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## A life course approach to understanding stress exposures and cognitive function among middle-aged and older adults

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#### Abstract

**Background:** Many studies have evaluated the stress-cognition association, but few have captured the cumulative nature of stress or distinguished the influences of stressors occurring in childhood versus adulthood. Using a lifecourse approach, we investigated whether cumulative stress exposures are associated with poorer cognitive function and faster cognitive decline.

**Methods:** We used data from the Midlife Development in the United States Study (N= 3,954, mean baseline age: 56 years). We fit marginal structural generalized estimating equations models to estimate the difference in baseline cognitive function per *SD* increment in the continuous stressor score, and, separately, between persons in each life course stressor profile and those who did not experience high stress in either childhood or adulthood. We also characterized differences in cognitive decline across levels of stress exposures.

**Results:** Higher cumulative stress exposure was associated with lower executive function (*difference per SD in* continuous stressor score = -0.12 SD units, 95% CI = -0.16, -0.08) and episodic memory (*difference* = -0.09 SD units, 95% CI = -0.13, -0.05). Baseline executive function and episodic memory were lower among those with high stress only in childhood, only in adulthood, and both, than among those without high stress in childhood or adulthood. There was little evidence that rate of change in executive function and episodic memory differed across levels of cumulative stress exposures.

**Conclusions:** These findings offer support to the hypothesis that stress exposures, accumulated over the life course, worsen cognitive performance, but limited support for the hypothesis that these exposures promote cognitive decline.

#### Keywords

Stress; Executive Function; Episodic Memory; Cognition; Lifecourse

Conflict of Interests: None Declared

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#### Introduction

Growing evidence suggests that individuals exposed to high levels of stress face faster cognitive declines and higher risk of mild cognitive impairment in older adulthood, an early stage of dementia (Aggarwal et al., 2014; Korten et al., 2017). These studies shed light on a highly prevalent potential risk factor for dementia that may be amenable to intervention. However, much of this research has used measures of perceived stress, which primarily assess recent perceptions of general stress, typically within the past month. With the restricted time frame and types of questions asked, such measures may not adequately capture stress exposures that have accumulated over the life course (Aggarwal et al., 2014; Korten et al., 2017). Even though stressors from multiple domains often co-occur (Sternthal et al., 2011), of the studies that have considered the cognitive effects of exposure to specific stressors, almost all have considered stress exposure occurring in only one domain (e.g., work stress) (Barnes et al., 2012; Deligkaris et al., 2014).

The stress and health literature has long suggested the importance of considering multiple types of stress exposures. The stress process framework proposed by Pearlin, for example, posits that stressors, whether in the form of an untoward event or a chronic strain associated with a social role, tend to give rise to additional stressors, a process called stress proliferation (Pearlin, 1989; Pearlin et al., 2005). According to this framework, stress proliferation often results in clusters of stressors and cumulative adversities that shape health and well-being later in life. Building upon Pearlin's stress proliferation framework, Wheaton further proposed the stress domain hypothesis, which posits that no single source of stress can capture the full impact of stress, and that assessment of the impacts of stress on health outcomes depends on considering multiple key sources of stress over significant periods of time in lives (Wheaton, 1999, 1994). Together, the stress domain hypothesis and the stress proliferation hypothesis suggest that both untoward events and chronic strains shape individuals' stressful experiences, and that, to validly estimate the consequences of these experiences, it is essential to assess constellations of stressors made up of both events and strains. Importantly, both Pearlin and Wheaton advocated the use of a life-course framework to study stress (Pearlin, 2010; Wheaton, 1994). As noted by Pearlin, "the observation and understanding of stress proliferation are best realized when viewed within a life-course framework" (Pearlin, 2010).

Given that stress exposures occur at multiple time points during the life course and that even early life exposures appear to influence later health outcomes, incorporating a life course approach into research on stress exposures and cognitive outcomes in middle and older adulthood may facilitate better understanding of the stress-cognition relationship. Epidemiologists and sociologists have proposed different, but not mutually exclusive, conceptual models to explain how exposures across the life course can act to influence health outcomes (Ferraro et al., 2009; Kuh et al., 2003). For example, the accumulation of risk model posits that cumulative exposures across the life course increase the risk of adverse health late in life, regardless of the stage of life when the exposures occur. Prior research using data from the Midlife Development in the United States Study, the Chicago Community Health study, and the Women's Health Study has shown consistent support for the accumulation of risk model (Albert et al., 2017; Cuevas et al., 2019; Slopen et al., 2012).

These studies, which typically aggregated multiple domains of acute and chronic stress exposures to create a composite score of cumulative stress exposures, found that higher levels of cumulative stress exposures across the life course were associated with higher odds of smoking, obesity, and cardiovascular diseases (Albert et al., 2017; Cuevas et al., 2019; Slopen et al., 2012).While the detrimental consequences of cumulative stress exposures have been evaluated in relation to these and other health outcomes, little is known about whether cumulative stress exposures are associated with worse cognitive outcomes.

Separately, the critical period model suggests that exposures at a particular stage in life (e.g., early childhood) lead to long-lasting and irreversible health consequences later in life. For example, a previous study reported that childhood distress was associated with higher cardiometabolic risk later in life, even when adjusting for adulthood stress, indicating that the health consequences of early childhood distress are independent from the effects of stress in other life stages (Winning et al., 2015). Because there has been little distinction between the adverse effects of childhood stressors and adulthood stressors on cognitive outcomes in adulthood, it is unclear whether this critical period model can be used to explain how stress exposures influence cognitive outcomes. Distinguishing the associations of childhood and adulthood stressors with cognitive function, separately and in combination, will add further insight to what is known about the sequelae of these exposures, and, by extension, provide further impetus for the benefits of viewing dementia prevention sa a series of life-long interventions, starting early in life, rather than interventions late in life.

Guided by the stress process framework, the stress domain hypothesis, and applying a life course approach to this inquiry, we used data from the Midlife Development in the United States Study (MIDUS) to investigate whether higher life course stress exposures are associated with lower levels of and a faster decline in cognitive function. MIDUS included a rich set of psychosocial measures, allowing us to characterize the relationships between cumulative stress exposures and cognitive function. Our study complements and expands on prior research in MIDUS that has assessed the associations between stressors and cognitive function. For example, Lynch and Lachman used MIDUS2 and MIDUS3 data to assess the relationship between lifetime stressful life events and cognitive function (Lynch and Lachman, 2020). They found that stressful life events predicted faster declines in cognitive function. Stawski et al used data from the MIDUS daily stressor project (n=1,500) and showed that better cognitive function was associated with healthier profiles of naturally occurring cortisol (Stawski et al., 2011). Other research has used MIDUS data to examine associations between chronic exposure to specific stressors and cognitive function. For example, Munoz et al used cross-sectional data from three samples including MIDUS to examine neighborhood stress and cognitive function. Their study showed that higher neighborhood stress was associated with lower executive function in MIDUS (Munoz et al., 2015). Grzywacz et al examined the cross-sectional associations between workplace exposures and cognitive function in MIDUS and found that greater physical strains were associated with poorer episodic memory and executive function (Grzywacz et al., 2016). Lindert et al assessed the associations between social stress and cognitive declines and showed that daily discrimination had adverse effects on executive function (Lindert et al., 2021). Together, findings of these studies inform our understanding of the stress-cognition association. However, most studies have considered a single type or domain of stress. As

older adults tend to experience multiple stress exposures across the life course, considering only one type of stress may underestimate or mischaracterize the total burdens of stress on cognitive function. Additionally, many of these studies included only a single assessment of cognitive function and therefore could not assess associations between stress and cognitive decline.

Building upon prior research on stress and cognitive function, we hypothesized that 1) higher cumulative life course stress exposures would be associated with lower level of and a faster decline in cognitive function; and 2) individuals who were exposed to high childhood stress would have worse cognitive outcomes later in life, regardless of their levels of adulthood stress exposures. We operationalized cumulative stress exposures by ten different domains of stress exposures, including childhood stress, financial stress, neighborhood stress, work psychological stress, work physical stress, work-family conflict, perceived discrimination, perceived inequality, relationship stress, and stressful life events during adulthood. We created a continuous cumulative stress exposure score, by combining measures of ten stress exposures across the life course and examined its relationship with cognitive function level and rate of cognitive decline. To investigate how stress exposures at specific times during the life course are associated with cognitive function, we created a profile of cumulative stress that characterized both level and timing of exposure to stressors over the life course.

A common practice in estimating the combined effects of stressors in different parts of the life course is to fit a regression model that includes terms for childhood and adulthood stressors, as well as terms for putative sources of confounding in the pre-childhood period (e.g., age, sex, parental education) and adulthood period (e.g., income, working status) (Sternthal et al., 2011). However, using this approach may yield biased estimates of the effects of childhood stressors and any joint effect of these stressors with adulthood stressors. This is, because adulthood confounders may also be affected by childhood stressors, a phenomenon called treatment-confounder feedback (Robins et al., 2000). To address this methodological challenge, we fit marginal structural models to account for potential stressor-confounder feedback.

#### Methods

#### Sample population

We used data from MIDUS, a national random digit dial (RDD) sample of noninstitutionalized adults. MIDUS began in 1995–1996 (MIDUS1) with a telephone survey followed by mailed self-administered questionnaires (Barry, 2014). The MIDUS1 sample also included twin pairs and non-twin siblings of the main RDD sample respondents. Of the 7,108 MIDUS1 respondents (aged 25–75 years), 4,963 were re-interviewed in 2004– 2006 (MIDUS2 national sample; 75% of surviving participants). At MIDUS2, to increase the representation of Black Americans, MIDUS investigators recruited 592 new Black participants from Milwaukee, WI. In 2013–2014, the third wave of data (MIDUS3) was collected from 3,294 previously enrolled national sample participants, and a second wave of data was collected from 389 Milwaukee participants.

The Cognitive Project began in MIDUS2. It involved separate telephone interviews of MIDUS2 participants (N= 4,814; 86% response rate) and continued in MIDUS3 with a follow-up assessment (Hughes et al., 2018). Compared with those who did not participate in the MIDUS cognitive project, individuals who participated in the cognitive project were more likely to be female, White, and have higher education levels and annual household income. MIDUS was approved by the Institutional Review Boards at all participating institutions. Additional details about MIDUS can be found elsewhere (Barry, 2014; Brim et al., 2004; Ryff, 2016).

For the current study, we considered MIDUS2 as the cognitive baseline. From the 5,555 individuals who participated in MIDUS2, we excluded those who did not complete the self-administered questionnaires (given at MIDUS2), which contained the stressor questions, and those without complete baseline cognitive function data. Compared with those in our analytic sample, individuals who were excluded were more likely to be Black adults, males, without a college degree, and have lower than \$25000 annual income. The analytic sample included 3,954 adults (of whom 1350 were twins or siblings; 259 were from the Milwaukee sample). Of these participants, 2,517 completed the follow-up cognitive function assessment at MIDUS3 (181 were from the Milwaukee sample). Of those who did not, 480 died prior to the scheduled assessment, and 957 did not complete the assessment for other reasons (see Figure 1). Compared with those who remained in MIDUS3, those who were lost to follow-up due to death were more likely to be older, male, with lower parental education, lower respondent education, lower household income, not working, had lower cumulative stress exposures, and had lower baseline cognitive function scores. Those who were lost to follow-up due to reasons other than death were more likely to be younger, male, with higher education levels, with lower education, had higher cumulative stress exposures, and had lower baseline cognitive function scores (see Appendix Table 6).

#### Measures

**Childhood Stress Exposures**—Childhood stress exposures were assessed at MIDUS2 with seven items from the revised Adverse Childhood Experience questionnaire and nine items from the MIDUS stressful life event inventory (*see* Appendix Table 8) (Turner and Wheaton, 1995). Participants who responded affirmatively to each item indicated their age at the time of the experience. We summed experiences that occurred before age 18 to obtain a score for childhood stress exposures and transformed the summed score into a z-score. Following previous research (Cuevas et al., 2019),we designated the top quartile of the z-score as "high childhood stress."

Adulthood Stress Exposures—We measured stress exposures that occurred during adulthood, also assessed at MIDUS2, using questions from nine stressor domains: financial stress, neighborhood stress, work psychological stress, work physical stress, work-family conflict, perceived discrimination, perceived inequality, relationship stress, and stressful life events during adulthood (Chen et al., 2022; Cuevas et al., 2020; Slopen et al., 2012). We used a broad array of measures designed to assess adulthood stress in various domains. Specifically, financial stress was assessed by asking participants if they had enough money to meet their needs and how difficult it was for them to pay their monthly

bills. Neighborhood stress was measured using a single scale with four items assessing neighborhood safety, perceived neighborhood support, and perceived neighborhood trust (Keyes, 1998). Work psychological stress was measured by a combination of five separate measures assessing skill discretion, decision authority, job demand, coworker support, and supervisor support (Karasek, 1985). Work physical stress was assessed by combining two measures assessing risk of injury or accident on the job and frequency of job strain. Workfamily conflict was assessed by two measures assessing negative work-to-family spillover and negative family-to-work spillover (Grzywacz et al., 2016). Relationship stress was characterized by combining four measures assessing family strain, friend strain, perceived troubles in marriage, and spouse/partner strain (Schuster et al., 1990; Walen and Lachman, 2000). Perceived inequality was derived from three separate measures assessing people's perceptions of inequality across child rearing (e.g., as a family, we have not had the resources to do many fun things together with the children), housing and neighborhood conditions, and work (Ryff et al., 1999). Perceived discrimination was measured by the lifetime discrimination inventory and the well-validated everyday discrimination scale (Williams et al., 1997). Stressful life events in adulthood were assessed by the stressful life event inventory, which includes 20 events that occurred after age 18 (Cohen et al., 1997). A single score was derived for stress exposure in each domain with higher scores indicating higher stress exposure in all domains (see Appendix Table 8). we recomputed the scales such that respondents would have received the lowest value of the scale if a given stressor did not apply to them (e.g., psychosocial work stress did not apply to those not working)(Cuevas et al., 2019; Slopen et al., 2012). To obtain a cumulative stressor score for adulthood stress exposure, we transformed the raw score in each stressor domain into a z-score and summed the standardized scores across the nine domains. We standardized this summed score to facilitate comparisons of the effects of adulthood stress exposures with those of childhood stress exposures. Consistent with "high childhood stress exposures," we designated the top quartile of the z-score as "high adulthood stress exposures."

**Life Course Cumulative Stress Exposures**—We created a life course cumulative stressor score by summing the z-scores of all stressor domains, including childhood stress exposures and the nine domains of adulthood stressor exposures. We standardized the continuous cumulative stressor score, with a higher score indicating a greater level of stressors.

Life Course Stressor Profiles—To characterize exposure to stressors at different points in the life course, we created four mutually exclusive stressor groups: 1) absence of high stressors (i.e., not in the top quartile for stressor exposure in childhood or adulthood) (reference); 2) high stress exposure in "childhood only"; 3) high stress exposure in "adulthood only"; 4) persistently high stress exposure (i.e., high stress exposure in both childhood and adulthood).

**Cognitive Function**—Cognitive function was measured with the Brief Test of Adult Cognition by Telephone (Tun and Lachman, 2006), which includes seven subtests evaluating: immediate recall, delayed recall, working memory span, verbal fluency, inductive reasoning, processing speed, and attention-switching tasks. Drawing on previous

confirmatory factor analyses (Lachman et al., 2014), we characterized cognitive function in two domains: executive function and episodic memory. The MIDUS2 executive function composite score was computed as the mean of standardized scores on the working memory, verbal fluency, inductive reasoning, processing speed, and the Stop & Go Switch Task (SGST) assessed at MIDUS2. Notably, the SGST includes both the accuracy and the latency scores. We focused on task switching latency, which was a composite score based on the average reaction time of the switch and non-switch trials. We reverse coded the score so that a higher score indicated faster reaction times. The episodic memory composite score was computed as the mean of standardized scores on immediate and delayed recalls subtests assessed at MIDUS2. Final MIDUS2 executive function and episodic memory scores were standardized to have a mean of 0 and a standard deviation of 1. To estimate the MIDUS3 executive function score, we first standardized the working memory, verbal fluency, inductive reasoning, processing speed, and attention-switching task subtests based on their raw test scores at MIDUS2. We then took the means of the five domains to obtain the final score for MIDUS3 executive function. In MIDUS3, the SGST values varied by phone type, and therefore we used the score that was corrected for the latency difference in phone type. To calculate the MIDUS3 episodic memory score, we standardized the immediate and delayed recall subtests based on their raw test scores at MIDUS2 and then took the mean score across the two domains to obtain the final MIDUS3 episodic memory score (Hughes et al., 2018).

**Covariates**—We selected covariates based on their observed and hypothesized relationships with the stress exposures and cognitive outcomes. Covariates included age (years), sex (male, female), self-identified race/ethnicity (White, Black, others), father's or mother's highest education level (less than high school, high school or GED, some college, college or more), respondent's education (less than high school, high school or GED, some college, college or more), current annual household income (<\$25,000, \$25,000–44,999, \$45,000–69,999, \$70,000), presence of chronic conditions (yes, no), currently working (yes, no), currently married (yes, no), and having any children (yes, no). All covariates were assessed at MIDUS2 except for parents' education in the national sample, which was assessed at MIDUS1.

#### **Statistical Analyses**

We first checked monotonicity in the relation between cumulative stressor scores and cognitive function by evaluating the relationship using tertiles of cumulative stresss exposures. Compared with those in the lowest tertile of cumulative stressors, those in the  $2^{nd}$  tertile of stressors had markedly lower executive function and episodic memory scores, on average. Both cognitive scores were lower still among those in the highest tertile of stressors, providing some evidence of monotonicity in the relationship (*see* Appendix Table 1). We fit marginal structural generalized estimating equations (GEE) models to estimate the difference in baseline cognitive function score per SD unit increment in the continuous stressor score. Following prior work in MIDUS (Nishimi et al., 2021; Slopen et al., 2016), we accounted for twin and sibling clustering (n = 1350) by including "family ID" in the repeated statement of the GEE models. We conducted separate analyses for each cognitive score, i.e., executive function and episodic memory. We then fit marginal structural GEE

models to examine the differences in baseline cognitive function score per SD between persons in each life course stressor profile and the reference group comprising those who did not experience high stress exposure in either childhood or adulthood. To provide context for the magnitude of the cognitive function differences by stress level, we ran a separate model to estimate the differences in each cognitive score per year in age, adjusting for sex and race/ethnicity.

To estimate corresponding differences in rates of change in cognition, we extended these analyses to GEE models for repeated measures. Specifically, we regressed each cognition score on the cumulative stress exposure variable(s), covariates, time (years since baseline, continuous), cross products of cumulative stress exposures with time, and the cross-products of covariates with time. The coefficient of interest was the cross-product term between stress exposures and time, the mean difference in rate of change in cognitive function per year, per SD unit increase in stress exposure measure. We did not assess the associations between lifecourse stress profiles and changes in cognitive function because the numbers in some strata were too small (e.g., n=10) after stratification by age.

Several potential confounders of the associations of adulthood stress exposures (e.g., education, income) with cognitive outcomes are plausibly affected by childhood stress exposures (see the directed acyclic graph in Figure 2). Including these covariates as terms in the regression models could induce collider bias and/or mask associations of childhood stress exposures on cognitive function. Thus, we calculated inverse probability-of-treatment weights to mitigate confounding of the estimated adult stressor effects in the context of "stressor-confounder feedback" (Hernán et al., 2004; Naimi et al., 2014; Robins et al., 2000). Additionally, to account for potential bias introduced by lost-to-follow-up, we used inverse probability-of-survival weights to separately account for attrition due to death and due to other reasons in longitudinal models of change (See Appendix for detail information regarding inverse probability weights).

**Missing data.**—Data missingness ranged from 0.05% (perceived inequality) to 14% (childhood stressors) of participants; most variables were missing for <5% of participants. We addressed missing stress exposure and covariates with multiple imputation.

**Secondary and sensitivity analyses.**—To probe the extent to which specific stress domains might drive the association of aggregate cumulative stressor score with levels of cognitive function, we fit separate models for each stressor domain in relation to cognitive scores. Since the definition of high vs. low stress was specific to this study sample, we calculated and presented the means and standard errors of individual stress domains across the four lifecourse stress profiles (see Appendix Table 3–4). Additionally, to evaluate the robustness of the results under a broader definition of "high stress exposure," we reran our analyses designating high stress exposure as the top tertile (instead of the top quartile) of each stressor domain. Because change in cognitive function is sensitive to age (Murman, 2015), we stratified these analyses on stressors and cognitive changes by baseline age (<65 years old). As prior research suggests that the association between stressors and changes in cognitive function may differ between men and women (Munro et al., 2019), we also assessed the associations between stressors and cognitive changes by gender. In

constructing work-related stress scores, we assigned the lowest scores for non-workers. Although this approach has been widely used in prior research (Lantz et al., 2005; Slopen et al., 2012), it may not accurately capture the effect of work stress since the sociodemographic profiles of non-workers may differ from that of those who have actually experienced the lowest levels of work-related stress, and also reasons for not working may vary in ways that also affect stress levels and health. In a sensitivity analysis, we restricted the sample to those who were working and reran the primary analyses to check whether this approach influenced the results All analyses were performed in SAS, Version 9.4.

#### Results

At baseline, participants were 28 to 84 years old (mean (*SD*), 56 (12)); approximately one-third were 65 years or older (Table 1). Compared with individuals in the lower two tertiles of the cumulative stress score, those in the highest tertile were more likely to be younger, Black, working, and unmarried; they were also more likely to have a high school degree or lower, have no children, and have chronic conditions. Both MIDUS2 executive function and episodic memory scores were standardized to have a mean of 0 and a standard deviation of 1. At MIDUS3, the average executive function and episodic memory scores were -0.16 (*SD* 0.76) and -0.03 (*SD* 0.98), respectively. The average change in executive function between MIDUS2 and MIDUS3 was -0.32 (*SD* 0.60), and the average change in episodic memory was -0.14 (*SD* 0.94).

Table 2 presents the associations between *continuous* cumulative stress exposures and levels of cognitive function in the full analytical sample and among working adults only. In results from the marginal structural models, a 1-*SD* higher cumulative stress exposure score corresponded to a baseline executive function score that was a mean 0.12 *SD* lower (95% CI, -0.16 to -0.08) and an episodic memory score that was 0.09 *SD* lower (95% CI, -0.13 to -0.05), on average. To place the magnitude of these differences in context, executive function and episodic memory scores were both 0.03 SD lower per year in baseline age. Among working adults only, a 1-*SD* higher cumulative stress exposure score corresponded to a baseline executive function score that was a mean 0.10 *SD* lower (95% CI, -0.14 to -0.05) and an episodic memory score that was 0.06 *SD* lower (95% CI, -0.11 to -0.01).

Table 3 shows the associations between four categorical lifecourse stressor profiles and levels of cognitive function in the full analytical sample and among working adults only. Relative to the mean executive function scores among respondents who did not experience high stress exposure in either childhood or adulthood, the mean score was 0.22 SD units lower (95% CI, -0.31 to -0.12) among those with high stress exposures only in childhood, 0.13 SD units lower (95% CI, -0.21 to -0.05) among those with high stress exposures only in adulthood, and 0.34 SD units lower among those with persistently high stress exposures (95% CI, -0.46 to -0.22). The corresponding differences in episodic memory score were similar to executive function, although the difference in episodic memory scores for those with persistently high stress was less pronounced than differences in executive function. The patterns of the associations of lifecourse profiles with executive function and episodic memory were similar among workers.

Regarding the associations between cumulative stress exposures and the rate of cognitive change, we found little evidence that cumulative stress exposure levels were associated with the rate of cognitive change in the full sample or among workers only (Table 4).

#### Secondary and Sensitivity Analyses

Higher stress in several specific domains—particularly, financial stress, childhood stress, neighborhood stress, perceived discrimination, and relationship stress—was associated with lower executive function scores. Similarly, financial stress, neighborhood stress, perceived inequality, and childhood stress were associated with lower levels of episodic memory (see Appendix Figure 1–2). When using the highest tertile (rather than quartile) as the cut-off defining high versus low stress, the results were similar for executive function, with scores being lower among those with high stress only in childhood or adulthood, and lowest among those with persistently high stress. For episodic memory, results were largely similar except that those with high stress only in childhood (rather than those with high stress in childhood and adulthood) had the lowest scores when compared with individuals in other stress profiles (*see* Appendix Table 2). Additionally, we found no evidence of age and gender differences in the associations between cumulative stressors and changes in cognitive function (*see* Appendix Table 5).

#### Discussion

In this large, longitudinal cohort of middle-aged and older adults, we found strong inverse associations between cumulative exposure to stressors over the life course and cognitive function. Additionally, high stress exposures only in childhood were associated with lower levels of cognitive function later in life. However, there is little evidence that changes in cognitive function over an average of 9 years of follow-up differed across levels of cumulative exposure to stressors.

The inverse associations demonstrated between stressor exposures and cognitive performance in the current study are consistent with most previous research on stress and cognitive aging (Aggarwal et al., 2014; Munoz et al., 2015; Turner et al., 2017), and they provide some evidence to the accumulation of risk model. Our study adds critical detail to the literature by including a more comprehensive assessment of stressors occurring throughout the life course. The findings suggest that cumulative stress exposures have strong impacts on cognitive function and highlight the importance for considering multiple domains of stress exposures when understanding differences in cognitive function at the population level. A variety of mechanisms could underlie the associations between cumulative stressor exposures and cognition. Exposures to stressors across the lifecourse may decrease gray matter and white matter volumes in the hippocampus, a brain region that is crucial for learning and memory (Gianaros et al., 2007). Stress exposures may activate biological stress responses via the hypothalamic-pituitary-adrenal axis, the system responsible for the release of glucocorticoids, including cortisol (Frodl and O'Keane, 2013). As individuals with cognitive impairment have been found to secrete more cortisol than normal older adults (Arsenault-Lapierre et al., 2010; Csernansky et al., 2006; Lind et al., 2007; Marin et al., 2011), it is possible that cumulative stress exposures are

linked to cognitive function through elevated cortisol. Exposure to stressors, especially during childhood, may influence educational attainment and employment opportunities in adulthood and reduce one's likelihood of engaging in cognitively stimulating activities, factors that have been found to be closely linked to cognitive function in late adulthood (Montez and Hayward, 2014). Stress exposures may also lead to unhealthy behaviors (e.g., physical inactivity, smoking) or induce biobehavioral responses (e.g., insomnia and sleep apnea) that, in turn, lead to poor cognitive functioning and dementia (Blondell et al., 2014; Rusanen et al., 2011; Yaffe et al., 2014). Identifying the mechanisms underlying cumulative stress exposures and cognitive function would be an important avenue for future research. Such knowledge will help evaluate whether intervening on these stress exposures has the potential to promote cognitive health at the population level.

Our findings also suggest that even in the absence of high adulthood stress exposure, respondents with a history of high exposure to childhood stressors had poorer cognitive function than those without high stress exposure across the life course. These findings are broadly consistent with previous research on cardiometabolic risk which found that childhood distress was associated with higher cardiometabolic risk in adulthood event after accounting for adulthood distress (Winning et al., 2015). While our childhood stress measure includes 16 items that capture several stressful events, it was less comprehensive than the adulthood stress measures. Our findings on childhood stress and cognitive function may have been even stronger if more stress experiences in childhood have been measured. Together, these findings provide some evidence to the critical period model and suggest that the deleterious impact of childhood stress exposures on adult health could be long-lasting and enduring.

In our sensitivity analysis, work psychological stress and work family conflicts appear to have weaker associations with cognitive function, particularly episodic memory, than other stress domains. It is worth noting that individually, some stress domains may have weaker associations with cognitive outcomes than others, but they may still exert strong impacts on cognitive function when they co-occur with other stress domains. A recurring question in this literature is whether one should assign weights to specific stress domains to create the cumulative stress score. Although weighing the relative importance of multiple stressors seems promising, it is difficult to do so empirically in a way that is generalizable across age and sociodemographic groups. Prior research on cumulative risk factors and developmental outcomes has suggested that unitary weights are more robust predictors than weighted scores (Evans et al., 2013; Slopen et al., 2018; Wainer, 1976). Further investigation is needed into the predictive performance of different operations of cumulative stress in relation to cognitive outcomes

In contrast with prior research, we found little evidence of an association between cumulative stress exposures and rate of cognitive change. Our study included stress exposures occurring across the life course, but it may be that recent stress exposures affect cognitive decline with more potency than remote or enduring stressful experiences. Indeed, a recent study using data from MIDUS data showed that those who were first exposed to traumatic events later in life had greater decline in executive function than those whose first traumatic events occurred earlier in life (Lynch and Lachman, 2020). Likewise, a study

using data from the Baltimore Epidemiologic Catchment Area Follow-up Study found that stressful life events occurring within the last year, but not those occurring prior to the past year, were associated with greater verbal memory decline over approximately 11 years among women but not men (Munro et al., 2019). Unfortunately, we did not know the exact timing of exposure to many of the chronic stressors (e.g., financial stress) assessed in our study. Therefore, we could not distinguish roles of recent from those of stressful experiences occurring well before the MIDUS assessment period, beyond separating out those exposures occurring in childhood. Taken together with findings from other work (Lynch and Lachman, 2020), our findings may suggest that it is important for future research to consider timing when constructing the cumulative stress metric and to evaluate the timing of cumulative stress exposures, including both untoward events and chronic stressors, and their relationships with cognitive declines. The null associations of stress exposure with cognitive decline may also suggest that the stress exposures we measured affect cognitive performance or, in the case of childhood exposures, the developmental processes leading to peak adult performance but not late-life neurodegeneration or decompensation, which is more relevant to cognitive decline and dementia.

#### Limitations

Our study has several limitations. First, childhood stress exposures were retrospectively reported and might have been subject to faulty recall. However, research on twins participating in MIDUS has found moderate– to high– sibling agreement on the self-reported childhood stressor indicators, suggesting retrospective recall bias of the childhood stress exposures may be minimal (Gruenewald et al., 2012). Second, cognitive function status at the time of recall might affect the report of stress exposures. If, for example, those who were with poorer cognitive performance at baseline were less likely to recall stress exposures, then our estimate of the stress-cognition association might have been underestimated. Third, although we included a broad range of potential confounders, unmeasured confounding is likely as in any observational study. Importantly, we did not have information about early life cognitive function, which may have affected both exposure to stressors and later cognitive function. Additionally, our definitions of "high stress" exposure are specific to this sample. However, concerns about such specificity are somewhat mitigated by findings in sensitivity analyses that found largely similar results using a different cut-off point.

Further, despite nine years of follow up, the average change in cognitive score was small, particularly among the younger group, which potentially limited our statistical power to detect factors that affect rate of change. The use of only two waves of cognitive data presents additional methodological challenges, as we were unable to characterize more detailed cognitive trajectories and evaluate their relationship with cumulative stress exposures. Although we have calculated inverse probability weights to account for bias from loss-to-follow-up, there might have been selection bias resulting from differential participation. It is important to recognize those with poorer cognitive function or who have experienced high stress exposures may be less likely to participate in studies like MIDUS or may have been excluded from our sample due to missing SAQ data. Lastly, while our stress measures are more comprehensive than many other previous studies, they might have failed

to capture some important stress exposures (e.g., caregiver stress) that could potentially affect middle-aged and older adults. Additionally, most stress exposure measures in MIDUS did not assess stress appraisals and thus we were not able to assess the extent to which participants experienced any given stress exposure as actually being stressful and then evaluating how much actual stress experienced cumulatively over the life course influenced cognitive function in late life (Brown et al., 2020; Morris et al., 2021). It would be helpful for future research to also consider stress appraisals in their measures of cumulative stress exposures.

#### Conclusion

In conclusion, higher cumulative exposure to stressors across the life course was strongly associated with lower levels of cognitive function. We found little evidence that cumulative stress exposures are associated with changes in cognitive function in this study sample. Our study advances the literature with its assessment of stress exposure in multiple domains. This allowed a more comprehensive assessment of stressor exposure across the life course and how such exposure may be associated with cognitive function. In particular, information on childhood and adulthood stressors allowed us to investigate how stressors at different times in the life course independently and cumulatively influence cognition. By using marginal structural models, our study was able to account for stressor exposure and the cognitive outcomes. Our findings may suggest the value of further investigation when more assessments of cognitive function data are available or with older samples. For future research, it would be useful to collect more detailed information on when stressors occurred in order to determine whether proximity plays a role in the associations between stress and cognitive decline.

#### Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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#### References

- Aggarwal NT, Wilson RS, Beck TL, Rajan KB, De Leon CFM, Evans DA, Everson-Rose SA, 2014. Perceived stress and change in cognitive function among adults aged 65 and older. Psychosom. Med. 76, 80. [PubMed: 24367123]
- Albert MA, Durazo EM, Slopen N, Zaslavsky AM, Buring JE, Silva T, Chasman D, Williams DR, 2017. Cumulative psychological stress and cardiovascular disease risk in middle aged and older women: Rationale, design, and baseline characteristics. Am. Heart J. 192, 1–12. [PubMed: 28938955]
- Arsenault-Lapierre G, Chertkow H, Lupien S, 2010. Seasonal effects on cortisol secretion in normal aging, mild cognitive impairment and Alzheimer's disease. Neurobiol. Aging 31, 1051–1054. [PubMed: 18703254]

- Barnes LL, Lewis TT, Begeny CT, Yu L, Bennett DA, Wilson RS, 2012. Perceived discrimination and cognition in older African Americans. J. Int. Neuropsychol. Soc. 18, 856–865. [PubMed: 22595035]
- Barry TR, 2014. The Midlife in the United States (MIDUS) series: A national longitudinal study of health and well-being. Open Health Data 2.
- Blondell SJ, Hammersley-Mather R, Veerman JL, 2014. Does physical activity prevent cognitive decline and dementia?: A systematic review and meta-analysis of longitudinal studies. BMC Public Health 14, 510. [PubMed: 24885250]
- Brim OG, Ryff CD, Kessler RC, 2004. The MIDUS national survey: An overview. Healthy Are We 1–36.
- Brown LL, Abrams LR, Mitchell UA, Ailshire JA, 2020. Measuring more than exposure: Does stress appraisal matter for Black–White differences in anxiety and depressive symptoms among older adults? Innov. Aging 4, igaa040.
- Chen R, Weuve J, Misra S, Cuevas A, Kubzansky LD, Williams DR, 2022. Racial disparities in cognitive function among middle-aged and older adults: The roles of cumulative stress exposures across the life course. J. Gerontol. Ser. A 77, 357–364.
- Cohen S, Kessler RC, Gordon LU, 1997. Measuring stress: A guide for health and social scientists. Oxford University Press on Demand.
- Csernansky JG, Dong H, Fagan AM, Wang L, Xiong C, Holtzman DM, Morris JC, 2006. Plasma cortisol and progression of dementia in subjects with Alzheimer-type dementia. Am. J. Psychiatry 163, 2164–2169. [PubMed: 17151169]
- Cuevas AG, Chen R, Slopen N, Thurber KA, Wilson N, Economos C, Williams DR, 2020. Assessing the role of health behaviors, socioeconomic status, and cumulative stress for racial/ethnic disparities in obesity. Obesity 28, 161–170. [PubMed: 31858741]
- Cuevas AG, Chen R, Thurber KA, Slopen N, Williams DR, 2019. Psychosocial stress and overweight and obesity: findings from the Chicago community adult health study. Ann. Behav. Med. 53, NP-NP.
- Deligkaris P, Panagopoulou E, Montgomery AJ, Masoura E, 2014. Job burnout and cognitive functioning: A systematic review. Work Stress 28, 107–123.
- Evans GW, Li D, Whipple SS, 2013. Cumulative risk and child development. Psychol. Bull. 139, 1342. [PubMed: 23566018]
- Ferraro KF, Shippee TP, Schafer MH, 2009. Cumulative inequality theory for research on aging and the life course.
- Frodl T, O'Keane V, 2013. How does the brain deal with cumulative stress? A review with focus on developmental stress, HPA axis function and hippocampal structure in humans. Neurobiol. Dis. 52, 24–37. [PubMed: 22426398]
- Gianaros PJ, Jennings JR, Sheu LK, Greer PJ, Kuller LH, Matthews KA, 2007. Prospective reports of chronic life stress predict decreased grey matter volume in the hippocampus. Neuroimage 35, 795–803. [PubMed: 17275340]
- Gruenewald TL, Karlamangla AS, Hu P, Stein-Merkin S, Crandall C, Koretz B, Seeman TE, 2012. History of socioeconomic disadvantage and allostatic load in later life. Soc. Sci. Med. 74, 75–83. [PubMed: 22115943]
- Grzywacz JG, Segel-Karpas D, Lachman ME, 2016. Workplace exposures and cognitive function during adulthood: Evidence from National Survey of Midlife Development and the O\* NET. J. Occup. Environ. Med. Coll. Occup. Environ. Med. 58, 535.
- Hernán MA, Hernández-Díaz S, Robins JM, 2004. A structural approach to selection bias. Epidemiology 615–625.
- Hughes ML, Agrigoroaei S, Jeon M, Bruzzese M, Lachman ME, 2018. Change in cognitive performance from midlife into old age: Findings from the Midlife in the United States (MIDUS) study. J. Int. Neuropsychol. Soc. 24, 805–820. [PubMed: 30019663]
- Karasek R, 1985. Job Content Questionnaire user's guide. Dep. Work Environemnt.
- Keyes CLM, 1998. Social well-being. Soc. Psychol. Q. 121-140.
- Korten NC, Comijs HC, Penninx BW, Deeg DJ, 2017. Perceived stress and cognitive function in older adults: which aspect of perceived stress is important? Int. J. Geriatr. Psychiatry 32, 439–445. [PubMed: 27059116]

- Kuh D, Ben-Shlomo Y, Lynch J, Hallqvist J, Power C, 2003. Life course epidemiology. J. Epidemiol. Community Health 57, 778. [PubMed: 14573579]
- Lachman ME, Agrigoroaei S, Tun PA, Weaver SL, 2014. Monitoring cognitive functioning: psychometric properties of the brief test of adult cognition by telephone. Assessment 21, 404–17. 10.1177/1073191113508807 [PubMed: 24322011]
- Lantz PM, House JS, Mero RP, Williams DR, 2005. Stress, life events, and socioeconomic disparities in health: results from the Americans' Changing Lives Study. J. Health Soc. Behav. 46, 274–288. [PubMed: 16259149]
- Lind K, Edman Å, Nordlund A, Olsson T, Wallin A, 2007. Increased saliva cortisol awakening response in patients with mild cognitive impairment. Dement. Geriatr. Cogn. Disord. 24, 389–395. [PubMed: 17943022]
- Lindert J, Paul KC, Lachman Margie E, Ritz B, Seeman T, 2021. Social stress and risk of declining cognition: a longitudinal study of men and women in the United States. Soc. Psychiatry Psychiatr. Epidemiol. 1–10.
- Lynch KS, Lachman ME, 2020. The effects of lifetime trauma exposure on cognitive functioning in midlife. J. Trauma. Stress 33, 773–782. [PubMed: 32516491]
- Marin M-F, Lord C, Andrews J, Juster R-P, Sindi S, Arsenault-Lapierre G, Fiocco AJ, Lupien SJ, 2011. Chronic stress, cognitive functioning and mental health. Neurobiol. Learn. Mem. 96, 583– 595. [PubMed: 21376129]
- Montez JK, Hayward MD, 2014. Cumulative childhood adversity, educational attainment, and active life expectancy among US adults. Demography 51, 413–435. [PubMed: 24281740]
- Morris EP, Brown LL, Zaheed AB, Palms JD, Sol K, Martino A, Zahodne LB, 2021. Effects of Stress Exposure Versus Appraisal on Episodic Memory Trajectories: Evidence for Risk and Resilience Among Black Older Adults. J. Gerontol. Ser. B.
- Munoz E, Sliwinski MJ, Scott SB, Hofer S, 2015. Global perceived stress predicts cognitive change among older adults. Psychol. Aging 30, 487. [PubMed: 26121285]
- Munro CA, Wennberg AM, Bienko N, Eaton WW, Lyketsos CG, Spira AP, 2019. Stressful life events and cognitive decline: Sex differences in the Baltimore Epidemiologic Catchment Area Follow-Up Study. Int. J. Geriatr. Psychiatry 34, 1008–1017. [PubMed: 30901482]
- Murman DL, 2015. The Impact of Age on Cognition. Semin Hear 36, 111–21. 10.1055/ s-0035-1555115 [PubMed: 27516712]
- Naimi AI, Moodie EE, Auger N, Kaufman JS, 2014. Constructing inverse probability weights for continuous exposures: a comparison of methods. Epidemiology 292–299.
- Nishimi KM, Koenen KC, Coull BA, Chen R, Kubzansky LD, 2021. Psychological resilience predicting cardiometabolic conditions in adulthood in the Midlife in the United States Study. Proc. Natl. Acad. Sci. 118.
- Pearlin LI, 2010. The life course and the stress process: Some conceptual comparisons. J. Gerontol. B. Psychol. Sci. Soc. Sci. 65, 207–215.
- Pearlin LI, 1989. The sociological study of stress. J. Health Soc. Behav. 241–256. [PubMed: 2674272]
- Pearlin LI, Schieman S, Fazio EM, Meersman SC, 2005. Stress, health, and the life course: Some conceptual perspectives. J. Health Soc. Behav. 46, 205–219. [PubMed: 16028458]
- Robins JM, Hernan MA, Brumback B, 2000. Marginal structural models and causal inference in epidemiology.
- Rusanen M, Kivipelto M, Quesenberry CP, Zhou J, Whitmer RA, 2011. Heavy smoking in midlife and long-term risk of Alzheimer disease and vascular dementia. Arch. Intern. Med. 171, 333–339. [PubMed: 20975015]
- Ryff C, 2016. Midlife in the United States (MIDUS refresher), 2011-2014.
- Ryff CD, Magee WJ, Kling KC, Wing EH, 1999. Forging macro-micro linkages in the study of psychological well-being. Self Soc. Aging Process. 247, 278.
- Schuster TL, Kessler RC, Aseltine RH, 1990. Supportive interactions, negative interactions, and depressed mood. Am. J. Community Psychol. 18, 423–438. [PubMed: 2264558]

- Slopen N, Chen Y, Priest N, Albert MA, Williams DR, 2016. Emotional and instrumental support during childhood and biological dysregulation in midlife. Prev. Med. 84, 90–96. [PubMed: 26708307]
- Slopen N, Dutra LM, Williams DR, Mujahid MS, Lewis TT, Bennett GG, Ryff CD, Albert MA, 2012. Psychosocial stressors and cigarette smoking among African American adults in midlife. Nicotine Tob. Res. 14, 1161–1169. [PubMed: 22367977]
- Slopen N, Meyer C, Williams DR, 2018. 6 Cumulative Stress and Health, in: The Oxford Handbook of Integrative Health Science. Oxford Library of Psychology, p. 75.
- Stawski RS, Almeida DM, Lachman ME, Tun PA, Rosnick CB, Seeman T, 2011. Associations between cognitive function and naturally occurring daily cortisol during middle adulthood: timing is everything. J. Gerontol. B. Psychol. Sci. Soc. Sci. 66, i71–i81. [PubMed: 21743054]
- Sternthal MJ, Slopen N, Williams DR, 2011. Racial disparities in health: how much does stress really matter? Bois Rev. Soc. Sci. Res. Race 8, 95–113.
- Tun PA, Lachman ME, 2006. Telephone assessment of cognitive function in adulthood: the Brief Test of Adult Cognition by Telephone. Age Ageing 35, 629–32. 10.1093/ageing/afl095 [PubMed: 16943264]
- Turner AD, James BD, Capuano AW, Aggarwal NT, Barnes LL, 2017. Perceived stress and cognitive decline in different cognitive domains in a cohort of older African Americans. Am. J. Geriatr. Psychiatry 25, 25–34. [PubMed: 28231871]
- Turner RJ, Wheaton B, 1995. Checklist measurement of stressful life events. Meas. Stress Guide Health Soc. Sci. 29–58.
- Wainer H, 1976. Estimating coefficients in linear models: It don't make no nevermind. Psychol. Bull. 83, 213.
- Walen HR, Lachman ME, 2000. Social support and strain from partner, family, and friends: Costs and benefits for men and women in adulthood. J. Soc. Pers. Relatsh. 17, 5–30.
- Wheaton B, 1999. Social stress, in: Handbook of the Sociology of Mental Health. Springer, pp. 277–300.
- Wheaton B, 1994. Sampling the stress universe, in: Stress and Mental Health. Springer, pp. 77-114.
- Williams DR, Yu Y, Jackson JS, Anderson NB, 1997. Racial differences in physical and mental health: Socio-economic status, stress and discrimination. J. Health Psychol. 2, 335–351. [PubMed: 22013026]
- Winning A, Glymour MM, McCormick MC, Gilsanz P, Kubzansky LD, 2015. Psychological distress across the life course and cardiometabolic risk: findings from the 1958 British Birth Cohort Study. J. Am. Coll. Cardiol. 66, 1577–1586. [PubMed: 26429083]
- Yaffe K, Falvey CM, Hoang T, 2014. Connections between sleep and cognition in older adults. Lancet Neurol. 13, 1017–1028. [PubMed: 25231524]



**Figure 1.** Flowchart of Participant Selection, Midlife Development in the United States Study, 2004–2014

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#### Figure 2.

Directed Acyclic Graph showing the hypothesized causal relationships between childhood stress exposures, adulthood stress exposures, cognitive function, and confounders in the Midlife Development in the United States Study, 2004–2014.  $L_0$ : Age, gender, and race/ ethnicity status.  $L_1$ : Education, income, parental status, spousal status, working status, whether had chronic conditions or not.

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## Table 1

Baseline Characteristics by Tertile of the Cumulative Stress Exposure Score (N=3954), Midlife Development in the United States Study, 2004–2014<sup>a</sup>

Intercritic All participants (N= 395) Lowest (N= 131) Highest (N= 131) Highest (N= 131)   Age group, N(%) $66$ (yau) $001$ ( $71.3$ ) $0106$ ( $71.3$ ) $1121$ ( $90.4$ ) $66$ (yaus old $081$ ( $327.3$ ) $667$ ( $30.6$ ) $0106$ ( $71.1$ ) $1121$ ( $90.4$ ) $66$ (yaus old $2373$ ( $72.7$ ) $667$ ( $30.6$ ) $700$ ( $53.1$ ) $1241$ ( $90.4$ ) $-660$ ( $90.6$ ) $726$ ( $90.7$ ) $700$ ( $53.1$ ) $744$ ( $56.6$ ) $744$ ( $56.6$ ) $-600$ ( $900$ ) $723$ ( $86.6$ ) $723$ ( $86.7$ ) $700$ ( $53.1$ ) $744$ ( $56.6$ ) $-100$ ( $1000$ ( $1000$ ) $723$ ( $86.6$ ) $723$ ( $80.7$ ) $744$ ( $56.7$ ) $744$ ( $56.7$ ) $-1000$ ( $10000$ ( $10000$ ( $10000$ ( $10000$ ( $10000$ ( $10000$ ( $10000$ ( $1000$ ( $10000$ ( $10000$ ( $1000$			Tertile of Cumulat	ive Stress Exposure	Score
Age group, N(6) $- = 65$ years old $- < 65$ years old<	Characteristic	All participants $(N = 3954)$	Lowest (N=1318)	Middle ( <i>N</i> =1318)	Highest (N=1318)
·> 65 years old 108 (27.3) 66 (49.4) 30 (2.2.9) 17 (9.6)   - 65 years old 2873 (72.7) 667 (50.6) 10 (677.1) 11 19 (90.4)   Gender, N (%) - 2226 (56.3) 728 (3.3.7) 67 (30.6) 10 (677.1) 11 91 (90.4)   Gender, N (%) - 2226 (56.3) 728 (3.3.7) 67 (46.5) 744 (56.4)   - Hemde 1728 (43.7) 53 (40.7) 61 (46.5) 744 (56.4) 744 (56.4)   - White 346 (60.0) 127 (46.2) 123 (40.7) 103 (78.4) 103 (78.4)   - White 346 (60.0) 126 (40.7) 23 (17.6) 23 (17.6) 23 (17.6)   - White 346 (60.0) 126 (40.7) 24 (45.6) 23 (17.6) 23 (17.6)   - White 346 (60.0) 123 (40.7) 23 (17.6) 23 (17.6) 23 (17.6)   - White - White 346 (7.7) 35 (4.9) 34 (35.1) 34 (35.1) 34 (35.1)   - High school Gender or N (%) - High school 11 (19.2) 34 (25.1) 34 (25.1) 34 (25.1) <td>Age group, <math>N(\%)</math></td> <td></td> <td></td> <td></td> <td></td>	Age group, $N(\%)$				
$- \cdot 65 yars old$ $- \cdot 65 yars old$ $1016(71.1)$ $191(90.4)$ $Cender, N(\%)$ $- Female$ $225(56.3)$ $782(9.3)$ $700(53.1)$ $74(56.4)$ $- Male$ $- Nale$ $2226(56.3)$ $782(9.3)$ $74(56.4)$ $74(56.4)$ $- Male$ $- Nale$ $- 1728(43.7)$ $538(40.7)$ $618(46.9)$ $574(43.6)$ $- Race, N(\%)$ $- 342(8.6)$ $1224(92.9)$ $116(688.5)$ $033(78.4)$ $- Race, N(\%)$ $- 342(8.6)$ $- 396(10.0)$ $65(4.9)$ $97(7.5)$ $232(17.6)$ $- Wale$ $- 396(10.0)$ $65(4.9)$ $97(7.5)$ $232(17.6)$ $- Race, N(\%)$ $- 396(10.0)$ $65(4.9)$ $97(7.5)$ $232(17.6)$ $- Wale$ $- 396(10.0)$ $65(4.9)$ $97(7.5)$ $232(17.6)$ $- Wale$ $- 396(10.0)$ $65(4.9)$ $97(7.5)$ $237(17.6)$ $- Race, N(\%)$ $- 123(18.4)$ $- 29(2.2)$ $37(4.0)$ $37(4.0)$ $- Hale stool GED- 104(27.6)21(9.2)37(4.0)37(4.0)- Hale stool GED- 104(27.6)21(9.2)37(4.0)37(4.0)- Hale stool GED- 104(27.6)- 29(2.2)37(4.0)37(4.0)- Some ollege- 104(27.6)- 29(2.2)37(2.6)37(2.6)- Some ollege- 104(27.6)- 23(22.1)37(2.6)37(2.6)- Hale stool on GED- 106(1900)- 106(1900)37(2.6)37(2.6)- Les han High school or GED- 176(7.1)- 236(19.0)317(2.6)$	- >= 65 years old	1081 (27.3)	651 (49.4)	302 (22.9)	127 (9.6)
Gender, N(%) $700 (53.1)$ $740 (54.3)$ $\cdot$ Female $\cdot$ Female $1728 (43.7)$ $536 (40.7)$ $618 (46.9)$ $741 (56.4)$ $\cdot$ Male $\cdot$ N(%) $3423 (86.6)$ $732 (4.2)$ $741 (56.4)$ $536 (4.2)$ $536 (4.2)$ $537 (4.2)$ Race, N(%) $= 3423 (86.6)$ $1224 (92.9)$ $1166 (88.5)$ $1033 (78.4)$ $536 (10)$ $56 (4.9)$ $90 (7.5)$ $232 (17.6)$ Race, N(%) $= 346 (10)$ $65 (4.9)$ $90 (7.5)$ $537 (4.0)$ $537 (4.0)$ $536 (10)$ $53 (4.0)$ $347 (25.6)$ $\cdot$ Huher's or mother's highest levels of education, N(%) $1104 (27.9)$ $1104 (27.9)$ $414 (31.4)$ $341 (26.1)$ $537 (4.0)$ $\cdot$ Huher's or mother's highest levels of education, N(%) $1104 (27.9)$ $237 (25.0)$ $337 (25.6)$ $337 (25.6)$ $337 (25.6)$ $\cdot$ Huher's or mother's highest levels of education, N(%) $1104 (27.9)$ $1104 (27.9)$ $341 (26.1)$ $347 (25.6)$ $\cdot$ Some college $384 (9.7)$ $337 (25.6)$ $337 (25.6)$ $337 (25.6)$ $337 (25.6)$ $\cdot$ Some college $1084 (77.4)$ $358 (72.2)$ $337 (25.6)$ $338 (29.4)$ $\cdot$ Les than High school $-108 (77.4)$ $235 (64.7)$ $237 (25.6)$ $337 (25.6)$ $\cdot$ Huher school or GED $235 (64.7)$ $237 (25.6)$ $337 (25.6)$ $337 (25.6)$ $\cdot$ Huhe school or GED $-108 (77.4)$ $236 (79.7)$ $347 (25.6)$ $337 (25.6)$ $\cdot$ Huhe school or GED $-108 (77.4)$ $236 (79.7)$ $347 (25.7)$ $347 (25.6)$ <tr< td=""><td>- &lt; 65 years old</td><td>2873 (72.7)</td><td>667 (50.6)</td><td>1016(77.1)</td><td>1191 (90.4)</td></tr<>	- < 65 years old	2873 (72.7)	667 (50.6)	1016(77.1)	1191 (90.4)
- Female- Temale- Tema	Gender, $N(\%)$				
· Male· Ma	- Female	2226 (56.3)	782 (59.3)	700 (53.1)	744 (56.4)
Race, $N(8)$ Interference $3423 (86.6)$ $1224 (92.9)$ $1166(88.5)$ $1033 (78.4)$ - White $396 (10.0)$ $65 (4.9)$ $90 (7.5)$ $232 (17.6)$ - Black $396 (10.0)$ $65 (4.9)$ $90 (7.5)$ $233 (4.0)$ - Black $396 (10.0)$ $65 (4.9)$ $90 (7.5)$ $233 (4.0)$ - Inters $1135 (3.4)$ $29 (2.2)$ $33 (4.0)$ $53 (4.0)$ - Less than High school $1104 (27.9)$ $1104 (27.9)$ $414 (31.4)$ $344 (26.1)$ - High school or GED $384 (97.1)$ $100 (27.6)$ $342 (32.1)$ $342 (32.1)$ - Some college $107 (27.7)$ $340 (25.0)$ $323 (32.0)$ - Some college $1057 (25.7)$ $380 (25.6)$ $342 (25.1)$ $316 (24.0)$ - College or more $1057 (25.7)$ $350 (25.6)$ $327 (25.7)$ $316 (24.0)$ - Less than High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - Less than High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - Less than S25 (00 $1084 (27.4)$ $358 (27.2)$ $327 (29.7)$ $313 (32.7)$ - Less than S25 (00 $1084 (27.4)$ $253 (10.6)$ $221 (10.6)$ $327 (20.7)$ - Less than S25 (00 $237 (35.6)$ $237 (25.6)$ $237 (25.6)$ $338 (22.9)$ - Less than S25 (00 $1084 (27.4)$ $237 (15.6)$ $237 (25.7)$ $327 (20.7)$ - Less than S25 (00 $237 (30.0)$ $237 (35.6)$ $237 (30.0)$ $237 (20.7)$ - Less than S25 (000 $237$	- Male	1728 (43.7)	536 (40.7)	618 (46.9)	574 (43.6)
· White· White· Mathematical mathem	Race, $N(\%)$				
· Black $36(100)$ $65(4,9)$ $9(7,5)$ $232(17,6)$ · Others· I35 (3,4) $29(2,2)$ $53(4,0)$ $53(4,0)$ Father's number's highest levels of education. $N(\%)$ $113(3,4)$ $29(2,2)$ $53(4,0)$ Father's or mother's highest levels of education. $N(\%)$ $1104(27,9)$ $414(31,4)$ $344(26,1)$ $53(4,0)$ - Less than High school $1104(27,9)$ $1104(27,9)$ $141(31,4)$ $344(26,1)$ $53(3,7)$ - High school $384(9,7)$ $1104(27,9)$ $110(9,9)$ $121(9,2)$ $133(10,1)$ - Some college $384(9,7)$ $332(26,6)$ $332(29,7)$ $316(24,0)$ - Some college $1057(26,7)$ $358(26,6)$ $332(29,7)$ $316(24,0)$ - College or more $1057(26,7)$ $358(72,2)$ $337(25,6)$ $338(29,4)$ - Less than High school $255(6,4)$ $67(5,2)$ $78(5,8)$ $310(23,7)$ - Less than High school $1084(27,4)$ $358(72,2)$ $337(25,6)$ $338(29,4)$ - Less than High school $1084(27,4)$ $358(72,2)$ $337(25,6)$ $338(29,4)$ - Less than S25,000 $544,999$ $737(18,6)$ $237(19,6)$ $302(29,1)$ $302(29,1)$ - Less than S25,000 $544,999$ $732(18,5)$ $251(19,0)$ $302(29,1)$ $302(29,1)$ - Less than S25,000 $544,999$ $732(18,5)$ $251(19,0)$ $302(29,1)$ $302(29,1)$ - Less than S25,000 $544,999$ $732(18,5)$ $251(19,0)$ $202(19,0)$ $202(19,0)$ - Less than S25,000 $543,999$	- White	3423 (86.6)	1224 (92.9)	1166(88.5)	1033 (78.4)
· Others· 135 (3.4) $29 (2.2)$ $53 (4.0)$ $53 (4.0)$ Father's or mother's highest levels of education, $N(\%)$ · 104 (279)· 414 (31.4)· 347 (26.1)· 347 (26.3)· Less than High school· 109 (279)· 104 (379)· 423 (32.1)· 462 (35.1)· 323 (397)· High school or GED· 384 (9.7)· 131 (9.9)· 121 (9.2)· 133 (10.1)· Some college· 384 (9.7)· 330 (26.6)· 392 (29.7)· 316 (24.0)· College or more· 1057 (26.7)· 350 (26.6)· 392 (29.7)· 316 (24.0)· College or more· 1057 (26.7)· 350 (26.6)· 322 (29.7)· 313 (23.7)· College or more· 1057 (26.7)· 353 (27.2)· 337 (25.6)· 338 (29.4)· Less than High school or GED· 108 (27.4)· 358 (27.2)· 337 (25.6)· 338 (29.4)· High school or GED· 108 (27.4)· 358 (27.2)· 337 (25.6)· 338 (29.4)· East than High school or GED· 106 (27.4)· 244 (25.9)· 244 (25.9)· 244 (25.9)· College or more· 1767 (37.1)· 639 (41.8)· 621 (39.5)· 244 (29.9)· 244 (29.9)· College or more· 1767 (37.1)· 639 (41.8)· 244 (29.9)· 244 (29.9)· 244 (29.9)· College or more· 1066 (20.0)· 244 (29.9)· 244 (29.9)· 244 (29.9)· 244 (29.9)· College or more· 1267 (37.1)· 232 (19.0)· 239 (19.0)· 239 (19.0)· 239 (20.0)· Less than \$55,000· 244 (29.9)· 244 (29.9)· 244 (29.9)	- Black	396 (10.0)	65 (4.9)	99 (7.5)	232 (17.6)
Father's or mother's highest levels of education, $N(\%)$ $1104 (27.9)$ $341 (3.1.4)$ $341 (2.6.1)$ $347 (2.6.3)$ - Less than High school1104 (27.9) $1104 (27.9)$ $131 (9.9)$ $341 (2.6.1)$ $347 (2.6.3)$ - High school or GED $384 (9.7)$ $384 (9.7)$ $131 (9.9)$ $1121 (9.2)$ $133 (10.1)$ - Some college $384 (9.7)$ $384 (9.7)$ $350 (2.6.6)$ $392 (29.7)$ $316 (24.0)$ - College or more $1057 (2.6.7)$ $357 (2.6.7)$ $350 (2.6.6)$ $392 (29.7)$ $316 (24.0)$ Participant's education level, $N(\%)$ $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - Less than High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $338 (29.4)$ - High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $338 (29.4)$ - Less than High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $338 (29.4)$ - Less than S25 000 $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $338 (29.4)$ - College or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $337 (25.6)$ $337 (25.6)$ - Less than S25 000 $900 (22.8)$ $250 (19.0)$ $291 (16.6)$ $263 (200)$ - Less than S25 000 $544,999$ $732 (18.5)$ $250 (19.0)$ $251 (19.0)$ $251 (19.0)$ - Less than S25 000 $544,999$ $732 (18.5)$ $250 (19.0)$ $251 (19.0)$ $253 (20.0)$ - Loss than S70 000 $544,999$ $732 (18.5)$ $251 (16.8)$ $251 (16.8)$ $253 (20.0)$ <td< td=""><td>- Others</td><td>135 (3.4)</td><td>29 (2.2)</td><td>53 (4.0)</td><td>53 (4.0)</td></td<>	- Others	135 (3.4)	29 (2.2)	53 (4.0)	53 (4.0)
- Less than High school1104 (27.9)414 (31.4)344 (26.1)347 (26.3)- High school or GED1408 (35.6)423 (32.1)462 (35.1)523 (39.7)- Some college384 (9.7)131 (9.9)121 (9.2)133 (10.1)- Some college or more1057 (26.7)350 (26.6)392 (29.7)316 (24.0)- Tricipant's education level, $N(%)$ 255 (6.4) $67 (5.2)$ 78 (5.8)110 (8.3)- High school255 (6.4) $67 (5.2)$ 78 (5.8)110 (8.3)- Less than High school255 (6.4) $67 (5.2)$ 78 (5.8)313 (23.7)- High school or GED849 (29.0)254 (25.9)337 (25.6)338 (29.4)- High school or GED849 (29.0)254 (55.9)337 (25.6)313 (23.7)- Use college1084 (77.4)358 (27.2)337 (25.6)313 (23.7)- Use college or more1767 (37.1)639 (41.8)621 (39.5)313 (23.7)- College or more1767 (37.1)639 (41.8)621 (39.5)302 (23.9)- Less than \$55,000900 (22.8)347 (26.3)251 (19.0)302 (23.9)- Less than \$55,000732 (18.5)250 (19.0)219 (16.6)263 (20.0)- Less than \$57,000544,999732 (18.5)251 (19.0)251 (19.0)263 (20.0)- Less than \$57,000544,999732 (18.5)251 (19.0)253 (20.0)263 (20.0)- Less than \$57,0001543 (39.0)500 (37.9)575 (43.6)263 (20.0)- More than \$70,0001543 (39.0)500 (37.9)5	Father's or mother's highest levels of education, $N(\%)$				
- High school or GED1408 (35.6)123 (32.1)423 (35.1)523 (39.7)- Some college $384 (9.7)$ $131 (9.9)$ $121 (9.2)$ $133 (10.1)$ - College or more $384 (9.7)$ $350 (26.6)$ $392 (29.7)$ $316 (24.0)$ Participant's education level. $N(\%)$ $1057 (26.7)$ $350 (26.6)$ $392 (29.7)$ $110 (8.3)$ - Less than High school $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - Less than High school $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - Less than High school $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $313 (23.7)$ - Less than High school $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - High school or GED $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ - Some college $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ - College or more $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ - Los than $$55,000$ $530 (41.8)$ $621 (39.5)$ $507 (38.5)$ - Les than $$55,000$ $530 (10.8)$ $273 (10.9)$ $273 (10.9)$ $263 (20.0)$ - Les than $$55,000$ $730 (10.5)$ $271 (16.8)$ $273 (20.7)$ $263 (20.0)$ - Les than $$55,000$ $590 (9999$ $730 (19.7)$ $271 (16.8)$ $273 (20.7)$ $263 (20.0)$ - More than $$70,000$ $1543 (39.0)$ $573 (37.9)$ $573 (20.7)$ $263 (20.1)$	- Less than High school	1104 (27.9)	414 (31.4)	344 (26.1)	347 (26.3)
- Some college $384 (9.7)$ $131 (9.9)$ $121 (9.2)$ $133 (10.1)$ - College or more $1057 (26.7)$ $350 (26.6)$ $392 (29.7)$ $316 (24.0)$ Participant's education level, $N(\%)$ $1057 (26.7)$ $350 (26.6)$ $392 (29.7)$ $316 (24.0)$ Participant's education level, $N(\%)$ $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - High school or GED $255 (6.4)$ $67 (5.2)$ $78 (29.1)$ $318 (29.4)$ - Some college $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - Some college $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ - College or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ - Muthood household income, $N(\%)$ $900 (22.8)$ $347 (26.3)$ $251 (19.0)$ $302 (22.9)$ - Less than $825,000$ $535,000$ $732 (18.5)$ $271 (16.8)$ $251 (19.0)$ $202 (29.0)$ - S55,000 to $869.999$ $732 (18.5)$ $273 (20.7)$ $263 (20.7)$ $263 (20.7)$ - More than $870,000$ $1543 (39.0)$ $500 (37.9)$ $575 (43.6)$ $467 (35.4)$	- High school or GED	1408 (35.6)	423 (32.1)	462 (35.1)	523 (39.7)
- College or more $1057(26.7)$ $350(26.6)$ $392(29.7)$ $316(24.0)$ Participant's education level, $N(\%)$ $255(6.4)$ $67(5.2)$ $78(5.8)$ $110(8.3)$ - Less than High school $255(6.4)$ $67(5.2)$ $78(5.8)$ $110(8.3)$ - High school or GED $1084(27.4)$ $358(27.2)$ $337(25.6)$ $388(29.4)$ - Some college $1964(27.4)$ $358(27.2)$ $337(25.6)$ $388(29.4)$ - Some college $1767(37.1)$ $639(41.8)$ $621(39.5)$ $507(38.5)$ - College or more $1767(37.1)$ $639(41.8)$ $621(39.5)$ $507(38.5)$ - Multhood household income, $N(\%)$ $900(22.8)$ $347(26.3)$ $21(19.0)$ $302(22.9)$ - Less than $525,000$ $536,999$ $732(18.5)$ $250(19.0)$ $219(16.6)$ $263(20.0)$ - S45,000 to $569,999$ $780(19.7)$ $211(16.8)$ $273(20.7)$ $286(21.7)$ - More than $570,000$ $570(300)$ $570(37.9)$ $575(43.6)$ $467(35.4)$	- Some college	384 (9.7)	131 (9.9)	121 (9.2)	133 (10.1)
Participant's education level, $N(\%)$ $2455 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - Less than High school $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - High school or GED $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - Some college $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ - College or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ - College or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ - Less than $525 (000$ $900 (22.8)$ $347 (26.3)$ $210 (16.6)$ $202 (22.9)$ - Less than $525 (000$ $536 (44.999$ $732 (18.5)$ $251 (19.0)$ $302 (22.9)$ - Less than $525 (000$ $544.999$ $732 (18.5)$ $251 (16.6)$ $263 (20.0)$ - More than $570 (000 (569.999)$ $733 (18.5)$ $271 (16.8)$ $273 (20.7)$ $286 (21.7)$ - More than $570 (000$ $500 (37.9)$ $577 (43.6)$ $467 (35.4)$	- College or more	1057 (26.7)	350 (26.6)	392 (29.7)	316 (24.0)
- Less than High school $255 (6.4)$ $67 (5.2)$ $78 (5.8)$ $110 (8.3)$ - High school or GED $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ - Some college $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ - Some college or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ Adulthood household income, $N(\%)$ $900 (22.8)$ $347 (26.3)$ $251 (19.0)$ $302 (22.9)$ - Less than \$55,000 $732 (18.5)$ $250 (19.0)$ $219 (16.6)$ $263 (20.0)$ - S55,000 to \$69,999 $732 (18.5)$ $221 (16.8)$ $273 (20.7)$ $286 (21.7)$ - More than \$70,000 $1543 (39.0)$ $500 (37.9)$ $575 (43.6)$ $467 (35.4)$	Participant's education level, $N(\%)$				
High school or GED $1084 (27.4)$ $358 (27.2)$ $337 (25.6)$ $388 (29.4)$ Some college $849 (29.0)$ $849 (29.0)$ $254 (25.9)$ $282 (29.1)$ $313 (23.7)$ College or more $1767 (37.1)$ $639 (41.8)$ $621 (39.5)$ $507 (38.5)$ Adulthood household income, $N(\%)$ $900 (22.8)$ $347 (26.3)$ $251 (19.0)$ $302 (22.9)$ Less than $525,000$ to $$44,999$ $732 (18.5)$ $250 (19.0)$ $219 (16.6)$ $263 (20.0)$ $545,000$ to $569,999$ $780 (19.7)$ $221 (16.8)$ $273 (20.7)$ $286 (21.7)$ Ome than $$70,000$ $1533 (39.0)$ $500 (37.9)$ $575 (43.6)$ $467 (35.4)$	- Less than High school	255 (6.4)	67 (5.2)	78 (5.8)	110 (8.3)
$\begin{array}{l l l l l l l l l l l l l l l l l l l $	- High school or GED	1084 (27.4)	358 (27.2)	337 (25.6)	388 (29.4)
	- Some college	849 (29.0)	254 (25.9)	282 (29.1)	313 (23.7)
Adulthood household income, $N(\%)$ 900 (22.8) $347$ (26.3) $251$ (19.0) $302$ (22.9)- Less than \$25,000 $532,000$ $732$ (18.5) $250$ (19.0) $219$ (16.6) $263$ (20.0)- \$45,000 to \$69,999 $780$ (19.7) $221$ (16.8) $273$ (20.7) $286$ (21.7)- More than \$70,000 $1543$ (39.0) $500$ (37.9) $575$ (43.6) $467$ (35.4)	- College or more	1767 (37.1)	639 (41.8)	621 (39.5)	507 (38.5)
- Less than \$25,000900 (22.8) $347 (26.3)$ $251 (19.0)$ $302 (22.9)$ - \$25,000 to \$44,999 $732 (18.5)$ $250 (19.0)$ $219 (16.6)$ $263 (20.0)$ - \$45,000 to \$69,999 $780 (19.7)$ $221 (16.8)$ $273 (20.7)$ $286 (21.7)$ - More than \$70,000 $1543 (39.0)$ $500 (37.9)$ $575 (43.6)$ $467 (35.4)$	Adulthood household income, $N(\%)$				
- \$25,000 to \$44,999 732 (18.5) 250 (19.0) 219 (16.6) 263 (20.0)   - \$45,000 to \$69,999 780 (19.7) 221 (16.8) 273 (20.7) 286 (21.7)   - More than \$70,000 1543 (39.0) 500 (37.9) 575 (43.6) 467 (35.4)	- Less than \$25,000	900 (22.8)	347 (26.3)	251 (19.0)	302 (22.9)
- \$45,000 to \$69,999 780 (19.7) 221 (16.8) 273 (20.7) 286 (21.7)   - More than \$70,000 1543 (39.0) 500 (37.9) 575 (43.6) 467 (35.4)	- \$25,000 to \$44,999	732 (18.5)	250 (19.0)	219 (16.6)	263 (20.0)
- More than \$70,000 1543 (39.0) 500 (37.9) 575 (43.6) 467 (35.4)	- \$45,000 to \$69,999	780 (19.7)	221 (16.8)	273 (20.7)	286 (21.7)
	- More than \$70,000	1543 (39.0)	500 (37.9)	575 (43.6)	467 (35.4)

		Tertile of Cumulat	ive Stress Exposure	Score
Characteristic	All participants $(N = 3954)$	Lowest (N=1318)	<b>Middle</b> ( <i>N</i> =1318)	Highest (N=1318)
- Yes	2564 (64.9)	520 (39.5)	973 (73.8)	1071 (81.3)
- No	1390 (35.2)	798 (60.5)	345 (26.3)	247 (18.7)
Currently married, $N(\%)$				
- Yes	2872 (72.6)	974 (73.9)	992 (75.3)	906 (68.7)
- No	1082 (27.4)	344 (26.1)	326 (24.7)	412 (31.3)
Has 1+ child, $N(\%)$				
- Yes	3464 (87.6)	1160 (88.1)	1153 (87.5)	1151 (87.3)
- No	490 (12.4)	158 (11.9)	165 (12.5)	167 (12.7)
Presence of chronic conditions, $N(\%)$				
- Yes	3100 (78.4)	1020 (77.4)	1003 (76.1)	1076 (81.6)
- No	854 (21.6)	298 (22.6)	315 (23.9)	242 (18.4)
Abbreviation: GED: General Educational Diploma; SD: 5	standard Deviation.			

 $^{a}$ Results were generated from 20 imputed datasets.

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# Table 2.

Adjusted Mean Difference in Baseline Cognitive Function Score per SD unit Increment in Cumulative Stress Exposure, and Per Year in Baseline Age, in the Midlife Development in the United States Study a

	Executive <b>]</b>	Junction Score	Episodi	c Memory Score
	Difference,	SD units (95% CI)	Difference,	SD units (95% CI)
All Participants (N=3,954)				
Per SD in Cumulative Stress Exposure	-0.12	-0.16, -0.08	-0.09	-0.13, -0.05
Per year in Baseline age	-0.03	-0.04, -0.03	-0.03	-0.03, -0.02
Workers only (N=2,564)				
Per SD in Cumulative Stress Exposure	-0.10	-0.14, -0.05	-0.06	-0.11, -0.01
Per year in Baseline age	-0.03	-0.033, -0.026	-0.02	-0.024, -0.016

Abbreviation: CI: confidence interval; SD: standard deviation

 $^{a}$ Multiple imputation results are summarized from 20 imputed data sets.

b Estimates are from marginal structural models and were additionally adjusted for childhood confounders including age, gender, race/ethnicity status, and parent's education levels. Stabilized inverse probability of treatment weights were included to adjust for "stressor-confounder" feedback. Author Manuscript

Adjusted Mean Difference in Baseline Cognitive Function score, by Life-course Stressor Profile, in the Midlife in the United States Study<sup>a</sup>

			Executive Functi	on Score	Episodic Memor	y Score
Childhood Stress Exposure <sup>b</sup>	Adulthood Stress Exposure $^{b}$	N	Difference, SD units	95% CI	Difference, SD units	95% CI
All Participants (N=3,954)						
Lower	Lower	2491	0, Ref		0, Ref	
High	Lower	475	-0.22	-0.31, -0.12	-0.17	-0.27, -0.07
Lower	High	6969	-0.13	-0.21, -0.05	-0.13	-0.22, -0.05
High	High	292	-0.34	-0.46, -0.22	-0.21	-0.33, -0.08
<i>Workers only (N=2,564)</i>						
Lower	Lower	1489				
High	Lower	247	-0.17	-0.30, -0.04	-0.14	-0.27, -0.02
Lower	High	589	-0.09	-0.19,0.02	-0.08	-0.20.0.04
High	High	239	-0.38	,-0.52,-0.23	-0.14	-0.31,0.04

<sup>a</sup>Multiple imputation results are from 20 imputed data sets. Generalized estimating equations (with normal distribution and identity link) were used to estimate the associations of cumulative stressor profiles with cognitive function, adjusting for clustering by sibling status.

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b. We designated the top quartile of the z-score of childhood and adulthood stress exposures as "high stress," and the lowest and the 2nd quartile of the z-score of childhood and adulthood stress exposures as "lower stress."

<sup>c</sup> Estimates are from marginal structural models and were additionally adjusted for childhood confounders included age, gender, race/ethnicity status, and parent's education levels. Stabilized inverse probability of treatment weights were included to adjust for "stressor-confounder" feedback. Author Manuscript

# Table 4.

Adjusted Mean Differences in Changes in Cognitive Function, by Baseline Cumulative Stress Exposures Score, in the Midlife Development in the United States Study, 2004–2014<sup>a</sup>

	Mean differences in changes in executive function scores	Mean differences in changes in episodic memory scores
	Difference, SD units, (95% CI)	Difference, SD units, (95% CI)
All Participants (N= 2,517)	0.004 (-0.0004, 0.01)	-0.00002(-0.01,0.01)
Workers only $(N=1,754)$	0.003 (-0.01,0.01)	0.01 (-0.01,0.03)

Abbreviation: CI: confidence interval.

stress, and the cross-products of time with covariates. Stabilized inverse probability of weights were included to account for "stressor-confounder" feedback, lost to follow up due to death, and lost to follow <sup>a</sup>Multiple imputation results were summarized from 20 imputed data sets. Models included age, race/ethnicity status, parent's education, cumulative stress exposures, time, the cross-products of time with up due to other reasons.