Risk of Toxic Stress for Children

Using Census Data to Target Services

The Institute for Urban Policy Research At The University of Texas at Dallas

Timothy M. Bray, PhD



Risk of Toxic Stress for Children

Using Census Data to Target Services

Stress, and the body's response to it, play an important part in shaping the current and future well-being of children. When the body's response is conditioned by constant stressors, such as living in extreme poverty, lacking caring adults in life, and other correlates of poverty and socioeconomic disadvantage, the body's response to stress becomes toxic, and it can have a profound impact on child development. In this paper, we use publicly available data from the American Community Survey's (ACS) Public Use Micro Sample (PUMS) dataset to estimate the number and percentage of children at risk for experiencing toxic stress. Figure 1 depicts the eight risk factors that we selected. We further explore the racial and ethnic distribution of toxic stress and the geographic distribution in Dallas County, Texas. We find that 438,000 Dallas County children (64%) have at least one of eight risk factors for toxic stress. Prevalence of toxic stress for children of color tends to be nearly double that for non-Hispanic White children.



Housing Distress



Over-crowding



Deep Poverty



Parental Unemployment



Lack of Transportation



Lack of Health Insurance



Single-Parent Households



Falling Behind in School

Figure 1. Overview of Census Demographics Tapped as Risk Factors for Toxic Stress

What Is Toxic Stress?

The concept of toxic stress emerged in 2005, when the National Scientific Council on the Developing Child (2014) published a report describing the effects of stress on brain development in children. Dr. Jack Shonkoff (2010), a physician and member of the Council, later reported that the Council was attempting to build a taxonomy to communicate the effects of prolonged exposure to stress to policy makers and the medical community. The term *toxic stress* is primarily used to set it apart from other, more common types of childhood stress.

Positive Stress

- •A stress response to normative developmental events
- •"Characterized by moderate, short-lived increases in heart rate, blood pressure . . . stress hormones . . . an important aspect of healthy development is experienced in the context of stable and supportive relationships that facilitate adaptive responses that restore the stress response system to baseline."
- •Example Precipitates: playground injury, shots at the doctor's office

Tolerable Stress

- A stress response to non-normative developmental events
- •"A physiological state that could potentially disrupt brain architecture but is buffered by supportive relationships that facilitate adaptive coping."
- •Example Precipitates: A serious injury, death of a loved one, natural disaster

Toxic Stress

- •"Strong, frequent, or prolonged activation of the body's stress response systems in the absense of the buffering protection of stable adult support."
- •Example Precipitates: Extreme poverty, family violence, abuse, and neglect

Figure 2. Shonkoff's (2010) Definitions for a Taxonomy of Childhood Stress Responses

Shonkoff (2010) provides a simple set of definitions for the three types of stressors identified by the Council, illustrated in Figure 2. As seen from the taxonomy proposed, toxic stress is not just about the stressor but also the availability of caring, responsive adults to help the child cope with the stress. Toxic stress, then, refers to the deleterious effects of prolonged exposure to serious stressors without recourse to a caring, responsive adult.

Why Toxic Stress Matters

At its core, the toxic stress responses are most detrimental to children, whose brains are developing at a rapid pace. Shonkoff et al. (2012) reported that "the essential characteristic of this phenomenon is the postulated disruption of brain circuitry and other organ and metabolic systems during sensitive developmental periods" (p. 236). They further suggested that toxic stress responses from adverse childhood experiences may account, to some degree, for the racial and ethnic disparities in health observed later in life. Shonkoff et al. framed the effects of exposure to toxic stress in two mediating

forces: (a) altered architecture of the brain during key developmental phases and (b) prolonged exposure to stress hormones that accompany toxic stress. A summary of the short- and long-term consequences outlined by Shonkoff et al. is provided in Figure 3.

Short-Term Effects

- Functional differences in learning, memory, and aspects of executive functioning during stress responses.
- More anxiety during stress responses, with less top-down control of the stress response and impaired memory and mood control.

Long-Term Effects

- A chronically activated stress response, with increased potential for fear and anxiety.
- Impairments in memory and moodrelated functions.
- Decreased ability for the hippocampus to promote contextual learning and disrupted memory encoding systems.
- Stunted development in executive functions, such as decision-making, working memory, behavioral selfregulation, and mood and impulse control.

Figure 3. A Summary of Effects of Toxic Stress Reported by Shonkoff et al., 2012

Methodology

The US Census Bureau conducts the ACS monthly; results are tabulated annually and every 5 years, producing what are known as the 1-Year and 5-Year estimates. From the individual survey responses, the Census Bureau pulls a representative subset and publishes them as the PUMS dataset. Each survey response is geographically coded according to the Public Use Micro Sample Area (PUMA) in which the respondent lived. PUMAs are drawn to be sufficiently large so as to ensure the anonymity of census respondents. Areas of larger population are more precisely subdivided than areas with fewer people. In North Texas, for instance, Dallas County is subdivided into more than 20 PUMAs, while Rockwall and Hunt counties are combined into a single PUMA. Here, we estimate the risk for toxic stress by aggregating the PUMAs that compose Dallas County.

A number of questions the ACS asks may indicate risk for toxic stress responses. Figure 4 presents a brief outline of the eight risk factors that we have selected, and their nominal definitions. Following Figure 4, we provide a more technical definition for each risk factor.

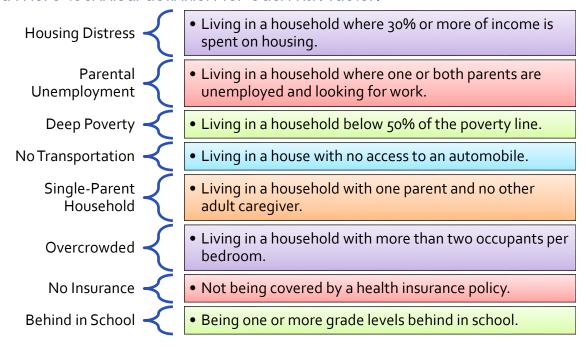


Figure 4. ACS-Based Toxic Stress Risk Factor Definitions

We defined children as those whose age, at the time of data collection, was under 18. We coded each child record with a 1 or o for each risk factor they possessed. Specifically, we defined the risk factors as follows.

- Children were defined at risk if their household's housing costs exceeded 30% of the household income and household income was below 200% of the federal poverty threshold: POVPIP < 200 AND [GRPIP ≥ 0.30 OR OCPIP ≥ 0.30] (see Kingsley & Austin Turner, 2008; Nepal, Tanton, & Harding, 2010; Quigley & Raphael, 2004).
- Parental unemployment was defined as living in a family with one or both spouses unemployed and looking for work (WORKSTAT = 2, 4, 6, 8, 11, or 14).
- Children were defined at risk by living in deep poverty: a household income below 50% of the poverty line, or POVPIP < 50 (see Ratcliffe & McKernan, 2012).
- No access to transportation was defined as living in a house with no access to an automobile (VEH = 0).
- Single-parent households were defined as living with one parent with no spouse, unmarried partner, or grandparent present (FES ≥ 5 AND PARTNER = 0 AND MULTGEN = 0).
- Risk due to overcrowding was defined as more than two people per bedroom and household income below 200% of the federal poverty threshold: POVPIP < 200 AND NP/BDSP > 2 (Blake, Kellerson, & Simic, 2007).

- Lack of health insurance was defined as indicating no health insurance coverage (HINS1 = 0 AND HINS2 = 0 AND HINS3 = 0 AND HINS4 = 0 AND HINS5 = 0 AND HINS6 = 0 AND HINS7 = 0).
- Risk from falling behind in school was defined as repeating a grade one or more times (SCHG ≥ 3 AND SCHG ≤ 14 AND SCHG < SCHL).

We computed confidence intervals for counts and proportions using the Census-provided replicate weights. This successive difference replication methodology allowed us to provide population-level estimates from the PUMS data, which is essentially a sample from a sample, while also providing confidence intervals and margins of error that respect this sampling design.

Findings

Table 1 presents the number of children estimated to be at risk for experiencing toxic stress in Dallas County by risk factor. Three factors each affect more than 100,000 children: housing distress, overcrowding, and behind grade level. The number of children with at least one risk factor is 438,525, or more than 2 in 3 Dallas County children.

Table 1. Number of Children by Risk Factor

Risk Factor	Number at Risk	95% Confidence Interval
Housing Distressed	255,888 38%	(±15,519) (±2.3%)
Parental Unemployment	36,252 5%	(±2,370) (±7,382) (±1.1%)
Deep Poverty	56,513 8%	(±8,750) (±1.3%)
No Transportation	29,306 4%	(±5,817) (±0.9%)
Single-Parent Household	41,928 6%	(±7,117) (±1.0%)
Overcrowded Housing	132,012 19%	(±12,467) (±1.8%)
No Insurance	72,577 11%	(±9,236) (±1.4%)
Behind Grade Level	111,543 16%	(±10,196) (±1.5%)
One or More Risk Factors	438,525 64%	(±13,236) (±1.9%)

Of the 682,000 children in Dallas County, only 36% (some 244,000) had no risk factors for toxic stress, which was also the modal category. The average child had 1.08 risk factors (±0.05). One-half of the children (54%) had one or two risk factors. The highest number of risk factors that any child had was six. Less than 1% of children had this high a risk. Figure 5 presents the number of children by their total number of risk factors. As seen in the graphic, nearly as many children had one risk factor as had no risk factors.

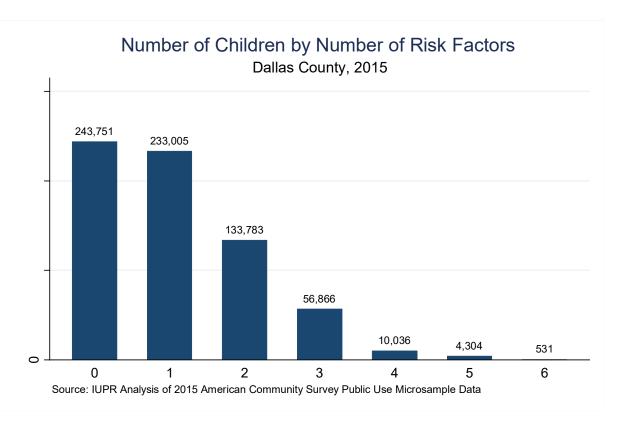


Figure 5. Number of Children by Number of Risk Factors.

Variation by Race

As one might expect, there is significant variation in the prevalence of toxic stress risk factors across racial, ethnic, and age categories. presents the percentage of Dallas County children with one or more and three or more risk factors by age and race/ethnicity. The 95% margin of error is displayed under each estimate in parentheses. Children of color are more likely to have risk factors than non-Hispanic White children are, and this effect is most pronounced for children with three or more risk factors. While 16% of non-Hispanic African American children under 5 have three or more risk factors, only 9% of Hispanic children do, and only 2% of non-Hispanic White children do. Moreover, age appears to make a bigger difference among non-Hispanic White children than non-White children. For both one or more risk factors and three or more risk factors, the prevalence among children ages 5 to 17 is roughly twice that of children under 5. Among children of color, the prevalence is more closely aligned.

Table 2. Percentage of Childhood Population by Number of Risk Factors, Age, and Race/Ethnicity.

	At Lease 1 Risk Factor		3 or More Risk Factors	
	Under 5	5 to 17	Under 5	5 to 17
Non-Hispanic White	29%	50%	2%	5%
	(±6.7%)	(±4.8%)	(±1.6%)	(±2.6%)
Non-Hispanic Black	65%	68%	16%	14%
	(±8.6%)	(±4.5%)	(±6.4%)	(±4.0%)
Non-Hispanic Asian	39%	53%	1%	9%
	(±14.6%)	(±10.6%)	(±2.7%)	(±5.7%)
Non-Hispanic Other	52%	60%	11%	13%
	(±14.6%)	(±12.7%)	(±7.8%)	(±9.3%)
Hispanic	63%	75%	9%	12%
	(±4.2%)	(±2.8%)	(±3.0%)	(±2.4%)

Table 3 belowpresents a more complete picture of the distribution of toxic stress risk by race and ethnicity, while Figure 6 provides a graphic illustration. As both the table and graphic attest, housing distress is the most prominent risk factor for all five racial and ethnic groups. For non-Hispanic White children, however, it is tied with falling behind in school, each with 18% of children at risk. Comparing risk among the groups reveals that, while all five groups lead with housing distress, it is far more prominent among non-Hispanic African American children (41% of children at risk) and Hispanic children (44% of children at risk).

For non-Hispanic African American children, the second and third leading risk factors were deep poverty (13% of children) and falling behind in school (14% of children). The story was markedly different for Hispanic children, however, with the second most prevalent risk factor being overcrowded housing, with 27% of children at risk. This was followed by risk of falling behind in school (16%) and having no health insurance (14% of Hispanic children—double that of non-Hispanic African American children, and more than double that of non-Hispanic Whites).

Table 3. Number and Percentage of Children at Risk of Toxic Stress by Risk Factor, Race, and Ethnicity, Dallas County 2015 (95% Confidence Intervals in Parentheses)

	Non-Hispanic White	Non-Hispanic Black	Non-Hispanic Asian	Non-Hispanic Other	Hispanic
Housing	22,111	60,100	6,605	7,876	159,196
Distressed	18%	41%	20%	36%	44%
	(±4,502)	(±7,200)	(±2,635)	(±3,200)	(±11,909)
	(±3.6%)	(±4.9%)	(±8.0%)	(±11.4%)	(±3.3%)
Parental	4,668	13,044	3,358	1,053	14,129
Unemployment	4%	9%	10%	5%	4%
	(±2,219)	(±4,795)	(±1,678)	(±890)	(±4,503)
	(±1.8%)	(±3.3%)	(±5.1%)	(±3.9%)	(±1.3%)
Deep Poverty	4,370	19,467	512	1,992	30 , 172
	4%	13%	2%	9%	8%
	(±1,885)	(±5,541)	(±597)	(±1,389)	(±6,576)
	(±1.5%)	(±3.8%)	(±1.8%)	(±6.2%)	(±1.8%)
No	¹ ,533	16,996	1,557	1,225	7,995
Transportation	1%	12%	5%	6%	2%
	(±848)	(±5,014)	(±1,196)	(±935)	(±3,108)
	(±0.7%)	(±3.5%)	(±3.6%)	(±4.2%)	(±0.9%)
Single-Parent	4,682	14,915	205	1,078	21,048
Household	4%	10%	1%	5%	6%
	(±2,067)	(±4,277)	(±353)	(±905)	(±5,751)
	(±1.7%)	(±3.0%)	(±1.1%)	(±4.1%)	(±1.6%)
Overcrowded	11,172	16,157	4,166	3,663	96 , 854
Housing	9%	11%	13%	17%	27%
	(±3,457)	(±4,962)	(±2,266)	(±2,274)	(±10,872)
	(±2.8%)	(±3.5%)	(±6.6%)	(±9.0%)	(±3.0%)
No Insurance	7,784	10,673	2,469	1,159	50,492
	6%	7%	8%	5%	14%
	(±2,393)	(±3,628)	(±1,483)	(±1,028)	(±7,912)
	(±1.9%)	(±2.5%)	(±4.4%)	(±4.7%)	(±2.2%)
Behind Grade	22,648	20,100	6,262	3,7 ¹ 3	58,820
Level	18%	14%	19%	17%	16%
	(±3,157)	(±4,267)	(±1,667)	(±1,576)	(±8,058)
	(±2.5%)	(±3.0%)	(±5.1%)	(±6.3%)	(±2.2%)

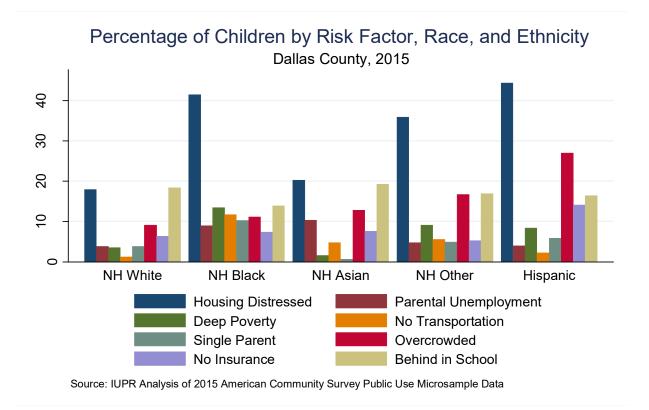


Figure 6. Percentage of Children at Risk of Toxic Stress by Risk Factor, Race, and Ethnicity, Dallas County 2015

Variation by Geography

Figure 7 presents the geographic distribution of children with at least one risk factor in Dallas County. Both northeast and south central sections of the county, including Rowlett and parts of Garland, Sachse, Desoto, and Glenn Heights, had the lowest risk, with less than 50% of children with one or more risk factors. Areas with the highest prevalence included northeast city of Dallas, as well as eastern Dallas, western Mesquite, and parts of Balch Springs, with 74% to 80% of children with one or more risk factors.

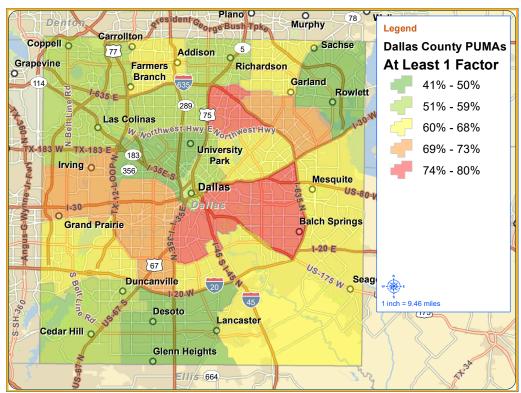


Figure 7. Distribution of Percentage of Children With One or More Risk Factors by PUMA, Dallas County 2015

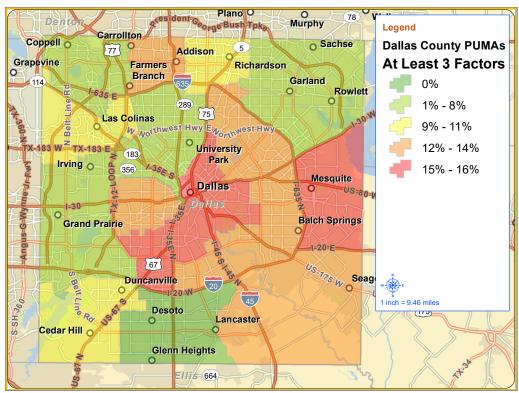


Figure 8. Distribution of Percentage of Children With Three or More Risk Factors by PUMA, Dallas County 2015

Figure 8 presents a map of the distribution of children with three or more of the eight risk factors. Similar patterns hold, with northeast and south central Dallas County with the lowest risk. However,

this map sees northwestern Dallas County and eastern Dallas County join those with the lowest risk. Old East Dallas remains a significant area of need, as does most of Oak Cliff. Southern Garland, however, drops in classification, as does eastern Dallas County (Mesquite, Balch Springs, and Sunnyvale), though less so.

Figure 9 presents the information using a different tact, presenting for each geography the median number of risk factors for children in the area—one half of children fall above this number and one half fall below. Again, northeast and south central Dallas County show the lowest risk rates, with the median having no risk factors. Most of the remainder of the county shows the median of one risk factors one half of children have one or more risk factors, one half have one or none. Only one PUMA, representing South Dallas and parts of Oak Cliff, indicates the highest level of need, where the median was two risk factors. One half of children had two or more risk factors.

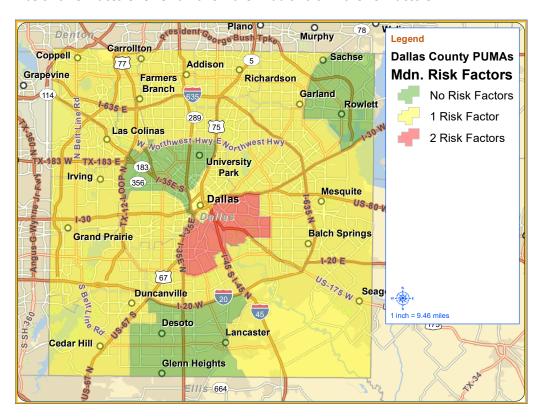


Figure 9. Distribution of Median Number of Risk Factors by PUMA, Dallas County 2015

Finally, Figure 10 represents the severity of need in the neighborhood by mapping the number of risk factors at the 90th percentile: 90% of children have the indicated number of risk factors or fewer. Now, only south central Dallas County (Desoto, Lancaster, parts of Glenn Heights) shows having the least relative need, with 90% of children having one or fewer risk factors. Western Dallas County, northwest city of Dallas, and northeast Dallas County (Garland, Sachse, and Rowlett) show moderate levels of need, with 90% of children having two or fewer risk factors. The PUMA representing South Dallas and parts of Oak Cliff continues to show the most severe need, with 90% of children having two or fewer risk factors.

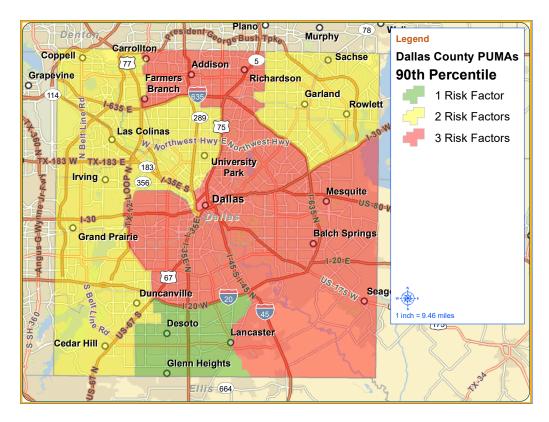


Figure 10. Distribution of 90th Percentile Risk Factors by PUMA, Dallas County 2015

Summary

Our work demonstrates that, at a minimum, there is utility in turning to the PUMS data to estimate the number of children at risk of toxic stress responses. Moreover, by using these tools to produce estimates of risk, more careful analysis can then explore variations in estimated risk by age, gender, race/ethnicity, and other demographic characteristics. Because of Dallas County's population, the county is divided into sufficiently small PUMAs, allowing the exploration of geographic variation in risk of toxic stress.

Care should be taken to refine this method of estimation. Chief among its weaknesses is that it is an estimate of risk rather than a prevalence of risk. Future clinical research should establish a more definitive relationship between these, and potentially other, indicators available via the ACS. By turning to advanced estimation techniques using publicly available data, we should be able to more precisely target specific resources to the communities and children most in need, mitigating the near- and long-term damage done by toxic stress.

References

- Blake, K. S., Kellerson, R. L., & Simic, A. (2007). *Measuring overcrowding in housing.* Washington, DC: U.S. Department of Housing and Urban Development.
- Kingsley, G. T., & Austin Turner, M. (2008). *Federal programs for addressing low-income housing needs.*Washington, DC: The Urban Institute.
- National Scientific Council on the Developing Child. (2014). *Excessive stress disrupts the architecture of the developing brain*. Cambridge, MA: National Scientific Council on the Developing Child.
- Nepal, B., Tanton, R., & Harding, A. (2010). Measuring housing stress: How much do definitions matter? *Urban Policy and Research*, 28(2), 211–224.
- Quigley, J. M., & Raphael, S. (2004). Is housing unaffordable? Why isn't it more affordable? *The Journal of Economic Perspectives*, 18(1), 191–214.
- Ratcliffe, C. E., & McKernan, S. (2012). *Child poverty and its lasting consequence.* Washington, DC: The Urban Institute.
- Shonkoff, J. P. (2010). Building a new biodevelopmental framework to guide the future of early childhood policy. *Child Development*, 81(1), 357–367.
- Shonkoff, J. P., Garner, A. S., Siegel, B. S., Dobbins, M. I., Earls, M. F., McGuinn, L., . . . Wood, D. L. (2012). The lifelong effects of early childhood adversity and toxic stress. *Pediatrics*, 129(1), 232–246.



800 West Campbell Road, WT20 Richardson, Texas 75080-3021

(972)883-5430 voice (972)883-5431 facsimile

Visit us on the Web! http://iupr.utdallas.edu

