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**STATE POLICY EFFECTS ON URBAN PERFORMANCE**

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## **Introduction**

Cities are creatures of their state governments. As such, state policy can have important effects, intended or otherwise, on the well-being of cities and their residents. States affect cities in a variety of ways, and the importance to local governments of the state government role has long been recognized. States determine the institutional forms of their local governments, the land use and regulatory frameworks under which they operate, and the revenue systems they may use. Cities and their residents are directly affected by state tax policies and by state programs and policies. While some states have explicit “urban policies” directed at promoting the well-being of their cities, in every state cities are affected, adversely or beneficially, by a range of state activities not necessarily devised with cities and their residents in mind.

This paper describes the initial stages in a research project that attempts to tease out how state policy effects the performance of cities. We first use factor analysis to explore the performance of central cities between 1990 and 2000 by measuring changes in a set of measures meant to broadly capture the economic and social well-being of city residents. We then employ linear regression to predict cities’ factor scores using a set of non-policy variables that describe the demographic characteristics and economic structures of the cities. By explaining performance using non-policy variables in this way, we attempt to isolate the potential impact of policy on performance, which should be captured in the unexplained variation in the cities’ factor scores. In the final part of our statistical analysis, we estimate how much of the unexplained variation can be attributed to state-level factors by using state fixed-effects models to predict the residuals from the previous stage regressions.

The next stage in our research will be to use the state-fixed effects models as guides in choosing states for case study research. We will conduct a series of intensive case studies in

both well- and poorly-performing states in order to determine how and why state-level policy affects the well-being of cities within those states.

## **Methods and Results**

### *Factor Analysis*

The purpose of our statistical analysis is to determine how much of the variation in city performance could potentially be due to differences in state policies. We therefore must first measure urban performance, broadly construed as the change in a broad array of economic and social indicators of the well-being of city residents. We use factor analysis to determine how these indicators vary in relation to one another and find that city performance can be evaluated in terms of categories of indicators (factors). Within these categories, the indicators are highly related to one another, but the categories themselves are statistically and intuitively distinct from one another.

Our population of cities includes all central cities with populations of more than 50,000 in 1990 (n=325).<sup>1</sup> We collected data on 27 indicators of the economic and social well-being of city residents. These include measures of income (per capita and median household), educational attainment, crime rates (both murder and larceny rates), housing costs (median home values and rents as well as housing affordability), racial and economic segregation, poverty (rates and concentration), and employment. (See Table 1 for a list of indicators and their definitions.) We are interested in the *change in* city performance, so all the indicators measure change from 1990 to 2000.

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<sup>1</sup> There was at least one central city of this size in every state except Vermont.

Factor analysis is used in order to condense our indicators of city and residential well-being into categories representing broad areas of performance. Four factors were retained, with eigenvalues ranging from 5.66 to 1.43. (See Table 2 for rotated factor loadings and eigenvalues.)

The four factors and their highly-loading variables<sup>2</sup> are:

Factor 1: Income and Education

- Percent with some college
- Median household income
- Per capita income
- Poverty rate
- Labor force participation rate
- Median rent
- Median home value

Factor 2: Population and Employment

- Population
- Jobs by place of residence
- Jobs by place of work

Factor 3: Housing Affordability

- Percent of households spending 30% or more of income on housing
- Percent of households spending 50% or more of income on housing

Factor 4: Concentrated Poverty

- Poverty rate
- Percent of all residents living in high-poverty neighborhoods

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<sup>2</sup> Variables considered highly-loading are those with factor loadings of at least  $\pm 0.55$ .

- Percent of poor residents living in high-poverty neighborhoods

The factor that explained the largest proportion of the variance among the indicators (33%) was that most related to income and education. The four factors cumulatively explained more than 77% of the total variance (see Table 2). We created new variables representing city performance in each of the preceding categories by using the factor scores for each factor. (See Table 3 for city rankings on each factor.)

### *Predicting the Factor Scores*

We then estimate how much of the variation in the categories of urban performance can be attributed to public policy by using linear regression to explain the portion of the variation that *cannot* be attributed to policy (or at least recent policy). A set of non-policy variables measuring the social and economic structures of cities is used to predict performance in each of the four categories. We then assume that the variation in performance that was not explained by the non-policy variables is the maximum portion of the variation that *could* be due to differences in policy.

Our predictive models of performance used the factors scores as dependent variables, and a set of variables measuring changes in cities' social and economic structures were the independent variables. The economic structure variables include the percent of the labor force that is working in the manufacturing industry as well as the percent in the finance, real estate, and insurance industries. Social structure and demographic variables include the percent of residents who are of working age, the percent who are enrolled in four-year colleges and universities, and the percent who are black (non-Hispanic) or Hispanic. (See Table 4 for a list of

independent variables.) We control for city characteristics at the beginning of the study period by including measures of these characteristics in 1990 in addition to the change from 1990 to 2000. Also included are controls for cities' climate (average July temperature), for whether or not the cities are state capitals, and for the geographic region of cities. (See Table 5 for the definition of the regions used in the analysis.) We thus attempt to explain as much of the variation in the factor scores as possible using this set of non-policy variables. The remaining, or unexplained, portion of the variation is the maximum portion which might be attributed to policy at the national, state, or city levels.

The four predictive models explain a significant amount of the variance in the factor scores, signifying that much of what effects city performance is unrelated to public policy. This is particularly true of the models of the Income and Education factor and the Population and Employment factor, in which 66 percent and 56 percent, respectively, of the variation in the factor scores is explained by the non-policy variables. The least predictive of the four models is that of the Housing Affordability factor, but the model's independent variables still explain a respectable 27 percent of the variance in factor scores. The model explained 34 percent of the variance in the Concentrated Poverty factor scores. Because these regression models are predictive, as opposed to causal, the coefficients on and significance of the independent variables is less important than the models' Adjusted-R<sup>2</sup>s. However, the four models do suggest some interesting causal relationships that may be explored in the future. (See Tables 6 through 9 for the results of the linear regressions.)

### *State Fixed-Effects Models*

As we are primarily interested in those policies that differ at the state level, we use state fixed-effects models to separate state-level policy effects from those that may occur at the national or city levels. In the final part of our statistical analysis, we regress the residuals from the four predictive models against a set of state dummy variables. For this, we used only cities that were in states with at least three central cities. This left us with 34 states<sup>3</sup> and 303 cities. Each set of residuals was used as the dependent variable in a linear regression with the 34 state dummy variables as independent variables. (The constant term was suppressed in the regressions in order to eliminate the need to leave out one of the state dummy variables as a reference group.) The state dummy variables with significant coefficients are those in which state-level policy could have played a significant role in the performance of their cities between 1990 and 2000. (See Tables 10 through 13 for the fixed-effects regression results.)

Using a statistical significance threshold of 0.1, there were seven significant state dummy coefficients when using the residuals from the Income and Education factor score model, five for the Population and Employment factor, ten for Housing Affordability, and nine for Concentrated Poverty. (See Table 14.) The low Adjusted-R<sup>2</sup>s suggest that most of the variation among states is not explained by state policy (or other state-level effects).

The performance of cities in Colorado, Illinois, Oregon, and Washington exceeded expectations on the Income and Education factor, while California, Michigan, and Pennsylvania under-performed on this factor. On the Population and Employment factor, Illinois, North Carolina, and Oregon performed well while Pennsylvania and Virginia performed more poorly than expected. The interpretation of the Housing Affordability factor is trickier: a positive sign

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<sup>3</sup> The states that are not included because they had too few central cities are: Alaska, Delaware, District of Columbia, Hawaii, Idaho, Maine, Maryland, Mississippi, Montana, Nebraska, Nevada, New Hampshire, North Dakota, South Dakota, West Virginia, and Wyoming.

on the state dummy coefficients is related to housing being less affordable in those states. While a lack of affordable housing can be bad for city residents, it is also often a sign of a vibrant, competitive city. Alabama, New Mexico, North Carolina, Pennsylvania, and Washington all had positive, significant coefficients in the Housing Affordability model, meaning cities in those states saw a greater decrease (or smaller increase) in housing affordability than was expected. Cities in California, Kansas, Louisiana, Michigan, and Wisconsin saw a larger increase (or smaller decrease) in housing affordability than the model predicted. On the Concentrated Poverty factor, a positive sign on the state dummy coefficients indicates an increase in poverty concentration. Arizona, California, Indiana, New Mexico, and Rhode Island performed worse than was expected on this factor, while Michigan, Ohio, Texas, and Washington performed better than the model predicted.

### *Selection of case studies*

We selected eight states in which to conduct case studies. Concentrating primarily on the Income and Education, Population and Employment, and Housing Affordability factors, seven states were selected that had significant coefficients on at least two of these three models. These states are:

- California (better than expected performance on Housing Affordability; worse than expected on Income and Education);
- Illinois (better than expected on both Income and Education, and Population and Employment);
- Michigan (better than expected on Housing Affordability; worse than expected on Income and Education);



- North Carolina (better than expected on Population and Employment; worse than expected on Housing Affordability);
- Oregon (better than expected on both Income and Education, and Population and Employment);
- Pennsylvania (worse than expected on Income and Education, Population and Employment, and Housing Affordability); and
- Washington (better than expected on Income and Education; worse than expected on Housing Affordability).

We also looked at lists of cities by state for each set of residuals to see whether some states would have been significant if not for one or two outlier cities. We found that an eighth state, Virginia, in addition to having a significant coefficient (worse than expected) in the Population and Employment factor, also performed poorly on Income and Education and well on Housing Affordability but were not found to be significant due to outliers (Suffolk and Danville on the former, Danville and Roanoke on the latter). We therefore intend to include Virginia in our case studies.

The case studies will focus on policy that was likely to have an effect on city performance during the period from 1990 to 2000<sup>4</sup> and will consist of both document searches and interviews. The interviews will be conducted with state and city policymakers as well as close observers such as journalists, academics, and other informed observers.

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<sup>4</sup> It is important to note that such policies could have been enacted well before the time period in question.

**TABLE 1: Indicators of city performance**

*All variables are measured using the change in well-being from 1990-2000, and all are city-level variables unless otherwise noted.*

<b>Variable</b>	<b>Variable label</b>	<b>Definition</b>
Population	chgpob	Total number of residents <sup>1</sup>
Percent with some college	chgsomecoll	Percent of residents aged 25+ with at least some college <sup>1</sup>
Jobs by place of residence	chgemployed	Total number of employed residents <sup>1</sup>
Jobs by place of work	chgjobsmsa	Total number of jobs in cities' MSA <sup>2</sup>
Vacancy rate	chgvacrate	Vacant units as a percent of total units <sup>1</sup>
Building permits	chgpermits	Number of building permits issued over the preceding 10 years <sup>3</sup>
In-migration	chginmig	Number of residents aged 5+ who are in-migrants over the preceding 5 years <sup>4</sup>
Racial segregation	chgbwdissim, chgbwexps, chghwdissim, chghwexps	Black-white and Hispanic-white dissimilarity and exposure indices <sup>5</sup>
Per capita income	chgpercap	Income per capita <sup>4</sup>
Median household income	chgmedhh	Median household income <sup>1</sup>
Poverty rate	chgpovrate	Individuals in poverty as a percent of total residents <sup>1</sup>
Unemployment rate	chgunemp	Percent of labor force that is unemployed <sup>1</sup>
Labor force participation rate	chglabfrc	Percent of residents aged 16+ who are in the labor force <sup>1</sup>
Homeownership rate	chgowrate	Owner-occupied units as a percent of total occupied units <sup>1</sup>
Median home value	chghmval	Median value of owner-occupied units <sup>1</sup>
Median rent	chgrent	Median gross rent <sup>1</sup>
Concentrated poverty	chgtophn, chgpoorhpn	Percent of all residents living in high-poverty neighborhoods; percent of poor residents living in high-poverty neighborhoods <sup>6</sup>
Housing affordability	chgafford50, chgafford30	Percent of residents spending at least 30% and more than 50% of income on housing <sup>4</sup>
Murder rate	chgmurder	Number of murders per 10,000 residents <sup>7</sup>
Larceny rate	chglarceny	Number of larcenies per 10,000 residents <sup>7</sup>
Income inequality	chginratio	Ratio of number of households with more than \$75,000 in income to the number of households living in poverty <sup>4</sup>
Young adults with no high school degree	chgythnsd	Percent of residents aged 18-24 without a high school degree or equivalent

Sources: <sup>1</sup> State of the Cities Data Sets, 1990 and 2000 Census data; <sup>2</sup> Bureau of Economic Analysis; <sup>3</sup> Census: Manufacturing, Mining, and Construction Statistics; <sup>4</sup> 1990 and 2000 Census; <sup>5</sup> Mumford Center; <sup>6</sup> Provided by Paul Jargowsky, tabulations based on 1990 and 2000 Census data; <sup>7</sup> Uniform Crime Reports, CJIS.

**TABLE 2: Rotated Factor Loadings<sup>5</sup>***See Table 1 for variable labels.**Shaded cells represent highly-loading variables, or those above 0.55.*

Variable	Common Factors				Uniqueness
	1	2	3	4	
chgpob	0.00157	0.95858	0.07211	0.06241	0.06558
chgsomecoll	0.61275	-0.21230	0.00499	-0.00038	0.57002
chgvacrate	0.00912	-0.35727	0.35298	-0.08732	0.68028
chgmedhh	0.82189	0.35531	-0.19806	-0.14596	0.13458
chgpovrate	-0.59665	-0.13368	0.30279	0.55558	0.19878
chgemployed	0.31968	0.88310	0.00731	-0.02593	0.11713
chgunemp	-0.19264	0.01296	0.52431	0.20773	0.61957
chglabfrc	0.65861	-0.13472	0.06456	-0.13799	0.52032
chgrent	0.69489	0.32723	0.27127	-0.05577	0.32646
chghmval	0.76642	0.22037	0.17385	-0.12120	0.31913
chgowndrate	0.14390	0.54530	-0.22790	-0.10660	0.61665
chgpercapp	0.88206	0.11188	-0.12316	-0.08972	0.18623
chgythnhsd	0.04495	0.14872	0.04058	0.12642	0.87239
chgjobsmsa	0.33736	0.65334	0.04583	-0.09444	0.41065
chgafford50	-0.12703	0.06408	0.79790	0.30393	0.24643
chgafford30	0.07090	0.02035	0.89683	0.05705	0.17922
chginmig	0.43782	-0.30312	0.19272	-0.09016	0.66584
chgbwdissim	-0.10099	0.17616	-0.20943	0.14032	0.86885
chgbwexps	0.37236	-0.22014	0.16380	-0.34501	0.60581
chghwdissim	0.22533	0.15241	-0.00475	0.03414	0.39779
chghwexps	0.16200	-0.07122	0.01320	0.06270	0.34310
chginratio	0.30456	0.38333	-0.20893	0.37245	0.56340
chgmurder	0.01758	0.12861	0.15100	0.12110	0.94006
chglarceny	0.14989	-0.16059	0.19277	-0.03851	0.91308
chgpermits	0.24560	0.09994	0.14330	-0.12699	0.85626
chgtothpn	-0.15090	-0.05452	0.16347	0.90611	0.12647
chgpooorhpn	-0.11093	0.06279	0.07505	0.89881	0.16630
Eigenvalue	5.65883	3.42507	2.67161	1.43216	
% of variance	33.11	20.04	15.63	8.38	
Total variance	33.11	53.15	68.78	77.15	

<sup>5</sup> Four factors were retained, all with eigenvalues greater than 1. Varimax rotation was used to orthogonally rotate the factors.

**TABLE 3: City rankings on factor scores**  
*Cities are ranked from best- to worst-performing.<sup>6</sup>*

City	State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
Abilene	TX	252	126	95	184
Akron	OH	79	236	207	42
Alameda	CA	133	239	28	321
Albany	GA	217	233	198	67
Albany	NY	283	315	317	191
Albuquerque	NM	156	79	176	231
Allentown	PA	295	242	296	307
Altoona	PA	96	281	248	208
Amarillo	TX	161	125	87	99
Anaheim	CA	324	53	156	159
Anchorage	AK	292	65	46	213
Anderson	IN	54	234	153	148
Ann Arbor	MI	201	145	37	207
Appleton	WI	72	160	31	294
Arlington	TX	282	34	106	187
Arlington CDP	VA	196	152	16	283
Asheville	NC	22	133	306	297
Atlanta	GA	15	182	113	137
Aurora	IL	17	18	197	192
Austin	TX	11	8	42	223
Bakersfield	CA	317	13	242	200
Baltimore	MD	248	321	279	94
Baton Rouge	LA	158	197	117	41
Battle Creek	MI	139	200	15	46
Bayonne	NJ	258	265	299	175
Baytown	TX	259	129	85	218
Beaumont	TX	216	228	134	28
Bellevue	WA	157	43	237	253
Bellingham	WA	185	48	325	144
Berkeley	CA	16	217	290	324
Bethlehem	PA	237	277	302	177
Billings	MT	131	113	112	165
Binghamton	NY	264	322	127	279
Birmingham	AL	115	303	252	93
Bloomington	IL	19	49	129	268
Bloomington	IN	51	101	191	314
Boca Raton	FL	153	76	110	288
Boise City	ID	103	6	297	179
Bossier City	LA	145	153	14	309
Boston	MA	197	213	32	241
Boulder	CO	36	73	235	254
Bridgeport	CT	305	257	36	139
Brockton	MA	298	215	8	281
Brownsville	TX	224	7	148	4

<sup>6</sup> On Housing Affordability, a decrease in the availability of affordable housing is assumed to be bad. However, it must be noted that an alternative interpretation of the lack of affordable housing is that it is a sign of a competitive, vibrant city.

**TABLE 3: City rankings on factor scores (continued)**

<b>City</b>	<b>State</b>	<b>Income and Education</b>	<b>Population and Employment</b>	<b>Housing Affordability</b>	<b>Concentrated Poverty</b>
Bryan	TX	263	52	249	43
Buffalo	NY	189	320	229	134
Cambridge	MA	47	156	315	269
Camden	NJ	254	313	175	25
Canton	OH	34	258	158	102
Cape Coral	FL	214	16	192	239
Cedar Rapids	IA	93	137	137	142
Champaign	IL	175	178	97	126
Charleston	SC	76	91	289	54
Charleston	WV	38	275	196	133
Charlotte	NC	74	38	281	57
Chattanooga	TN	49	260	201	170
Cheyenne	WY	137	140	91	230
Chicago	IL	99	191	74	81
Cincinnati	OH	70	307	171	88
Clarksville	TN	172	19	200	87
Clearwater	FL	162	173	225	190
Cleveland	OH	40	294	145	89
College Station	TX	64	31	164	304
Colorado Springs	CO	35	23	84	259
Columbia	MO	63	58	179	64
Columbia	SC	199	95	307	63
Columbus	GA	114	186	116	98
Columbus	OH	85	131	241	53
Corpus Christi	TX	140	118	119	82
Council Bluffs	IA	27	192	250	247
Dallas	TX	274	69	194	68
Danbury	CT	302	111	50	222
Danville	VA	155	299	314	110
Davenport	IA	109	181	103	91
Dayton	OH	116	288	187	14
Daytona Beach	FL	210	223	212	34
Dearborn	MI	271	146	323	153
Decatur	IL	166	273	219	284
Denton	TX	136	41	35	11
Denver	CO	25	61	173	97
Des Moines	IA	147	212	120	85
Detroit	MI	69	224	6	3
Dothan	AL	230	147	125	7
Dubuque	IA	65	244	170	233
Duluth	MN	33	220	230	198
Durham	NC	83	30	238	150
East Lansing	MI	205	311	319	317
Eau Claire	WI	20	136	51	115
El Paso	TX	241	99	165	66
Elgin	IL	132	47	226	295
Elyria	OH	44	240	224	164
Erie	PA	221	269	292	30
Escondido	CA	318	50	34	221
Eugene	OR	95	51	310	104

**TABLE 3: City rankings on factor scores (continued)**

City	State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
Evanston	IL	123	261	266	313
Evansville	IN	105	289	131	160
Everett	WA	87	39	308	212
Fairfield	CA	234	45	132	290
Fall River	MA	236	271	19	258
Fargo	ND	128	59	39	132
Fayetteville	NC	167	4	256	35
Flint	MI	218	231	4	13
Fort Collins	CO	3	12	57	302
Fort Lauderdale	FL	176	199	90	275
Fort Smith	AR	211	149	231	131
Fort Wayne	IN	174	87	221	141
Fort Worth	TX	229	74	128	83
Fresno	CA	294	71	253	211
Gainesville	FL	242	97	293	18
Galveston	TX	148	230	185	121
Gary	IN	52	310	83	257
Gastonia	NC	90	96	316	215
Grand Rapids	MI	121	210	60	163
Great Falls	MT	146	187	154	118
Greeley	CO	92	27	277	59
Green Bay	WI	78	158	68	224
Greensboro	NC	207	103	257	95
Greenville	SC	50	276	282	143
Hamilton	OH	31	245	104	10
Hampton	VA	232	161	143	242
Harrisburg	PA	164	296	278	234
Hartford	CT	321	314	217	79
High Point	NC	21	109	133	267
Honolulu CDP	HI	301	209	300	189
Houston	TX	261	56	126	55
Huntington	WV	144	305	311	185
Huntsville	AL	250	232	183	161
Indianapolis	IN	111	193	262	111
Iowa City	IA	107	135	267	229
Irvine	CA	312	37	115	265
Irving	TX	256	40	72	251
Jackson	MS	255	304	177	31
Jacksonville	FL	141	104	142	136
Janesville	WI	57	92	109	272
Jersey City	NJ	233	243	76	188
Joliet	IL	37	22	202	92
Kalamazoo	MI	180	262	203	21
Kansas City	KS	125	274	69	107
Kansas City	MO	118	237	178	135
Kenosha	WI	14	124	157	248
Killeen	TX	73	17	12	270
Knoxville	TN	106	202	285	193
La Crosse	WI	61	205	45	225
Lafayette	LA	67	63	58	9

**TABLE 3: City rankings on factor scores (continued)**

City	State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
Lake Charles	LA	98	207	82	17
Lakeland	FL	190	155	223	186
Lancaster	CA	325	78	295	282
Lancaster	PA	253	218	288	77
Lansing	MI	209	259	44	60
Laredo	TX	183	5	239	2
Las Cruces	NM	260	66	210	264
Las Vegas	NV	270	1	280	100
Lawrence	KS	23	60	21	149
Lawrence	MA	309	175	2	16
Lawton	OK	188	120	70	201
Lexington-Fayette	KY	71	122	182	236
Lincoln	NE	81	88	195	183
Little Rock	AR	127	203	100	171
Lodi	CA	288	148	286	205
Long Beach	CA	320	179	121	320
Longmont	CO	4	15	270	299
Longview	TX	231	183	166	86
Lorain	OH	62	266	135	155
Los Angeles	CA	311	198	146	255
Louisville	KY	24	302	254	249
Lowell	MA	275	195	5	49
Lubbock	TX	245	138	139	39
Lynchburg	VA	163	248	247	56
Lynn	MA	290	169	7	125
Macon	GA	246	300	216	124
Madison	WI	89	128	190	157
Manchester	NH	257	184	18	300
Mansfield	OH	150	227	240	58
McAllen	TX	198	20	159	1
Melbourne	FL	219	77	96	174
Memphis	TN	126	185	260	24
Merced	CA	314	86	276	70
Meriden	CT	299	282	141	217
Mesa	AZ	238	11	24	240
Miami	FL	235	162	138	75
Miami Beach	FL	2	235	3	23
Midland	TX	297	117	186	8
Milwaukee	WI	206	285	88	27
Minneapolis	MN	68	222	54	76
Mobile	AL	152	208	218	80
Modesto	CA	300	89	174	123
Monroe	LA	18	278	23	112
Montgomery	AL	203	170	233	44
Muncie	IN	88	292	244	244
Napa	CA	208	83	41	261
Nashua	NH	277	150	13	280
Nashville-Davidson	TN	97	168	245	105
New Bedford	MA	251	308	30	308
New Haven	CT	307	298	204	292

**TABLE 3: City rankings on factor scores (continued)**

City	State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
New Orleans	LA	129	216	10	47
New York	NY	269	139	272	146
Newark	NJ	278	253	227	214
Newport News	VA	225	196	79	276
Niagara Falls	NY	195	318	149	101
Norfolk	VA	200	312	78	120
Norman	OK	120	55	59	286
North Charleston	SC	306	72	52	169
North Little Rock	AR	60	270	111	109
Norwalk	CT	279	190	150	243
Oakland	CA	130	141	22	210
Odessa	TX	284	154	155	22
Ogden	UT	53	68	318	199
Oklahoma City	OK	202	121	124	113
Olathe	KS	26	9	9	319
Omaha	NE	48	127	147	119
Orem	UT	1	21	268	278
Orlando	FL	171	142	172	156
Oshkosh	WI	66	93	56	130
Owensboro	KY	108	188	136	12
Palm Bay	FL	310	24	61	180
Palo Alto	CA	10	174	107	316
Pasadena	CA	262	241	73	237
Pawtucket	RI	304	264	140	256
Pensacola	FL	151	204	151	38
Peoria	IL	100	256	243	266
Philadelphia	PA	265	297	255	197
Phoenix	AZ	266	25	102	158
Pine Bluff	AR	192	272	75	29
Pittsburgh	PA	104	319	228	62
Pontiac	MI	55	279	66	40
Port Arthur	TX	173	252	89	36
Port St. Lucie	FL	308	2	144	293
Portland	ME	168	251	26	145
Portland	OR	9	62	312	116
Portsmouth	VA	149	287	167	26
Providence	RI	291	246	65	322
Provo	UT	12	33	162	84
Pueblo	CO	45	143	236	69
Racine	WI	112	238	180	78
Raleigh	NC	143	35	258	65
Rapid City	SD	122	110	33	252
Reading	PA	296	254	320	306
Redding	CA	247	54	264	195
Reno	NV	268	26	208	196
Richmond	VA	187	293	214	138
Riverside	CA	315	108	163	273
Roanoke	VA	138	267	271	162
Rochester	MN	135	70	168	245
Rochester	NY	285	306	251	262



**TABLE 3: City rankings on factor scores (continued)**

City	State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
Rockford	IL	223	164	263	108
Sacramento	CA	289	114	169	128
Saginaw	MI	86	249	1	45
Salem	OR	39	44	284	246
Salinas	CA	316	10	98	202
Salt Lake City	UT	5	75	298	206
San Angelo	TX	165	172	38	154
San Antonio	TX	80	42	40	20
San Bernardino	CA	319	107	269	182
San Buenaventura (Ventura)	CA	286	167	62	238
San Diego	CA	244	130	43	289
San Francisco	CA	6	165	11	301
San Jose	CA	101	84	20	305
Santa Ana	CA	323	80	17	147
Santa Barbara	CA	213	177	213	232
Santa Clara	CA	29	81	48	325
Santa Fe	NM	117	98	291	209
Santa Maria	CA	322	36	220	127
Santa Rosa	CA	212	32	49	260
Sarasota	FL	186	211	259	203
Savannah	GA	142	284	273	103
Schenectady	NY	273	309	206	312
Scottsdale	AZ	94	3	81	291
Scranton	PA	124	295	199	235
Seattle	WA	7	144	287	228
Shreveport	LA	159	201	27	52
Sioux City	IA	59	171	189	72
Sioux Falls	SD	46	64	99	220
South Bend	IN	178	250	283	181
Spokane	WA	58	112	303	114
Springfield	IL	84	180	188	168
Springfield	MA	303	286	93	106
Springfield	MO	134	163	261	71
Springfield	OH	30	291	92	287
St. Charles	MO	154	116	80	274
St. Joseph	MO	41	194	209	74
St. Louis	MO	102	325	122	178
St. Paul	MN	110	219	67	129
St. Petersburg	FL	75	225	160	151
Stamford	CT	293	157	309	167
Stockton	CA	272	94	265	173
Suffolk	VA	13	82	77	216
Sunnyvale	CA	28	115	25	323
Syracuse	NY	281	323	305	166
Tacoma	WA	32	151	304	122
Tallahassee	FL	222	90	313	194
Tampa	FL	56	189	193	152
Tempe	AZ	220	85	55	227
Terre Haute	IN	42	176	274	271

**TABLE 3: City rankings on factor scores (continued)**

<b>City</b>	<b>State</b>	<b>Income and Education</b>	<b>Population and Employment</b>	<b>Housing Affordability</b>	<b>Concentrated Poverty</b>
Toledo	OH	170	290	123	51
Topeka	KS	193	214	222	219
Trenton	NJ	287	268	322	176
Troy	NY	226	317	301	226
Tucson	AZ	227	46	63	61
Tulsa	OK	194	166	130	90
Tuscaloosa	AL	82	206	246	6
Tyler	TX	169	100	94	15
Utica	NY	179	324	232	318
Vallejo	CA	249	132	118	277
Victoria	TX	113	105	64	50
Vineland	NJ	239	221	234	250
Virginia Beach	VA	243	119	29	296
Visalia	CA	240	57	105	140
Waco	TX	160	134	215	32
Waltham	MA	77	247	108	303
Warren	OH	191	283	161	37
Warwick	RI	215	263	101	263
Washington	DC	177	301	181	315
Waterbury	CT	313	280	211	298
Waterloo	IA	43	229	152	73
Waukesha	WI	91	106	114	285
West Palm Beach	FL	204	102	275	172
Wichita	KS	184	123	47	96
Wichita Falls	TX	182	159	184	33
Wilmington	DE	228	226	294	311
Wilmington	NC	8	28	321	19
Winston-Salem	NC	181	67	205	117
Worcester	MA	276	255	53	204
Yakima	WA	267	29	324	48
Youngstown	OH	119	316	71	5
Yuma	AZ	280	14	86	310

**TABLE 4: Independent variables used in linear regression models**

*Dependent variables are the set of four factor scores. Most variables represent both level in 1990 and change between 1990 and 2000.<sup>7</sup>*

<b>Variable</b>	<b>Variable label</b>	<b>Definition</b>
Percent manufacturing	pctmanuf90, chgmanuf	Percent of the labor force working in the manufacturing industry <sup>1</sup>
Percent FIRE	pctfire90, chgfire	Percent of the labor force working in the finance, insurance, and real estate industries <sup>1</sup>
Dependent population	pctdepend90, chgdepend	Percent of residents aged 16 and younger or 65 and over <sup>1</sup>
Average July temperature	avgjulytemp	Average July temperature <sup>2</sup>
Annexed population	pctpopannx	Population annexed 1991-2000 as a percent of total 1990 population <sup>3</sup>
Percent Black	pctblk90, chgpctblk	Percent of residents who are Black, non-Hispanic <sup>1</sup>
Percent Hispanic	pcthis90, chgpcthis	Percent of residents who are Hispanic <sup>1</sup>
College enrollment	pctenroll90	Number of students enrolled in cities' 4-year institutions as a percent of total population <sup>4</sup>
Average wage	wagperjob90, chgwagperjob	Average wage and salary disbursements per job, MSA level <sup>5</sup>
City age	cityage	Years in 1990 since city passed 50,000 in population <sup>6</sup>
Capital dummy variable	capital	1= City is state capital
Regional dummy variables	coastse, contfarwest, greatlake, inlandse, noncontfarw, nthmideast, nthneweng, plains, rockymtn, sthneweng, sthwest, newyork	Dummy variables for region state is in (See Table 5 for list of states by region.)

Sources: <sup>1</sup> State of the Cities Data System, 1990 and 2000 Census data; <sup>2</sup> Federal Research Division of the Library of Congress; <sup>3</sup> Census Boundary and Annexation Survey; <sup>4</sup> Integrated Postsecondary Education Data System; <sup>5</sup> Bureau of Economic Analysis; <sup>6</sup> Census data 1890-1990.

<sup>7</sup> Exceptions are average July temperature, annexed population, and percent of residents enrolled in college.

**TABLE 5: Description of regional dummy variables**

<b>Region</b>	<b>Dummy Variable Label</b>	<b>Definition</b>
Coastal Southeast	coastse	Florida, Georgia, North Carolina, South Carolina, Virginia
Continental Far West	contfarwest	California, Nevada, Oregon, Washington
Great Lakes	greatlake	Illinois, Indiana, Michigan, Ohio, Wisconsin, all New York state MSAs west of Albany, all Pennsylvania MSAs west of Philadelphia
Inland Southeast	inlandse	Alabama, Arkansas, Kentucky, Louisiana, Mississippi, Tennessee, West Virginia
Non-continental Far West	noncontfarw:	Alaska, Hawaii
Northern Mideast	nthmideast	New Jersey except those in NY CMSA, New York except those in Great Lakes or New York CMSA, Pennsylvania except those in Great Lakes
Northern New England	nthneweng	Maine, New Hampshire, Vermont
Plains	plains	Iowa, Kansas, Minnesota, Missouri, Nebraska, North Dakota, South Dakota
Rocky Mountains	rockymtn	Colorado, Idaho, Montana, Utah, Wyoming
Southern Mideast	sthmideast	Delaware, District of Columbia, Maryland
Southern New England	sthneweng	Connecticut, Massachusetts, Rhode Island
Southwest	sthwest	Arizona, New Mexico, Oklahoma, Texas
New York City CMSA	newyork	New York City CMSA

**TABLE 6: Linear regression model of Income and Education factor***Dependent variable is the income and education factor score.*

Variable	Coefficient	t-statistic	p-value
constant term	1.0486	1.19	0.236
pctmanuf90	0.0286	3.41***	0.001
chgmanuf	0.0729	2.74***	0.007
pctfire90	0.0187	0.84	0.403
chgfire	0.1766	3.74***	0.000
pctdepend90	-0.0439	-3.45***	0.001
chgdepend	-0.2175	-9.20***	0.000
avgjulytemp	-0.0051	-0.65	0.516
pctpopannx	0.0316	1.98**	0.048
pctblk90	0.0023	0.87	0.386
chgpctblk	-0.0800	-6.01***	0.000
pcthis90	-0.0065	-2.11**	0.036
chgpcthis	-0.0829	-8.19***	0.000
pctenroll90	-0.0091	-2.04**	0.042
cityage	0.0030	2.75***	0.006
capital	-0.1056	-0.93	0.351
coastse	0.8412	2.37**	0.019
contfarwest	0.6453	1.72*	0.086
greatlake	0.8907	2.44**	0.015
inlandse	1.1021	3.01***	0.003
noncontfarw	-0.2196	-0.39	0.694
nthmideast	0.2993	0.71	0.478
nthneweng	0.0339	0.07	0.946
plains	0.9234	2.46**	0.014
rockymtn	1.8288	4.62***	0.000
sthneweng	0.1409	0.36	0.720
sthwest	1.0502	2.81***	0.005
newyork	-0.1445	-0.37	0.711
N		325	
R <sup>2</sup>		0.6855	
Adjusted R <sup>2</sup>		0.6569	
F-statistic (27, 297)		23.98 (p-value 0.000)	

\* = significant at 0.1 level; \*\* = significant at 0.05 level; \*\*\* = significant at 0.01 level

Regional reference group = Southern Mideast

The highly loading variables on this factor are (all are change from 1990-2000):

- percent of adult residents with some college
- median household income
- per capita income
- poverty rate
- labor force participation rate
- median home value
- median rent

**TABLE 7: Linear regression model of Population and Employment factor**  
*Dependent variable is the population and employment factor score.*

Variable	Coefficient	t-statistic	p-value
constant term	-0.2240	-0.19	0.849
pctmanuf90	-0.0136	-1.28	0.201
chgmanuf	-0.0118	-0.38	0.705
pctfire90	0.0333	1.22	0.224
chgfire	-0.0490	-0.90	0.369
pctdepend90	-0.0690	-4.30***	0.000
chgdepend	-0.0171	-0.62	0.538
avgjulytemp	0.0420	4.62***	0.000
pctpopannx	0.0568	3.06***	0.002
pctblk90	-0.0098	-3.10***	0.002
chgpctblk	-0.0197	-1.28	0.200
pcthis90	0.0043	1.22	0.222
chgpcthis	0.0351	3.01***	0.003
pctenroll90	-0.0202	-3.77***	0.000
cityage	-0.0049	-3.78***	0.000
wagprjob90	-0.0001	-2.87***	0.004
chgwagprjob	0.0129	3.41***	0.001
capital	-0.1986	-1.53	0.127
coastse	0.3306	0.81	0.421
contfarwest	0.4695	1.16	0.247
greatlake	0.2955	0.74	0.458
inlandse	0.0721	0.17	0.865
noncontfarw	0.9348	1.48	0.141
nthmidwest	-0.1621	-0.34	0.737
plains	0.2120	0.51	0.607
rockymtn	0.7683	1.75*	0.080
sthmidwest	0.0027	0.00	0.996
sthneweng	-0.0768	-0.18	0.858
sthwest	0.2041	0.48	0.632
newyork	-0.1360	-0.30	0.763
N		325	
R <sup>2</sup>		0.5994	
Adjusted R <sup>2</sup>		0.5600	
F-statistic (29, 295)		15.22 (p-value 0.000)	

\* = significant at 0.1 level; \*\* = significant at 0.05 level; \*\*\* = significant at 0.01 level

Region reference group = Northern New England

The highly loading variables on this factor are (all are change from 1990-2000):

population

employment by place of residence (city level)

employment by place of work (MSA level)

**TABLE 8: Linear regression model of Housing Affordability factor**  
*Dependent variable is the housing affordability factor score.*

Variable	Coefficient	t-statistic	p-value
constant term	0.8632	0.61	0.539
pctmanuf90	0.0091	0.69	0.494
chgmanuf	-0.0318	-0.82	0.414
pctfire90	0.0139	0.41	0.684
chgfire	0.0632	0.93	0.352
pctdepend90	0.0483	2.41**	0.016
chgdepend	0.0763	2.21**	0.028
avgjulytemp	-0.0123	-1.09	0.277
pctpopannx	0.0404	1.75*	0.081
pctblk90	-0.0062	-1.58	0.116
chgpctblk	-0.0236	-1.23	0.218
pcthis90	-0.0091	-2.08**	0.038
chgpcthis	0.0251	1.73*	0.085
pctenroll90	0.0284	4.26***	0.000
cityage	0.0052	3.26***	0.001
wagprjob90	-0.0001	-4.55***	0.000
chgwagprjob	0.0013	0.27	0.789
capital	0.5747	3.55***	0.000
coastse	0.3884	1.86*	0.064
contfarwest	0.6608	3.11***	0.002
inlandse	0.0756	0.34	0.738
noncontfarw	0.7967	1.23	0.221
nthmideast	0.7119	1.93*	0.054
nthneweng	-1.3783	-2.78***	0.006
plains	-0.3569	-1.73*	0.084
rockymtn	0.2509	0.87	0.383
sthmideast	1.1529	2.21**	0.028
sthneweng	-1.1065	-4.12***	0.000
sthwest	-0.0547	-0.23	0.822
newyork	1.2854	3.89***	0.000
N		325	
R <sup>2</sup>		0.3371	
Adjusted R <sup>2</sup>		0.2719	
F-statistic (29, 295)		5.17 (p-value 0.000)	

\* = significant at 0.1 level; \*\* = significant at 0.05 level; \*\*\* = significant at 0.01 level

Region reference group = Great Lakes

The highly loading variables on this factor are (all are change from 1990-2000):

- percent of households spending more than 30% of income on housing
- percent of households spending more than 50% of income on housing

**TABLE 9: Linear regression model of Concentrated Poverty factor**

*Dependent variable is the concentrated poverty factor score.*

Variable	Coefficient	t-statistic	p-value
constant term	1.5905	1.16	0.248
pctmanuf90	-0.0225	-1.78*	0.077
chgmanuf	-0.0692	-1.87*	0.063
pctfire90	-0.0147	-0.45	0.653
chgfire	-0.0190	-0.29	0.770
pctdepend90	-0.0450	-2.35**	0.019
chgdepend	0.0391	1.18	0.239
avgjulytemp	-0.0032	-0.29	0.768
pctpopannx	-0.0359	-1.62	0.106
pctblk90	-0.0210	-5.59***	0.000
chgpctblk	0.0051	0.28	0.782
pcthis90	-0.0226	-5.38***	0.000
chgpcthis	0.227	1.64	0.102
pctenroll90	-0.0091	-1.43	0.155
cityage	0.0010	0.66	0.511
wagprjob90	0.00004	1.57	0.117
chgwagprjob	0.0069	1.52	0.129
capital	-0.0154	-0.10	0.921
coastse	0.1062	0.40	0.692
contfarwest	0.2423	0.92	0.358
greatlake	-0.0287	-0.10	0.917
inlandse	-0.2372	-0.80	0.423
noncontfarw	-0.5035	-0.77	0.444
nthmideast	0.2392	0.57	0.567
nthneweng	-0.0307	-0.06	0.953
plains	-0.1280	-0.46	0.643
sthmideast	1.2597	2.31**	0.021
sthneweng	-0.1832	-0.56	0.579
sthwest	-0.3081	-1.16	0.248
newyork	-0.0757	-0.20	0.838
N		325	
R <sup>2</sup>		0.4013	
Adjusted R <sup>2</sup>		0.3424	
F-statistic (29, 295)		6.82 (p-value 0.000)	

\* = significant at 0.1 level; \*\* = significant at 0.05 level; \*\*\* = significant at 0.01 level

Region reference group = Rocky Mountains

The highly loading variables on this factor are (all are change from 1990-2000):

poverty rate

percent of population living in poor neighborhoods

percent of poor population living in poor neighborhoods



**TABLE 10: Linear regression model of residuals from Income and Education factor model**

State	Coefficient	t-statistic	p-value
Alabama	-0.159	-0.73	0.469
Arizona	-0.045	-0.20	0.838
Arkansas	0.067	0.25	0.802
California	-0.162	-1.81 **	0.071
Colorado	0.405	2.00 **	0.047
Connecticut	-0.198	-1.04	0.299
Florida	0.029	0.24	0.808
Georgia	0.034	0.14	0.887
Illinois	0.306	1.89 **	0.060
Indiana	0.193	1.08	0.282
Iowa	0.015	0.08	0.936
Kansas	0.278	1.16	0.248
Kentucky	0.004	0.01	0.990
Louisiana	0.212	1.04	0.297
Massachusetts	0.093	0.57	0.566
Michigan	-0.394	-2.43 **	0.016
Minnesota	-0.099	-0.37	0.712
Missouri	-0.044	-0.20	0.841
New Jersey	0.274	1.25	0.213
New Mexico	0.219	0.71	0.479
New York	-0.226	-1.33	0.184
North Carolina	0.198	1.17	0.244
Ohio	0.204	1.42	0.156
Oklahoma	-0.337	-1.26	0.210
Oregon	0.863	2.79 **	0.006
Pennsylvania	-0.301	-1.77 **	0.077
Rhode Island	-0.184	-0.59	0.554
South Carolina	-0.421	-1.57	0.118
Tennessee	-0.098	-0.41	0.684
Texas	0.031	0.32	0.749
Utah	0.198	0.74	0.462
Virginia	-0.096	-0.59	0.555
Washington	0.337	1.66 **	0.098
Wisconsin	0.100	0.62	0.535
N		303	
R <sup>2</sup>		0.1553	
Adjusted R <sup>2</sup>		0.0486	
F-statistic <sub>(34, 269)</sub>		1.46 (p-value 0.0558)	

\*\* = significant at the 0.10 level

**TABLE 11: Linear regression model of residuals from Population and Employment factor model**

State	Coefficient	t-statistic	p-value
Alabama	0.014	0.06	0.955
Arizona	0.189	0.78	0.327
Arkansas	-0.368	-1.23	0.218
California	-0.159	-1.60	0.112
Colorado	-0.073	-0.32	0.747
Connecticut	-0.177	-0.84	0.401
Florida	-0.135	-1.01	0.312
Georgia	-0.299	-1.12	0.264
Illinois	0.445	2.48 **	0.014
Indiana	-0.146	-0.74	0.462
Iowa	-0.254	-1.20	0.230
Kansas	0.204	0.77	0.445
Kentucky	-0.192	-0.56	0.578
Louisiana	0.320	1.42	0.157
Massachusetts	0.096	0.54	0.592
Michigan	0.181	1.01	0.315
Minnesota	0.156	0.52	0.601
Missouri	-0.050	-0.20	0.838
New Jersey	-0.112	-0.46	0.646
New Mexico	-0.216	-0.63	0.531
New York	-0.138	-0.73	0.464
North Carolina	0.680	3.61 **	0.000
Ohio	0.005	0.03	0.976
Oklahoma	-0.023	-0.08	0.938
Oregon	0.635	1.85 **	0.066
Pennsylvania	-0.448	-2.38 **	0.018
Rhode Island	-0.085	-0.25	0.805
South Carolina	0.194	0.65	0.515
Tennessee	0.143	0.54	0.591
Texas	-0.013	-0.12	0.905
Utah	0.348	1.17	0.244
Virginia	-0.307	-1.71 **	0.088
Washington	0.197	0.88	0.382
Wisconsin	0.137	0.76	0.446
N		303	
R <sup>2</sup>		0.1570	
Adjusted R <sup>2</sup>		0.0504	
F-statistic <sub>(34, 269)</sub>		1.47 (p-value 0.0501)	

\*\* = significant at the 0.10 level

**TABLE 12: Linear regression model of residuals from Housing Affordability factor model**

State	Coefficient	t-statistic	p-value
Alabama	0.598	1.97 **	0.049
Arizona	-0.492	-1.62	0.105
Arkansas	-0.047	-0.13	0.900
California	-0.253	-2.04 **	0.042
Colorado	0.044	0.16	0.877
Connecticut	-0.182	-0.70	0.488
Florida	-0.082	-0.49	0.622
Georgia	0.070	0.21	0.833
Illinois	0.316	1.41	0.159
Indiana	0.059	0.24	0.813
Iowa	0.307	1.17	0.243
Kansas	-0.689	-2.08 **	0.039
Kentucky	0.199	0.46	0.643
Louisiana	-1.017	-3.63 **	0.000
Massachusetts	-0.157	-0.70	0.484
Michigan	-0.488	-2.18 **	0.030
Minnesota	0.212	0.57	0.568
Missouri	0.178	0.59	0.557
New Jersey	0.398	1.31	0.190
New Mexico	0.713	1.66 **	0.097
New York	0.157	0.67	0.504
North Carolina	0.403	1.72 **	0.087
Ohio	0.023	0.12	0.907
Oklahoma	-0.277	-0.75	0.455
Oregon	0.583	1.36	0.175
Pennsylvania	0.621	2.65 **	0.009
Rhode Island	0.030	0.07	0.944
South Carolina	0.013	0.04	0.972
Tennessee	0.361	1.09	0.277
Texas	0.062	0.47	0.642
Utah	0.151	0.41	0.684
Virginia	-0.254	-1.14	0.257
Washington	1.036	3.69 **	0.000
Wisconsin	-0.549	-2.45 **	0.015
N		303	
R <sup>2</sup>		0.2257	
Adjusted R <sup>2</sup>		0.1278	
F-statistic <sub>(34, 269)</sub>		2.31 (p-value 0.0001)	

\*\* = significant at the 0.10 level

**TABLE 13: Linear regression model of residuals from Concentrated Poverty factor model**

State	Coefficient	t-statistic	p-value
Alabama	-0.432	-1.47	0.142
Arizona	0.565	1.93 **	0.055
Arkansas	0.354	0.98	0.326
California	0.224	1.87 **	0.063
Colorado	-0.115	-0.42	0.672
Connecticut	-0.134	-0.53	0.598
Florida	-0.011	-0.07	0.946
Georgia	0.327	1.02	0.310
Illinois	0.260	1.20	0.232
Indiana	0.576	2.41 **	0.017
Iowa	-0.073	-0.29	0.773
Kansas	0.295	0.92	0.359
Kentucky	-0.003	-0.01	0.995
Louisiana	0.071	0.26	0.794
Massachusetts	-0.213	-0.99	0.325
Michigan	-0.545	-2.52 **	0.012
Minnesota	-0.153	-0.42	0.671
Missouri	-0.143	-0.49	0.627
New Jersey	0.204	0.70	0.487
New Mexico	1.286	3.10 **	0.002
New York	0.283	1.25	0.214
North Carolina	-0.248	-1.09	0.276
Ohio	-0.563	-2.93 **	0.004
Oklahoma	0.333	0.93	0.355
Oregon	-0.542	-1.31	0.193
Pennsylvania	0.181	0.80	0.426
Rhode Island	0.805	1.94 **	0.053
South Carolina	-0.126	-0.35	0.726
Tennessee	0.084	0.26	0.793
Texas	-0.277	-2.14 **	0.033
Utah	0.213	0.59	0.553
Virginia	0.143	0.66	0.511
Washington	-0.729	-2.68 **	0.008
Wisconsin	0.089	0.41	0.681
N		303	
R <sup>2</sup>		0.2044	
Adjusted R <sup>2</sup>		0.1039	
F-statistic (34, 269)		2.03 (p-value 0.0010)	

\*\* = significant at the 0.10 level

**TABLE 14: Significant states by model**

*Dependent variables are the residuals from the non-policy models of the factor scores; independent variables are the state dummy variables*

State	Income and Education	Population and Employment	Housing Affordability	Concentrated Poverty
Alabama			+	
Arizona				+
Arkansas				
California	-		-	+
Colorado	+			
Connecticut				
Florida				
Georgia				
Illinois	+	+		
Indiana				+
Iowa				
Kansas			-	
Kentucky				
Louisiana			-	
Massachusetts				
Michigan	-		-	-
Minnesota				
Missouri				
New Jersey				
New Mexico			+	+
New York				
North Carolina		+	+	
Ohio				-
Oklahoma				
Oregon	+	+		
Pennsylvania	-	-	+	
Rhode Island				+
South Carolina				
Tennessee				
Texas				-
Utah				
Virginia		-		
Washington	+		+	-
Wisconsin			-	

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