117269	.0308861	0.380	0.704	-.0488087	.0722625
black	-.0620607	.0252616	-2.457	0.014	-.1115724	-.012549
female	-.1023033	.0254831	-4.015	0.000	-.1522494	-.0523573
marrd	.0284369	.0194959	1.459	0.145	-.0097744	.0666482
so	-.0988534	.0214922	-4.600	0.000	-.1409773	-.0567295
union	-.0179135	.0526235	-0.340	0.734	-.1210537	.0852268
indbc1	.0334642	.0290664	1.151	0.250	-.0235049	.0904332
indbc2	-.0428138	.0381695	-1.122	0.262	-.1176246	.0319971
indbc3	-.1528852	.0328403	-4.655	0.000	-.217251	-.0885195
occbc1	-.1583059	.0405208	-3.907	0.000	-.2377253	-.0788865
occbc2	-.1286459	.0376436	-3.417	0.001	-.202426	-.0548657
occbc3	.0097676	.0469066	0.208	0.835	-.0821675	.1017028
occbc4	-.3153684	.0411119	-7.671	0.000	-.3959462	-.2347906
tenure	.0006072	.0001062	5.718	0.000	.0003991	.0008153
tenure2	-4.13e-07	1.08e-07	-3.841	0.000	-6.24e-07	-2.02e-07
tothr	.0000176	4.31e-06	4.078	0.000	9.12e-06	.000026
tothr2	-1.19e-10	7.12e-11	-1.673	0.094	-2.59e-10	2.04e-11
newrural	-.2854531	.0955027	-2.989	0.003	-.4726349	-.0982713
lnurate	-.1130204	.023458	-4.818	0.000	-.1589973	-.0670436
nrlurate	.0845757	.0543379	1.556	0.120	-.0219246	.191076
_cons	6.426452	.150074	42.822	0.000	6.132312	6.720592
-----						
wageobs						
age79	-.0284001	.0097146	-2.923	0.003	-.0474405	-.0093598
hgcslf	.0296838	.0166058	1.788	0.074	-.0028631	.0622306
hispanic	.0025449	.0641634	0.040	0.968	-.1232131	.1283029
black	.067915	.056518	1.202	0.229	-.0428583	.1786882
female	.1768272	.0474558	3.726	0.000	.0838155	.2698389
marrd	.1988318	.0457379	4.347	0.000	.1091871	.2884765
so	.1072951	.0468797	2.289	0.022	.0154126	.1991777
newrural	.6971023	.2998196	2.325	0.020	.1094666	1.284738
lnurate	-.0599792	.0558314	-1.074	0.283	-.1694068	.0494485
nrlurate	-.3062882	.1643737	-1.863	0.062	-.6284547	.0158783
hgcpant	-.0067575	.006767	-0.999	0.318	-.0200206	.0065056
children	.0078364	.0140644	0.557	0.577	-.0197294	.0354022
tothr	.0000878	6.93e-06	12.666	0.000	.0000742	.0001013
tothr2	-6.02e-10	1.45e-10	-4.142	0.000	-8.87e-10	-3.17e-10
_cons	-1.049633	.2698891	-3.889	0.000	-1.578606	-.5206606
-----						
/athrho	.0481632	.0362872	1.327	0.184	-.0229583	.1192848
/lnsigma	-.7572299	.06248	-12.120	0.000	-.8796884	-.6347714
-----						
rho	.048126	.0362031			-.0229543	.1187222
sigma	.4689637	.0293008			.4149122	.5300566
lambda	.0225694	.0168387			-.0104339	.0555727
-----						

Wald test of indep. eqns. (rho = 0): chi2(1) = 1.76 Prob > chi2 = 0.1844

**Table A.2: 1998 Weekly Wage Estimation With Local Unemployment Rate – High School Education or Less**

Heckman selection model  
(regression model with sample selection)

Number of obs = 3546  
Censored obs = 1005  
Uncensored obs = 2541

Wald chi2(21) = 483.04  
Prob > chi2 = 0.0000

Log likelihood = -3713.324

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
lnwkwg						
age79	-.0165649	.0051561	-3.213	0.001	-.0266707	-.0064592
hgcs1f	.0423919	.0100579	4.215	0.000	.0226788	.0621049
hispanic	.0433566	.0327889	1.322	0.186	-.0209085	.1076217
black	-.0055492	.0291054	-0.191	0.849	-.0625948	.0514964
female	-.1483837	.0306611	-4.839	0.000	-.2084784	-.0882889
marrd	.0009578	.0227744	0.042	0.966	-.0436791	.0455947
so	-.0827034	.0256274	-3.227	0.001	-.1329322	-.0324745
union	.0499202	.0635247	0.786	0.432	-.0745859	.1744264
indbc1	.0789229	.0323944	2.436	0.015	.0154311	.1424147
indbc2	-.0781284	.0458268	-1.705	0.088	-.1679472	.0116904
indbc3	-.1878674	.0376676	-4.988	0.000	-.2616945	-.1140403
occbc1	-.1312854	.0495055	-2.652	0.008	-.2283143	-.0342565
occbc2	-.1451947	.0459813	-3.158	0.002	-.2353165	-.055073
occbc3	.0543836	.0550616	0.988	0.323	-.0535352	.1623024
occbc4	-.394743	.0514919	-7.666	0.000	-.4956654	-.2938207
tenure	.000616	.0001216	5.065	0.000	.0003776	.0008545
tenure2	-4.42e-07	1.23e-07	-3.596	0.000	-6.83e-07	-2.01e-07
tothr	.0000223	4.77e-06	4.673	0.000	.0000129	.0000317
tothr2	-5.86e-11	7.50e-11	-0.781	0.435	-2.06e-10	8.84e-11
newrural	-.1869535	.1360517	-1.374	0.169	-.45361	.0797031
lnurate	-.1182485	.0269484	-4.388	0.000	-.1710663	-.0654306
nrlurate	.0196388	.0817732	0.240	0.810	-.1406336	.1799113
_cons	10.09611	.1765426	57.188	0.000	9.75009	10.44212
wageobs						
age79	-.028383	.0097138	-2.922	0.003	-.0474217	-.0093442
hgcs1f	.0294054	.0165954	1.772	0.076	-.0031209	.0619318
hispanic	.0030406	.0642275	0.047	0.962	-.1228429	.1289242
black	.0675499	.0566369	1.193	0.233	-.0434564	.1785563
female	.1778179	.0474051	3.751	0.000	.0849055	.2707303
marrd	.1984602	.0457723	4.336	0.000	.1087482	.2881722
so	.1084716	.0468613	2.315	0.021	.0166252	.2003181
newrural	.699225	.2995782	2.334	0.020	.1120626	1.286387
lnurate	-.0590462	.0557801	-1.059	0.290	-.1683732	.0502808
nrlurate	-.3075786	.1642827	-1.872	0.061	-.6295668	.0144096
hgcparrt	-.006452	.0067556	-0.955	0.340	-.0196927	.0067887
children	.0084394	.0142032	0.594	0.552	-.0193984	.0362772
tothr	.0000876	6.97e-06	12.576	0.000	.000074	.0001013
tothr2	-5.99e-10	1.46e-10	-4.094	0.000	-8.86e-10	-3.12e-10
_cons	-1.05248	.2703205	-3.893	0.000	-1.582298	-.5226611
/athrho	.0142742	.0260695	0.548	0.584	-.0368212	.0653695
/lnsigma	-.5815187	.0486971	-11.942	0.000	-.6769632	-.4860742
rho	.0142732	.0260642			-.0368045	.0652765
sigma	.5590487	.027224			.5081578	.6150362
lambda	.0079794	.0145125			-.0204646	.0364234
Wald test of indep. eqns. (rho = 0): chi2(1) = 0.30 Prob > chi2 = 0.5840						





**Table A.4: 1998 Hourly Wage Estimation With Local Unemployment Rate – More Than High School Education**

Heckman selection model  
(regression model with sample selection)

Number of obs = 2623  
Censored obs = 464  
Uncensored obs = 2159

Log likelihood = -2406.127

Wald chi2(21) = 1055.22  
Prob > chi2 = 0.0000

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
lnhrwg						
age79	-.003278	.004657	-0.704	0.481	-.0124055	.0058495
hgcslf	.0746036	.0069005	10.811	0.000	.0610788	.0881284
hispanic	.0115946	.0307419	0.377	0.706	-.0486583	.0718476
black	-.0505677	.025415	-1.990	0.047	-.1003803	-.0007552
female	-.141477	.0260566	-5.430	0.000	-.192547	-.090407
marrd	.0890568	.0225463	3.950	0.000	.0448669	.1332468
so	-.0448759	.0219863	-2.041	0.041	-.0879681	-.0017836
union	.0138956	.0406387	0.342	0.732	-.0657548	.0935459
indbc1	-.0623271	.0283327	-2.200	0.028	-.1178581	-.006796
indbc2	-.2048047	.0283809	-7.216	0.000	-.2604303	-.1491792
indbc3	-.3755569	.0418602	-8.972	0.000	-.4576014	-.2935123
occbc1	-.2962722	.0349711	-8.472	0.000	-.3648144	-.2277301
occbc2	-.3090146	.032733	-9.440	0.000	-.37317	-.2448591
occbc3	.0113783	.0270073	0.421	0.674	-.0415551	.0643118
occbc4	-.421747	.0444355	-9.491	0.000	-.5088389	-.334655
tenure	.0005646	.0001267	4.455	0.000	.0003162	.000813
tenure2	-3.61e-07	1.36e-07	-2.649	0.008	-6.29e-07	-9.40e-08
tothr	.0000233	4.77e-06	4.879	0.000	.0000139	.0000326
tothr2	-1.76e-10	7.34e-11	-2.402	0.016	-3.20e-10	-3.24e-11
newrural	-.2801946	.1712129	-1.637	0.102	-.6157657	.0553765
lnurate	-.0442	.0237119	-1.864	0.062	-.0906745	.0022745
nrlurate	.0446	.0951201	0.469	0.639	-.1418319	.2310319
_cons	5.958565	.1586367	37.561	0.000	5.647643	6.269488
-----						
wageobs						
age79	-.0474519	.0124887	-3.800	0.000	-.0719292	-.0229746
hgcslf	.0594139	.0163717	3.629	0.000	.0273259	.0915018
hispanic	-.089979	.0845531	-1.064	0.287	-.2557	.0757419
black	.1409946	.0750637	1.878	0.060	-.0061276	.2881167
female	-.4103127	.061392	-6.683	0.000	-.5306389	-.2899866
marrd	-.0468879	.0633527	-0.740	0.459	-.1710569	.0772811
so	.0753175	.0565224	1.333	0.183	-.0354644	.1860993
newrural	-.4549338	.4747455	-0.958	0.338	-1.385418	.4755503
lnurate	-.0826528	.063155	-1.309	0.191	-.2064344	.0411287
nrlurate	.3466509	.2724349	1.272	0.203	-.1873118	.8806135
hgcparrt	-.035238	.0079059	-4.457	0.000	-.0507331	-.0197428
children	-.0391356	.0205149	-1.908	0.056	-.079344	.0010729
tothr	.0000789	.0000137	5.740	0.000	.000052	.0001058
tothr2	-3.20e-10	2.80e-10	-1.141	0.254	-8.69e-10	2.30e-10
_cons	-.1307772	.3984428	-0.328	0.743	-.9117108	.6501564
-----						
/athrho	.121973	.0503845	2.421	0.015	.0232211	.2207249
/lnsigma	-.7316999	.0416312	-17.576	0.000	-.8132955	-.6501042
-----						
rho	.1213717	.0496423			.0232169	.2172088
sigma	.4810905	.0200284			.4433945	.5219914
lambda	.0583908	.0232907			.0127419	.1040396
-----						
Wald test of indep. eqns. (rho = 0):	chi2(1) =	5.86	Prob > chi2 =	0.0155		
-----						

**Table A.5: 1998 Weekly Wage Estimation With Local Unemployment Rate – More Than High School Education**

Heckman selection model  
(regression model with sample selection)

Number of obs = 2623  
Censored obs = 464  
Uncensored obs = 2159

Wald chi2(21) = 1057.26  
Prob > chi2 = 0.0000

Log likelihood = -2863.122

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
lnwkwg						
age79	-.016645	.0059231	-2.810	0.005	-.028254	-.0050359
hgcslf	.0782242	.0087204	8.970	0.000	.0611325	.0953158
hispanic	.0265021	.0366793	0.723	0.470	-.045388	.0983923
black	.0138561	.0318003	0.436	0.663	-.0484715	.0761836
female	-.2832273	.0310051	-9.135	0.000	-.3439962	-.2224583
marrd	.0146067	.0280975	0.520	0.603	-.0404634	.0696768
so	.0035801	.026987	0.133	0.894	-.0493135	.0564736
union	.1114947	.0500679	2.227	0.026	.0133633	.209626
indbc1	-.0394343	.033515	-1.177	0.239	-.1051224	.0262539
indbc2	-.2569128	.0355916	-7.218	0.000	-.326671	-.1871546
indbc3	-.4295369	.0496005	-8.660	0.000	-.5267521	-.3323216
occbc1	-.2734558	.0452227	-6.047	0.000	-.3620907	-.184821
occbc2	-.2840817	.0409974	-6.929	0.000	-.3644351	-.2037282
occbc3	.0917785	.0321893	2.851	0.004	.0286887	.1548683
occbc4	-.5363156	.0606841	-8.838	0.000	-.6552542	-.4173769
tenure	.0007194	.0001516	4.746	0.000	.0004223	.0010164
tenure2	-5.20e-07	1.62e-07	-3.215	0.001	-8.38e-07	-2.03e-07
tothr	.0000311	6.42e-06	4.849	0.000	.0000185	.0000437
tothr2	-1.31e-10	9.12e-11	-1.433	0.152	-3.09e-10	4.80e-11
newrural	-.0775503	.2201435	-0.352	0.725	-.5090237	.3539231
lnurate	-.0399386	.0312427	-1.278	0.201	-.1011731	.0212959
nrlurate	-.0655251	.1241189	-0.528	0.598	-.3087936	.1777435
_cons	9.544031	.2146691	44.459	0.000	9.123288	9.964775
-----						
wageobs						
age79	-.0475314	.0125072	-3.800	0.000	-.0720451	-.0230178
hgcslf	.0572595	.0162471	3.524	0.000	.0254159	.0891032
hispanic	-.0789846	.084584	-0.934	0.350	-.2447663	.086797
black	.1535051	.075195	2.041	0.041	.0061256	.3008845
female	-.4109185	.0617366	-6.656	0.000	-.5319201	-.2899169
marrd	-.044109	.0633131	-0.697	0.486	-.1682004	.0799823
so	.0780984	.0567403	1.376	0.169	-.0331104	.1893073
newrural	-.4407172	.475791	-0.926	0.354	-1.373251	.4918161
lnurate	-.0798842	.0634085	-1.260	0.208	-.2041626	.0443942
nrlurate	.3370651	.2730267	1.235	0.217	-.1980573	.8721876
hgcparrt	-.0335657	.0077711	-4.319	0.000	-.0487969	-.0183346
children	-.039284	.0206007	-1.907	0.057	-.0796607	.0010926
tothr	.0000784	.0000138	5.661	0.000	.0000512	.0001055
tothr2	-3.08e-10	2.83e-10	-1.090	0.276	-8.62e-10	2.46e-10
_cons	-.1236542	.3980171	-0.311	0.756	-.9037535	.656445
-----						
/athrho	-.0076302	.0336773	-0.227	0.821	-.0736364	.058376
/lnsigma	-.5222819	.0319841	-16.329	0.000	-.5849696	-.4595942
-----						
rho	-.0076301	.0336753			-.0735036	.0583098
sigma	.5931655	.0189719			.5571228	.6315399
lambda	-.0045259	.0199686			-.0436637	.0346119
-----						
Wald test of indep. eqns. (rho = 0):	chi2(1) =	0.05	Prob > chi2 =	0.8208		
-----						

**Table A.6: 1998 Employment Estimation With Local Unemployment Rate – More Than High School Education**

Logit estimates  
 Log likelihood = -1034.6596  
 Number of obs = 3407  
 Wald chi2(13) = 90.10  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.2369

empstat2	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
age79	-.0683365	.0255739	-2.672	0.008	-.1184604	-.0182125
hgcs1f	.1221394	.0357527	3.416	0.001	.0520655	.1922133
hispanic	-.1711292	.1805652	-0.948	0.343	-.5250305	.1827722
black	.2589687	.1538562	1.683	0.092	-.0425839	.5605214
female	-.7507444	.1402474	-5.353	0.000	-1.025624	-.4758645
marrd	-.112712	.1313885	-0.858	0.391	-.3702288	.1448048
so	-.0328751	.1251301	-0.263	0.793	-.2781257	.2123754
newrural	-.7151046	1.021848	-0.700	0.484	-2.71789	1.287681
lnurate	-.0978523	.1397925	-0.700	0.484	-.3718405	.1761359
nrlurate	.5709972	.5725789	0.997	0.319	-.5512368	1.693231
hgcparrnt	-.0655911	.0171563	-3.823	0.000	-.0992168	-.0319655
children	-.0543416	.0414384	-1.311	0.190	-.1355594	.0268762
tothr	.0001014	.0000281	3.606	0.000	.0000463	.0001565
tothr2	2.58e-10	6.20e-10	0.416	0.677	-9.58e-10	1.47e-09
_cons	-.0975307	.8237348	-0.118	0.906	-1.712021	1.51696