



Lower Cardiovascular Reactivity Is Associated With More Childhood Adversity and Poorer Midlife Health: Replicated Findings From the Dunedin and MIDUS Cohorts

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Abstract

Cardiovascular reactivity has been proposed as a biomarker linking childhood adversity and poorer health. In the current study, we examined the association of childhood adversity, cardiovascular reactivity, and health in the Dunedin Multidisciplinary Health and Development Study ($n = 922$) and Midlife in the United States (MIDUS) studies ($n = 1,015$). In both studies, participants who experienced more childhood adversity had lower cardiovascular reactivity. In addition, people with lower cardiovascular reactivity had poorer self-reported health and greater inflammation. Dunedin participants with lower cardiovascular reactivity were aging biologically faster, and MIDUS participants with lower heart rate reactivity had increased risk of early mortality. Cardiovascular reactivity was not associated with hypertension in either study. Results were partially accounted for by greater reactivity among more conscientious, less depressed, and higher functioning participants. These results suggest that people who experienced childhood adversity have a blunted physiological response, which is associated with poorer health. The findings highlight the importance of accounting for individual differences when assessing cardiovascular reactivity using cognitive stressor tasks.

Keywords

cardiovascular reactivity, blood pressure, heart rate, health, well-being, cognition

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To better understand why some people are more likely than others to develop physical health problems, clinical psychological scientists often use physiological biomarkers to try to capture psychological vulnerabilities. In particular, there is a long tradition of using measures of cardiovascular function in clinical psychology to assist in the dual purposes of inferring psychological states and linking these states with health outcomes. The assessment of people's cardiovascular physiological response to a stressor—termed *cardiovascular reactivity* (Allen & The Psychosocial Working Group, 2000)—has been used productively for both purposes.

The cardiovascular-reactivity hypothesis (Blascovich & Katkin, 1993; Manuck, 1994; Treiber et al., 2003) arose from the observation that people with hypertension seemed to evidence greater increases in their blood pressure during cold pressor tasks compared with people without hypertension (Manuck, 1994). If people with higher levels of physiological reactivity in response

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to lab stressors were more likely to respond in an exaggerated way to stressors in their environment, it was thought that this could result in poorer cardiovascular health over time through atherosclerotic processes and damage to the cardiovascular system broadly (Manuck, 1994; Treiber et al., 2003). Empirical evidence began to provide support for the cardiovascular-reactivity hypothesis, which was summarized in a 2010 meta-analysis of 36 studies that found that people with higher cardiovascular reactivity were more likely to have cardiovascular risk factors in the form of high blood pressure and hypertension (Chida & Steptoe, 2010), particularly in the case of reactivity to cognitive task stressors. People with higher cardiovascular reactivity were also shown to have greater risk of mortality (Carroll, Ginty, et al., 2012).

During the early 2000s, life-course researchers proposed an extension of the cardiovascular-reactivity hypothesis as a potential mechanism to explain the well-established association between childhood adversity and poorer health in adulthood (Allen & The Psychosocial Working Group, 2000; Cohen et al., 2010). Children who experience adverse childhood experiences (ACEs) and socioeconomic disadvantage are more likely to have poorer health in later life (Cohen et al., 2010; Galobardes et al., 2004; Moffitt, 2013). Alterations in cardiovascular reactivity were proposed as one way by which adversity could become biologically embedded and result in poorer health. The MacArthur SES & Health Network's psychosocial notebook, for example, highlighted the possibility that "exposure to a more threatening or challenging environment by lower SES individuals results in greater reactivity in various organ systems in response to this exposure" (Allen & The Psychosocial Working Group, 2000). Over time, these exaggerated responses could then translate to poorer health outcomes, particularly in the cardiovascular system. This possibility, in part, prompted the adoption of cardiovascular reactivity as a physiological biomarker that might help explain how childhood adversity could affect health in longitudinal cohort studies, such as the Dunedin Multidisciplinary Health and Development Study (Poulton et al., 2015).

As the cardiovascular-reactivity hypothesis was applied more broadly to areas of health outside the cardiovascular system, a new set of findings began to challenge the traditional paradigm linking increased reactivity to poorer health. The aptly titled article "Are Large Physiological Reactions to Acute Stress Always Bad for Health?" Carroll et al. (2009) reviewed empirical evidence that *lower* cardiovascular reactivity was associated with poorer health outcomes outside the cardiovascular system, including obesity (Carroll et al., 2007) and poorer self-reported health (Phillips, 2011). Concurrently, a number of studies also found that trauma

and childhood adversity were associated with lower levels of cardiovascular reactivity (McLaughlin et al., 2014, 2015). These results were organized by researchers into a model suggesting that lower reactivity indexed dysregulation in the fronto-limbic system responsible for marshaling resources for motivated behavior (Carroll et al., 2017; Ginty et al., 2013). This motivational dysregulation was posited to result in poorer health because of differences in health behaviors, such as increased rates of smoking and obesity (Phillips et al., 2013), rather than—or perhaps in addition to—direct cardiovascular pathophysiological mechanisms, such as atherosclerosis. These findings became the basis of the blunting hypothesis of cardiovascular reactivity. These competing models raise important questions: Is lower cardiovascular reactivity in the lab a marker of people's psychological characteristics, specifically fronto-limbic dysregulation related to motivated behavior, or does greater reactivity hold direct relevance to cardiovascular physiological functioning and later health?

Cardiovascular reactivity has been measured in response to a wide array of tasks in the lab, and it seems unlikely that reactivity to different stressful tasks—such as the Trier Social Stress Test, trauma recalls, cold pressors, and cognitive stressors such as serial subtraction and Stroop tasks—all reflect the same "reactivity" (Manuck, 1994). Such measurement issues would be particularly relevant for stressors that might be more or less engaging or stressful for people because of the individual characteristics they bring into the lab while performing cardiovascular lab protocols. For example, lower cardiovascular reactivity is associated with less conscientiousness (Sesker, 2019) and lower cognitive ability (Carroll et al., 2017) as well as depression (Salomon et al., 2009). People who are more conscientious or have better cognitive ability might engage more fully with cognitive tasks, which could affect their cardiovascular reactivity to such tasks. Likewise, people who are depressed might be less willing or able to fully engage with cognitive tasks. Accounting for participants' individual differences in such measures could provide additional context important to interpreting the associations between childhood adversity, cardiovascular reactivity, and health.

Present Study

In the present study, we report the associations between childhood adversity, health, and cardiovascular reactivity to two commonly studied cognitive stressors: the Stroop task and a mental arithmetic task. This study evolved over several stages that are relevant to interpreting the findings. In 2005, we introduced a cardiovascular reactivity protocol into the Dunedin longitudinal study (Poulton et al., 2015) when participants were 32 years

old to investigate prospectively measured childhood adversity and adult cardiovascular reactivity. Resulting data yielded associations between childhood adversity and cardiovascular reactivity in the opposite direction of what was expected: People with more adverse childhoods had *lower* levels of cardiovascular reactivity at age 32. At the time, we did not publish these results because they ran counter to the contemporary understanding of the cardiovascular-reactivity hypothesis, and we were uncertain what to make of them. Now, 13 years later, we have collected health outcome data in the Dunedin study that enabled us to investigate the association between cardiovascular reactivity at age 32 and health at age 45. In addition, we identified data from the Midlife in the United States (MIDUS) study (Brim et al., 1994; Ryff et al., 2017) from a cardiovascular reactivity protocol administered during 2004 to 2006 that used the same cognitive stressors as those in the Dunedin study. This allowed us to replicate our results in a second data set and ensure robustness of findings that had initially puzzled us in the mid-2000s.

This process resulted in the present report, which used cardiovascular reactivity data from the Dunedin study ($n = 922$) and MIDUS study ($n = 1,015$) to examine the association of childhood adversity, cardiovascular reactivity, and health in two longitudinal studies. In particular, we were interested in examining reactivity in blood pressure and heart rate as indicators of cardiovascular reactivity. The original grant proposal in the Dunedin study hypothesized—in line with the cardiovascular-reactivity hypothesis—that people with more adversity in childhood would have greater cardiovascular reactivity and that this would explain their poorer adult health. In the years following the data collection in Dunedin, new evidence regarding cardiovascular blunting and health presented an alternative possibility, that lower cardiovascular reactivity, rather than higher, might be associated with poorer health. To provide additional context to our primary results related to childhood adversity, cardiovascular reactivity, and health, we conducted secondary analyses examining whether individual differences in psychological characteristics—specifically conscientiousness, depression, and cognitive ability—could help explain responses to the two cognitive stressful tasks.

Method

Participants and study design

Participants were drawn from two longitudinal studies, the Dunedin study (Poulton et al., 2015) and the MIDUS study (Brim et al., 1994; Ryff et al., 2017).

Dunedin study. The Dunedin Multidisciplinary Health and Development Study is a longitudinal investigation of

health and behavior in a representative birth cohort. The 1,037 study members (91% of eligible births; 52% male) were all individuals born between April 1972 and March 1973 in Dunedin, New Zealand, who were eligible on the basis of residence in the province and who participated in the first assessment at age 3 years (Poulton et al., 2015). The cohort represented the full range of socioeconomic status (SES) in the general population of New Zealand's South Island. As adults, the cohort matches the results from the New Zealand National Health and Nutrition Survey on key adult health indicators (Poulton et al., 2015). The cohort also matches the distribution of educational attainment of citizens of the same age from the New Zealand census (Richmond-Rakerd et al., 2020). The cohort is predominantly White (93%), matching South Island demographic characteristics (Poulton et al., 2015). Fourteen assessments were performed from birth to age 45 years, when 938 of the 997 participants (94.1%) still alive participated. Figure S1 in the Supplemental Material available online presents the results of an attrition analysis of individuals who participated in the study at age 45, which showed that they were representative of the sample as a whole according to childhood SES and IQ. During data-collection assessments, each participant was brought to the research unit for interviews and examinations. Written informed consent was obtained from cohort participants, and study protocols were approved by the appropriate institutional ethical review boards. Research reported here was conducted in accordance with the provisions of the World Medical Association Declaration of Helsinki. The assessments at age 32 years included the collection of heart rate and blood pressure reactivity during an experimental protocol. The study sample included participants who had heart rate and blood pressure reactivity scores from these laboratory tasks ($n = 922$, 49.1% women).

MIDUS study. The original MIDUS sample (Brim et al., 1994) included 7,108 participants. The MIDUS study used a random-digit-dialing procedure to recruit noninstitutionalized, English-speaking people between the ages of 25 and 74 years in the United States in 1995 and 1996. Participants provided informed consent and completed a variety of measures via telephone. This sample was assessed 10 years later ($n = 3,487$) for the MIDUS 2 assessments (Ryff et al., 2017), which repeated much of the original MIDUS assessment. This second wave of data collection also included a subsample of participants ($n = 1,255$) who completed a lab visit in which biomarkers and measures of cardiovascular reactivity (Ryff et al., 2004) were collected. Project staff obtained indicators of heart rate and blood pressure reactivity during the experimental protocol. The present study included MIDUS participants who had heart rate or blood pressure scores from the laboratory tasks ($n = 1,015$). The sample was 53.0 years old ($SD = 11.1$), on average, and 56.2% were women.

Measures

Cardiovascular reactivity. Cardiovascular reactivity was measured in the Dunedin study and MIDUS using cardiovascular responses to two stressful tasks—a computerized Stroop task and a mental arithmetic task. This protocol was adapted from exportable technology developed as part of an NHLBI Cooperative Agreement (HL41340) that aimed to create standardized tasks to evoke cardiovascular reactivity in population-based epidemiological investigations (Kamarck et al., 1992). The principles from this work formed the basis of these two tasks, which have been widely used.

In the Dunedin study, data were collected in four tasks in the following order: 10-min “vanilla” baseline (for a discussion of simple resting and “vanilla” baselines, see Jennings et al., 1992), 4-min Stroop task stressor, 10-min “vanilla” baseline, and 4-min mental arithmetic stressor. During the baseline conditions, participants watched a colored rectangle. This changed color every few seconds, and participants were asked to count the number of occurrences of a certain color, determined randomly. During the Stroop task condition, participants were presented with a word, the name of one of the colors red, green, blue, or yellow, in the middle of the computer monitor. They were asked to indicate what color type the word appeared in by selecting one of four options with a keypad operated by their dominant hand. The four options named colors and appeared in colored type on the computer screen. The combinations of colored type and color names were randomized. Furthermore, a recorded voice recited the names of the colors in a random order. Participants’ responses were timed, and software tracked the response time window, adjusting it to become shorter when the participants made correct responses and longer when participants made incorrect responses. Feedback was provided about the correctness and timeliness of each response.

In the mental arithmetic task, participants were presented with either an addition or subtraction problem on the computer monitor. The problem disappeared, and the word “equals” appeared, then a number, the answer, was presented with the words “yes” and “no” beneath it. Participants were required to select either yes or no using the keypad mentioned above depending on whether they thought the answer presented was correct or incorrect. Difficulty was varied by using three-digit, two-digit, or single-digit numbers in the problems. In this condition, the number of digits in the problems was increased in response to more correct responses and decreased in response to more incorrect responses. There was a time limit on responding, and feedback was given on response correctness and timeliness. During

all tasks, participants wore a Dinamap blood pressure cuff on their nondominant arm and a heart rate monitor (Polar Electro, Kempele, Finland). Blood pressure and pulse rate were taken automatically seven times during the baseline condition and four times during the stressor conditions at 17 s, 1 min 47 s, 3 min 17 s, and 4 min 47 s in all conditions, plus 6 min 17 s, 7 min 47 s, 9 min 17 s in the baseline conditions. Heart rate was monitored continuously from the beginning to end of each condition.

The protocol described above was largely followed in the same way for data collection during the MIDUS 2 biomarker sample psychophysiological assessment (Ryff et al., 2004). In brief, participants completed the same tasks—two 10-min baseline conditions and the two 6-min stressor conditions—with two differences: (a) The baseline in MIDUS was a simple resting baseline (for a discussion of simple resting and “vanilla” baselines, see Jennings et al., 1992) that did not include viewing colored rectangles, and (b) the order of the two stressors in MIDUS was randomized. In addition, blood pressure was measured continuously using a Finometer blood pressure cuff placed on the middle finger of the nondominant hand, and spectral analysis of the blood pressure waveform was used to process the data. More specific details of the protocol are provided in the MIDUS study documentation (Ryff et al., 2017).

In both studies, baseline heart rate and blood pressure were calculated using the mean of the two baseline periods. For both heart rate and blood pressure, cardiovascular reactivity was calculated using difference scores between the appropriate baseline level and subsequent stressful task. If one baseline was missing, the available task baseline was used for both scores. The reactivity scores for each task were then averaged to create a single mean reactivity score for heart rate, systolic blood pressure, and diastolic blood pressure. For full reactivity scores, see Table S1 in the Supplemental Material. Mean arterial pressure was calculated using the standard formula: $[(2 \times \text{diastolic blood pressure}) + \text{systolic blood pressure}] / 3$. Mean arterial pressure was used as the measure of blood pressure reactivity to simplify the reporting of results in the main text; however, all substantive results reported in the study were replicated when systolic or diastolic blood pressure was used as the predictor or outcome (see Table S2 in the Supplemental Material). Heart rate reactivity ($M = 3.5$, $SD = 5.4$) and blood pressure reactivity ($M = 6.0$, $SD = 4.2$) correlated in the Dunedin study, $r = .50$, $p < .001$. Heart rate reactivity ($M = 3.6$, $SD = 3.7$) and blood pressure reactivity ($M = 8.2$, $SD = 6.4$) also correlated in the MIDUS study, $r = .31$, $p < .001$.

Childhood predictors. We examined two measures of early childhood experiences that were hypothesized to

be associated with later cardiovascular reactivity—ACEs and childhood SES. The measures were assessed prospectively in Dunedin and retrospectively in MIDUS.

Adverse childhood experiences. In Dunedin, adverse childhood experiences were generated from archival Dunedin study records gathered during study assessments from ages birth to 15 years. As described previously (Reuben et al., 2016), archival study data were reviewed by four independent raters to determine whether study members experienced 10 types of events that best matched current guidelines, including five types of child harm (physical abuse, emotional abuse, physical neglect, emotional neglect, and sexual abuse) and five types of household dysfunction (incarceration of a family member, household substance abuse, household mental illness, loss of a parent, and household partner violence). Counts greater than four were recoded to four, in line with the U.S. Centers for Disease Control (2016) procedure. The distribution of ACEs in Dunedin ($M = 1.0$, $SD = 1.2$) matched the distribution observed in the Centers for Disease Control study (Reuben et al., 2016).

In MIDUS, ACEs were measured using the Childhood Trauma Questionnaire (Bernstein et al., 1998), which assessed retrospective reports of abuse and neglect that participants experienced in childhood. The five subscales in the measure (emotional abuse, physical abuse, sexual abuse, emotional neglect, physical neglect) were summed, and higher scores represented higher levels of adversity ($M = 7.8$, $SD = 3.0$).

Childhood SES. In Dunedin, the childhood SES of participants' families was measured using the 6-point Elley-Irving Socioeconomic Index for New Zealand (Elley & Irving, 1976). Childhood SES represented the highest SES level of either parent averaged across the interviews of the Dunedin study from the study member's birth through age 15 years ($M = 3.8$, $SD = 1.1$).

In MIDUS, childhood SES was assessed using study members' retrospective reports of their parents' occupations. The maximum value of the occupational Socioeconomic Index score between the male and female head of household at MIDUS 1 assessment was used as the measure of childhood SES, and higher scores represented higher SES ($M = 39.8$, $SD = 14.1$).

Health outcomes. Five health outcomes across Dunedin and MIDUS were studied, three of which were measured in both studies (self-reported health, inflammation, and hypertension) and two of which were measured in only one study (biological aging in Dunedin and mortality in MIDUS).

Self-reported health. In Dunedin and MIDUS, self-reported health was assessed using an item that asked participants, "In

general, would you say your health is?" Responses were given using a 5-point scale (5 = *excellent*, 4 = *very good*, 3 = *good*, 2 = *fair*, and 1 = *poor*). Scores were coded so that higher scores represented better health. In Dunedin, self-reported health was assessed at age 45 years, 13 years after the cardiovascular reactivity assessment ($M = 3.7$, $SD = 0.9$), whereas in MIDUS, self-reported health was assessed at the MIDUS 2 assessment (Ryff et al., 2017), approximately concurrent with the cardiovascular reactivity assessment ($M = 3.7$, $SD = 0.9$).

Inflammation. In Dunedin and MIDUS, systemic inflammation was assessed using blood samples. In Dunedin, samples were collected at age 45 years. The collection was performed between 4:15 p.m. and 4:45 p.m. for all participants. Serum high-sensitivity C-reactive protein (CRP; mg/L) was measured on a Modular P analyzer (Roche Diagnostics GmbH, Mannheim, Germany) using a particle-enhanced immunoturbidimetric assay. In MIDUS, participants who completed the biomarker lab visit (Ryff et al., 2017) provided fasting blood samples between 6:30 a.m. and 7:00 a.m., which were collected and processed using procedures described by Ryff and colleagues (2004). CRP was measured using a nephelometer (BNII; Dade Behring, now Siemens Medical Solutions USA, Inc., Malvern, PA) using a particle-enhanced immunonephelometric assay. Samples falling below the assay range for CRP by this method were reassayed by immunoelectrochemiluminescence using a high-sensitivity assay kit (Ryff et al., 2004). CRP in Dunedin and MIDUS was not normally distributed and was log-transformed for further analyses, as is commonly done (Emerging Risk Factors Collaboration, 2012). Values above 10 mg/L were recoded as missing, following standard practice (Pearson et al., 2003), because such values often represent acute infection rather than systemic inflammation. Mean values were 0.9 ($SD = 0.8$) in Dunedin and 0.9 ($SD = 0.6$) in MIDUS.

Hypertensive status. In Dunedin and MIDUS, hypertensive status was assessed using assessments of resting systolic and diastolic blood pressure. In Dunedin, resting blood pressure was assessed at age 45 years and measured using the average of three readings taken at 5-min intervals. In MIDUS, resting blood pressure was assessed as part of the biomarker assessment (Ryff et al., 2004). In both studies, participants were coded as having hypertension if their systolic blood pressure was 130 mmHg or higher or if their diastolic blood pressure was 80 mmHg or higher, matching current definitions of hypertension from the American Heart Association (Whelton et al., 2018). The percentage of participants in the hypertensive range was 50.5% in Dunedin and 35.7% in MIDUS.

Biological aging. Biological aging was assessed using a previously validated approach called the *pace of aging*

(Belsky et al., 2015). The pace of aging was measured for each participant using repeated assessments of a 19-biomarker panel taken at ages 26, 32, 38, and 45 years. Biomarkers included body mass index, waist-to-hip ratio, glycosylated hemoglobin level, leptin level, blood pressure, cardiorespiratory fitness, forced expiratory volume in 1 s, ratio of forced expiratory volume in 1 s to forced vital capacity, total cholesterol level, triglyceride level, high-density lipoprotein cholesterol level, ratio of apolipoprotein B100 to apolipoprotein A1, lipoprotein(a) level, creatinine clearance, blood urea nitrogen level, C-reactive protein level, white blood cell count, gum health, and caries-affected tooth surfaces. Change over time in each biomarker was modeled with a mixed-effects growth model, and these 19 rates of change were combined into a single index scaled (within sex) in years of physiological change occurring per 1 chronological year ($SD = 0.3$). Participants' pace of aging ranged from approximately 0.4 years to 2.4 years of biological aging per chronological year.

Mortality. MIDUS 3 data collection included information on all known MIDUS and Milwaukee decedents as of November 2017. This date was combined with the date of physiological assessment to create a time variable for survival analyses. Mortality tracing was conducted by the University of Wisconsin Survey Center. Methods included National Death Index searches and ongoing longitudinal sample maintenance (Ryff et al., 2016). Among the MIDUS sample included in this study, 61 deaths were recorded up to the censor date (6.0%).

Individual-difference variables. Three individual-difference variables were included as covariates to help contextualize the main study findings.

Conscientiousness. In Dunedin, study members' personalities were assessed using reports by coinformants at ages 26, 32, and 38 years. At each age, Dunedin study members nominated three people "who knew them well." Informants were mailed questionnaires asking them to describe the study member using a brief, 25-item version of the Big 5 Inventory (Benet-Martínez & John, 1998). Complete Big 5 data were obtained for 946 (96%) participating study members at age 26 years, 935 (96%) study members at age 32 years, and 933 (97%) study members at age 38 years. Conscientiousness was assessed to index the degree to which study members were responsible, attentive, careful, persistent, orderly, planful, and future-oriented (Benet-Martínez & John, 1998). Scores from each informant were averaged across informants and across study occasions to provide an average measure of conscientiousness. A higher score reflects greater conscientiousness ($M = 7.2$, $SD = 1.5$).

In MIDUS 2, participants responded to 31 self-descriptive adjectives designed to assess the Big 5 personality

traits (Ryff et al., 2017). The adjectives used to create the measure of conscientiousness included *organized*, *responsible*, *hardworking*, *careless* (reverse-coded), and *thorough*. Self-reports were completed on a 4-point scale ranging from 1 (*a lot*) to 4 (*not at all*). Higher scores corresponded to relatively higher levels of conscientiousness ($M = 3.5$, $SD = 0.4$).

Cognitive ability. In Dunedin, cognitive ability in adulthood was assessed using the Wechsler Adult Intelligence Scale-IV (WAIS-IV) at age 38 years (Wechsler, 2008). The WAIS-IV generates an IQ score (standardized $M = 100$, $SD = 15$).

In MIDUS, cognitive ability was assessed using the Brief Test of Adult Cognition by Telephone (Lachman & Tun, 2008; Tun & Lachman, 2008), which was collected as part of the MIDUS 2 cognitive project (Ryff & Lachman, 2006). Tasks included word recall (immediate and delayed), digit span backward, category fluency, number series, and backward counting. The variable was calculated as the mean of the z scores of the five subtests ($M = 0.2$, $SD = 0.9$).

Depressive symptoms. In Dunedin, symptoms of depression were assessed as part of the Diagnostic Interview Schedule (Robins et al., 1989). The scale represented a sum of the criteria for major depressive disorder according to the fourth edition of the *Diagnostic and Statistical Manual of Mental Disorders* (American Psychiatric Association, 1994). Scores ranged from 0 to 9, and higher scores represent relatively greater depression ($M = 1.5$, $SD = 2.9$).

In MIDUS, the Center for Epidemiological Studies Depression Scale (Radloff, 1977) was used to assess depressive symptoms. The scale consists of 20 items that are summed together. Higher scores represent relatively more depressive symptoms ($M = 7.8$, $SD = 12.0$).

Data analysis

We used a series of multiple regression models to test whether childhood adversity was associated with later cardiovascular reactivity as well as whether cardiovascular reactivity was associated with later health outcomes in each study. The associations were first tested among the Dunedin study members, then replicated in the MIDUS study using the best available matching variables. All models used multiple regression with the exception of models predicting hypertension and mortality in MIDUS, which instead used logistic regression and a survival model predicting time to death controlling for the date of cardiovascular reactivity assessment, respectively. All analyses in both Dunedin and MIDUS were adjusted for sex and baseline cardiovascular activity. MIDUS analyses were also adjusted for age (the

Table 1. Association of Childhood Adversity and Health Outcomes With Cardiovascular Reactivity in the Dunedin and MIDUS Studies

Predictor	Dunedin (<i>n</i> = 922)		MIDUS (<i>n</i> = 1,015)	
	HR reactivity	BP reactivity	HR reactivity	BP reactivity
Childhood adversity				
Adverse childhood experiences	-0.16* [-0.22, -0.09]	-0.12* [-0.18, -0.06]	-0.08* [-0.18, -0.02]	-0.11* [-0.17, -0.05]
Childhood SES	0.12* [0.06, 0.18]	0.10* [0.04, 0.16]	0.03 [-0.04, 0.10]	0.06 [-0.01, 0.13]
Health outcomes				
Self-reported health	0.10* [0.04, 0.17]	0.19* [0.12, 0.25]	0.10* [0.04, 0.17]	0.13* [0.06, 0.20]
Inflammation	-0.09* [-0.16, -0.01]	-0.13* [-0.20, -0.06]	-0.06* [-0.13, -0.00]	-0.09* [-0.15, -0.02]
Hypertensive status ^a	-0.00 [-0.03, 0.02]	-0.02 [-0.06, 0.01]	0.07* [0.04, 0.11]	-0.02 [-0.05, 0.02]
Pace of aging	-0.17* [-0.23, -0.11]	-0.22* [-0.28, -0.15]	—	—
Mortality ^a	—	—	-0.08* [-0.16, -0.01]	-0.03 [-0.07, 0.01]

Note: Values are standardized regression coefficients (β) unless otherwise noted. Values in brackets are 95% confidence intervals. Dunedin longitudinal study models are adjusted for sex and baseline cardiovascular activity, and MIDUS study models are adjusted for sex, age, and baseline cardiovascular activity. MIDUS childhood predictors were retrospective reports. MIDUS = Midlife in the United States; HR = heart rate, BP = blood pressure; SES = socioeconomic status.

^aValues are log survival odds.

* $p < .05$.

Dunedin participants were all the same age, i.e., born in a 1-year period).

After testing these models, we conducted five additional sets of analyses. First, we examined whether childhood adversity was indirectly associated with health outcomes via cardiovascular reactivity using mediation models. Second, we examined whether there was a nonlinear association between cardiovascular reactivity and the health outcomes, which we tested by including a quadratic term in our models. Third, we tested whether winsorizing cardiovascular reactivity scores greater than 2 *SD* from the mean within each sample attenuated the associations. Fourth, we examined sex differences in the primary results using moderation in a multiple regression framework. Fifth, we controlled for three individual-difference variables—conscientiousness, cognitive ability, and depression—to test whether including these measures as covariates attenuated the primary results. To account for missing data in our models, we used full information maximum likelihood (Graham, 2009) in Mplus (Version 8.3, Muthén & Muthén, 2012).

Results

Childhood predictors of cardiovascular reactivity

Adverse childhood experiences. Dunedin study members with more ACEs had lower heart rate reactivity, $\beta = -0.16$, 95% confidence interval [CI] = [-0.22, -0.09], $p < .001$, and blood pressure reactivity, $\beta = -0.12$, 95% CI = [-0.18, -0.06], $p < .001$. These results were replicated in the

MIDUS study; participants who retrospectively reported more abuse and neglect in childhood had lower heart rate reactivity, $\beta = -0.08$, 95% CI = [-0.18, -0.02], $p = .012$, and lower blood pressure reactivity, $\beta = -0.11$, 95% CI = [-0.17, -0.05], $p = .001$.

Childhood SES. Dunedin study members with lower SES in childhood had lower heart rate reactivity, $\beta = 0.12$, 95% CI = [0.06, 0.18], $p < .001$, and lower blood pressure reactivity, $\beta = 0.10$, 95% CI = [0.04, 0.16], $p = .002$. MIDUS participants' retrospectively reported childhood SES was not significantly associated with heart rate reactivity, $\beta = 0.03$, 95% CI = [-0.04, 0.10], $p = .430$, or blood pressure reactivity, $\beta = 0.06$, 95% CI = [-0.01, 0.13], $p = .088$, although both associations were in the same direction as Dunedin participants. The full results of the associations between childhood predictors and cardiovascular reactivity are presented in Table 1.

Health outcomes and cardiovascular reactivity

Self-rated physical health. Among Dunedin study participants, lower cardiovascular reactivity at age 32 years was associated with worse self-reported health 13 years later—heart rate reactivity, $\beta = 0.10$, 95% CI = [0.04, 0.17], $p = .002$; blood pressure reactivity, $\beta = 0.19$, 95% CI = [0.12, 0.25], $p < .001$. These associations were replicated in the MIDUS study: Lower cardiovascular reactivity during the MIDUS 2 biomarker assessment was associated with worse self-reported health assessed at MIDUS 2—heart rate reactivity, $\beta = 0.10$, 95% CI = [0.04, 0.17], $p = .002$; blood pressure reactivity, $\beta = 0.13$, 95% CI = [0.06, 0.20], $p < .001$.

Inflammation. Among Dunedin study participants, lower cardiovascular reactivity at age 32 years was associated with higher inflammation levels 13 years later—heart rate reactivity, $\beta = -0.09$, 95% CI = $[-0.16, -0.01]$, $p = .018$; blood pressure reactivity, $\beta = -0.13$, 95% CI = $[-0.20, -0.06]$, $p < .001$. These associations were replicated in the MIDUS study: Lower cardiovascular reactivity was associated with higher inflammation levels at the MIDUS 2 biomarker assessment—heart rate reactivity, $\beta = -0.06$, 95% CI = $[-0.13, -0.00]$, $p = .044$; blood pressure reactivity, $\beta = -0.09$, 95% CI = $[-0.15, -0.02]$, $p = .008$.

Hypertension. Among Dunedin study participants, cardiovascular reactivity at age 32 years was not significantly associated with hypertensive status 13 years later—heart rate reactivity, hazard ratio (*HR*) = 1.00, 95% CI = $[0.97, 1.03]$, $p = .845$; blood pressure reactivity, *HR* = 0.98, 95% CI = $[0.94, 1.01]$, $p = .192$. In the MIDUS study, blood pressure reactivity was not significantly associated with hypertensive status at the MIDUS 2 biomarker assessment, *HR* = 0.98, 95% CI = $[0.95, 1.02]$, $p = .537$, although higher heart rate reactivity was associated with a higher likelihood of hypertension, *HR* = 1.08, 95% CI = $[1.04, 1.12]$, $p < .001$. Given that hypertension depends on a cut-off for resting blood pressure, we also tested these associations between blood pressure reactivity and hypertension without controlling for resting blood pressure. Neither study evidenced a significant association—Dunedin, *HR* = 1.00, 95% CI = $[0.97, 1.03]$, $p = .971$; MIDUS, $\beta = 0.99$, 95% CI = $[0.97, 1.01]$, $p = .245$.

Biological aging. Dunedin study participants with lower cardiovascular reactivity had a faster pace of biological aging—heart rate reactivity, $\beta = -0.17$, 95% CI = $[-0.23, -0.11]$, $p < .001$; blood pressure reactivity, $\beta = -0.22$, 95% CI = $[-0.28, -0.15]$, $p < .001$.

Mortality. MIDUS study participants with lower heart rate reactivity were at greater risk of early mortality, *HR* = 1.09, 95% CI = $[1.01, 1.17]$, $p = .026$, although there was not a significant association between blood pressure reactivity and early mortality, *HR* = 1.03, 95% CI = $[0.99, 1.07]$, $p = .139$. Note that the sample included only 61 deaths (6.0%), which likely limited the power to detect associations in MIDUS. The full results of the association between cardiovascular reactivity and health outcomes are presented in Table 1.

Secondary analyses: indirect effects of childhood adversity on health via cardiovascular reactivity

Having established direct effects between childhood adversity, cardiovascular reactivity, and health in both

samples, we next tested whether there were significant indirect effects of childhood adversity on health via cardiovascular reactivity using a series of mediation models. We ran all possible models independently for the two measures of childhood adversity, two measures of cardiovascular reactivity, and four health outcomes. All mediation models were fully saturated, run using bootstrapping ($n = 1,000$), and controlled for baseline cardiovascular activity.

A limited number of significant indirect effects linked childhood adversity (childhood ACEs and SES) to health outcomes consistently across the two samples. ACEs were significantly associated with later self-reported health and inflammation via blood pressure reactivity in both MIDUS and Dunedin (Table 2). No other significant indirect effects between childhood adversity and health via cardiovascular reactivity were replicated between the two samples. Of the measures available in only one sample, ACEs and childhood SES were indirectly associated with biological aging as assessed in the Dunedin study via both measures of cardiovascular reactivity. The full results of the mediation models are presented in Table 2. Broadly, we observed nonsignificant indirect effects or significant indirect effects that were small in size. Direct effects between childhood adversity and health were generally larger in size, and these effects remained significant when accounting for cardiovascular reactivity.

Secondary analyses: probing the robustness of the study findings

Having established that more adverse childhoods were associated with less cardiovascular reactivity and that less cardiovascular reactivity was associated with poorer midlife health outcomes, we conducted additional analyses to verify and better understand the study findings.

Testing for nonlinearity. We first tested whether there were nonlinear associations between cardiovascular reactivity and the health outcomes used in the study. Prior research has suggested that it might be detrimental to health to be on either extreme in terms of cardiovascular reactivity (Phillips, 2016), and one method to test this possibility is to test whether there were nonlinear associations in our data. To do so, we created quadratic transformations of our cardiovascular reactivity variables by standardizing and squaring them (e.g., Standardized Heart Rate Reactivity \times Standardized Heart Rate Reactivity). We then ran our primary models of interest including both the linear and quadratic forms of the cardiovascular reactivity scores as predictors. Neither cardiovascular reactivity measure evidenced a consistent nonlinear association with any of the health outcomes in the primary

Table 2. Effects of Childhood Adversity on Health Via Cardiovascular Reactivity

Model and effect	Dunedin (<i>n</i> = 922)		MIDUS (<i>n</i> = 1,015)	
	HR reactivity	BP reactivity	HR reactivity	BP reactivity
ACEs → self-reported health				
Direct effect	−0.12* [−0.19, −0.06]	−0.12* [−0.19, −0.05]	−0.25* [−0.32, −0.17]	−0.25* [−0.32, −0.17]
Indirect effect	−0.01* [−0.02, −0.00]	−0.02* [−0.03, −0.01]	−0.01 [−0.02, −0.00]	−0.01* [−0.02, −0.00]
ACEs → inflammation				
Direct effect	0.09* [0.02, 0.15]	0.10* [0.03, 0.17]	0.08* [0.01, 0.14]	0.08* [0.02, 0.15]
Indirect effect	0.01 [−0.00, 0.02]	0.02* [0.00, 0.03]	0.01 [−0.00, 0.01]	0.01* [0.00, 0.02]
ACEs → hypertensive status [†]				
Direct effect	−0.01 [−0.13, 0.12]	0.06 [−0.07, 0.18]	−0.04 [−0.09, 0.01]	−0.02 [−0.13, 0.04]
Indirect effect	0.00 [−0.02, 0.02]	0.01 [−0.01, 0.03]	−0.01* [−0.01, −0.00]	0.00 [−0.01, 0.01]
ACEs → pace of aging				
Direct effect	0.18* [0.11, 0.25]	0.18* [0.11, 0.25]	—	—
Indirect effect	0.02* [0.01, 0.03]	0.02* [0.01, 0.04]	—	—
ACEs → mortality ^a				
Direct effect	—	—	0.12* [0.04, 0.20]	0.12* [0.04, 0.20]
Indirect effect	—	—	0.01 [−0.00, 0.02]	0.01 [−0.01, 0.02]
Childhood SES → self-reported health				
Direct effect	0.15* [0.09, 0.22]	0.15* [0.10, 0.24]	0.12* [0.06, 0.19]	0.12* [0.06, 0.19]
Indirect effect	0.01* [0.00, 0.02]	0.02* [0.01, 0.03]	0.00 [−0.00, 0.01]	0.01 [−0.00, 0.02]
Childhood SES → inflammation:				
Direct effect	−0.13* [−0.19, −0.06]	−0.13* [−0.19, −0.06]	−0.08* [−0.15, −0.01]	−0.08* [−0.15, −0.02]
Indirect effect	−0.01 [−0.02, 0.00]	−0.01* [−0.02, −0.00]	−0.00 [−0.01, 0.00]	−0.01 [−0.00, 0.00]
Childhood SES → hypertensive status ^a				
Direct effect	−0.14* [−0.26, −0.01]	−0.17* [−0.31, −0.04]	0.00 [−0.01, 0.01]	0.00 [−0.02, 0.02]
Indirect effect	0.00 [−0.02, 0.02]	−0.01 [−0.02, 0.01]	0.00 [−0.00, 0.00]	−0.00 [−0.00, 0.00]
Childhood SES → pace of aging				
Direct effect	−0.21* [−0.29, −0.17]	−0.22* [−0.27, −0.16]	—	—
Indirect effect	−0.02* [−0.03, −0.01]	−0.02* [−0.03, −0.01]	—	—
Childhood SES → mortality ^a				
Direct effect	—	—	−0.00 [−0.03, 0.03]	0.00 [−0.03, 0.03]
Indirect effect	—	—	−0.00 [−0.01, 0.01]	−0.00 [−0.01, 0.01]

Note: Values are standardized regression coefficients (β) unless otherwise noted. Values in brackets are 95% confidence intervals. MIDUS models are also adjusted for age. MIDUS = Midlife in the United States; ACEs = adverse childhood experiences; HR = heart rate; BP = blood pressure; SES = socioeconomic status.

^aValues are log survival odds.

* $p < .05$.

models. One exception was the association of heart rate reactivity with inflammation in MIDUS, $\beta = -0.08$, 95% CI = [−0.15, −0.01], $p = .043$. This effect was not replicated within the sample using the second measure of reactivity, nor was it replicated for either measure of reactivity in the other study cohort. All other associations were non-significant and relatively small in terms of effect size. For full results, see Table S3 in the Supplemental Material. Figure 1 illustrates mean levels among the continuously measured health outcomes for participants grouped into high, low, and average cardiovascular reactivity.

Cardiovascular reactivity outliers. We next tested whether these associations were due to outliers influencing the associations. We identified participants within each

study who had heart rate and blood pressure reactivity scores greater or lesser than 2 *SD* around the mean and winsorized these scores to ± 2 *SD* from the mean. In Dunedin, this included 43 (4.7%) heart rate scores and 39 (4.2%) blood pressure scores. In MIDUS, this included 43 (4.2%) heart rate scores and 53 (5.2%) blood pressure scores. When winsorizing these outlier scores, all effect sizes changed a negligible amount when compared with the primary analyses, $|\Delta\beta| \leq 0.03$ (for full results, see Table S4 in the Supplemental Material), providing evidence that the main findings were not due to outliers in cardiovascular reactivity.

Testing for sex differences. We also tested whether the associations might evidence sex differences. To do so, we

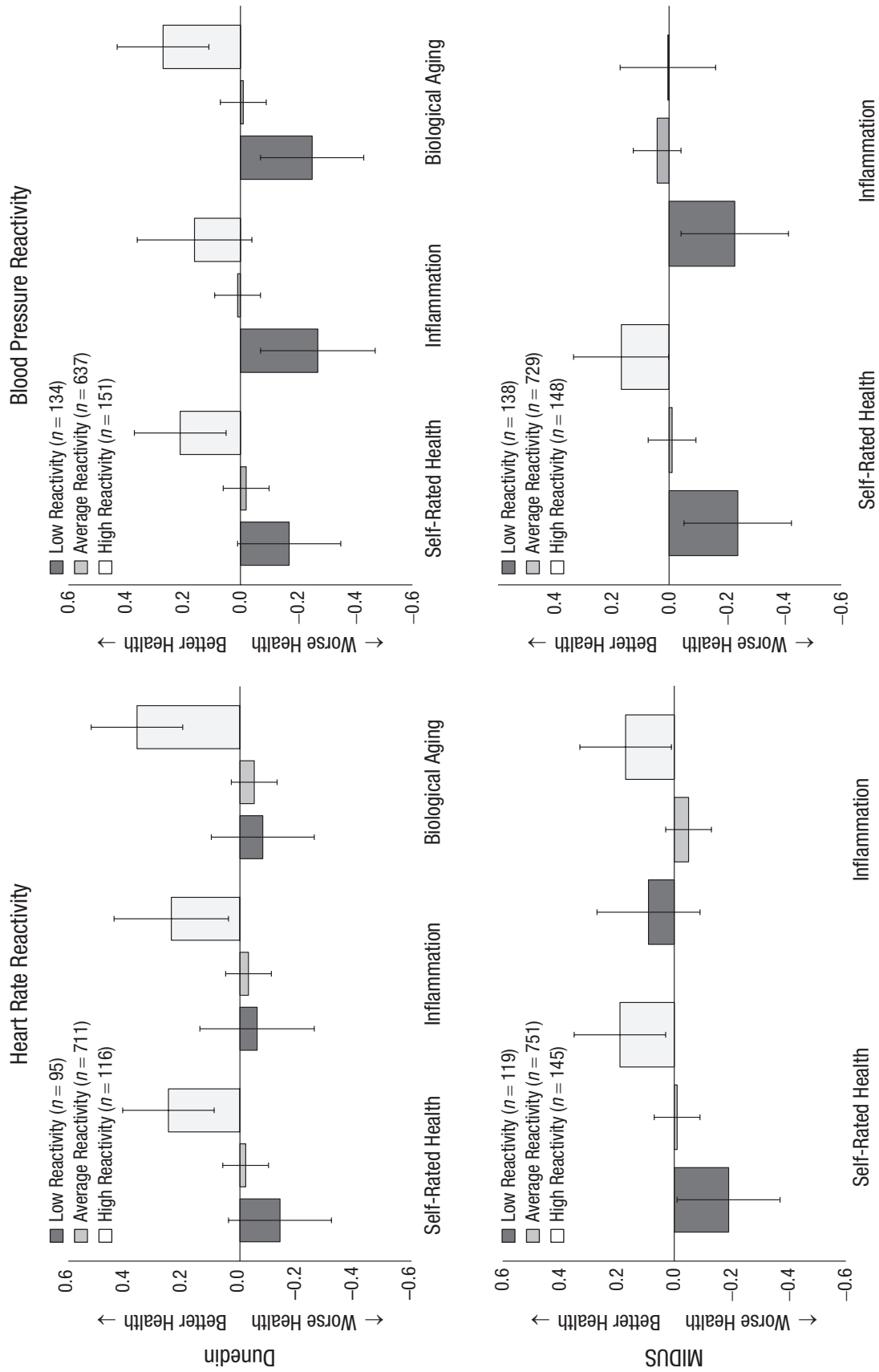


Fig. 1. Health outcomes illustrating the mean levels by levels of heart rate reactivity and blood pressure reactivity. Participants with reactivity more than 1 SD below the mean are categorized as having low reactivity. Participants with reactivity between -1 and 1 SD from the mean are categorized as having average reactivity. Participants with reactivity more than 1 SD above the mean are categorized as having high reactivity. Outcomes were standardized and scaled so that higher scores equaled better health. Ns for C-reactive protein (CRP) do not include participants missing data because their values were > 10 mg/L.

ran the models for our primary results while adding the appropriate interaction term to the predictors. The results did not indicate any consistent patterns of sex differences that were replicated across the cohorts. There was one significant interaction term, $\beta = 0.08$, 95% CI = [0.01, 0.15], $p = .027$, in which lower blood pressure reactivity was associated with greater risk for early mortality among women in MIDUS but not men. This result was not replicated for heart rate reactivity. For full results, see Table S5 in the Supplemental Material.

Evidence from individual-difference variables. Given the findings supported a blunting model of cardiovascular reactivity (Carroll et al., 2009, 2017; Ginty et al., 2013), we tested whether three individual differences—conscientiousness, cognitive ability, and depressive symptoms—might help explain these results. First, we tested whether the lab stressors elicited greater reactivity among highly conscientious research participants. People who are more conscientious are also generally healthier (Roberts et al., 2005), and conscientious participants may have been more likely to fully engage with the computerized cognitive stressor tasks. Indeed, conscientious participants in both Dunedin and MIDUS were generally more reactive to the stressor tasks (Dunedin: heart rate reactivity, $\beta = 0.13$, 95% CI = [0.07, 0.19], $p < .001$, blood pressure reactivity, $\beta = 0.18$, 95% CI = [0.11, 0.24], $p < .001$; MIDUS: blood pressure reactivity, $\beta = 0.11$, 95% CI = [0.04, 0.18], $p = .002$, but not heart rate reactivity, $\beta = 0.04$, 95% CI = [-0.02, 0.11], $p = .189$).

Second, we tested whether the lab stressors elicited greater reactivity among high-functioning participants. People with greater cognitive ability are generally healthier (Batty et al., 2009), and these participants may have more fully engaged with the computerized cognitive stressor tasks as well. High-functioning participants in both Dunedin and MIDUS were more reactive to the cognitive stressor tasks (Dunedin: heart rate reactivity, $\beta = 0.20$, 95% CI = [0.13, 0.26], $p < .001$, blood pressure reactivity, $\beta = 0.26$, 95% CI = [0.20, 0.32], $p < .001$; MIDUS: heart rate reactivity, $\beta = 0.10$, 95% CI = [0.04, 0.17], $p = .002$, blood pressure reactivity, $\beta = 0.17$, 95% CI = [0.11, 0.23], $p < .001$).

Finally, we tested whether the lab stressors elicited less reactivity among participants with higher levels of depressive symptoms. Depression is cited as a possible explanation for why blunted reactivity is associated with poorer health outcomes (Carroll et al., 2017; Ginty et al., 2013; Phillips et al., 2013) and might attenuate the observed associations. Similar to the other individual differences, participants with more depressive symptoms were less reactive to the cognitive stressor tasks in both studies (Dunedin: heart rate reactivity, $\beta = -0.08$, 95% CI = [-0.14, -0.02], $p = .015$, blood pressure

reactivity, $\beta = -0.17$, 95% CI = [-0.24, -0.11], $p < .001$; MIDUS: heart rate reactivity, $\beta = -0.14$, 95% CI = [-0.20, -0.08], $p < .001$, blood pressure reactivity, $\beta = -0.21$, 95% CI = [-0.27, -0.15], $p < .001$).

We repeated our primary analyses controlling for conscientiousness, cognitive ability, and depression to test whether these factors may have helped explain the associations between childhood adversity, cardiovascular reactivity, and health. Controlling for these individual differences attenuated the associations observed in the Dunedin and MIDUS studies, although effects were less attenuated in MIDUS. In Dunedin, controlling for these measures attenuated the primary associations, on average, by 0.06 in terms of absolute standardized effect size for heart rate (52.3% of original effect sizes) and by 0.08 for blood pressure (67.8% of original effect sizes). In MIDUS, controlling for these measures attenuated the primary associations, on average, by 0.03 in absolute effect size for heart rate (48.2% of original effect sizes) and 0.03 for blood pressure (47.4% of original effect sizes). In total, two thirds of the significant associations observed in the primary results were no longer significant (12 of 18 associations) when accounting for these variables. These results suggest that some participants' conscientiousness, cognitive ability, or depression may have influenced their engagement during the cardiovascular reactivity tasks. Figure 2 illustrates the attenuation of effects (for full results, see Table S6 in the Supplemental Material).

Discussion

In the current study, we used two longitudinal cohorts, the Dunedin ($n = 922$) and the MIDUS ($n = 1,015$) studies, to examine the links between early life adversity, cardiovascular reactivity, and health outcomes in adulthood. The primary results showed that people with more adversity in childhood consistently had lower levels of later cardiovascular reactivity and that lower cardiovascular reactivity was consistently associated with poorer health—with the exception of hypertensive status. These findings were largely replicated across the two cohorts. Although not empirically tested, descriptively, when the two cohorts differed on how constructs were measured, associations were stronger in the cohort that used more proximal or reliable measurement (e.g., prospectively measured childhood SES and ACEs in Dunedin compared with retrospective reports in MIDUS; full-scale IQ in Dunedin compared with a shorter cognitive battery in MIDUS).

The secondary results revealed two main findings. First, when testing indirect effects of childhood adversity on health via cardiovascular reactivity, significant effects were observed across the two cohorts linking

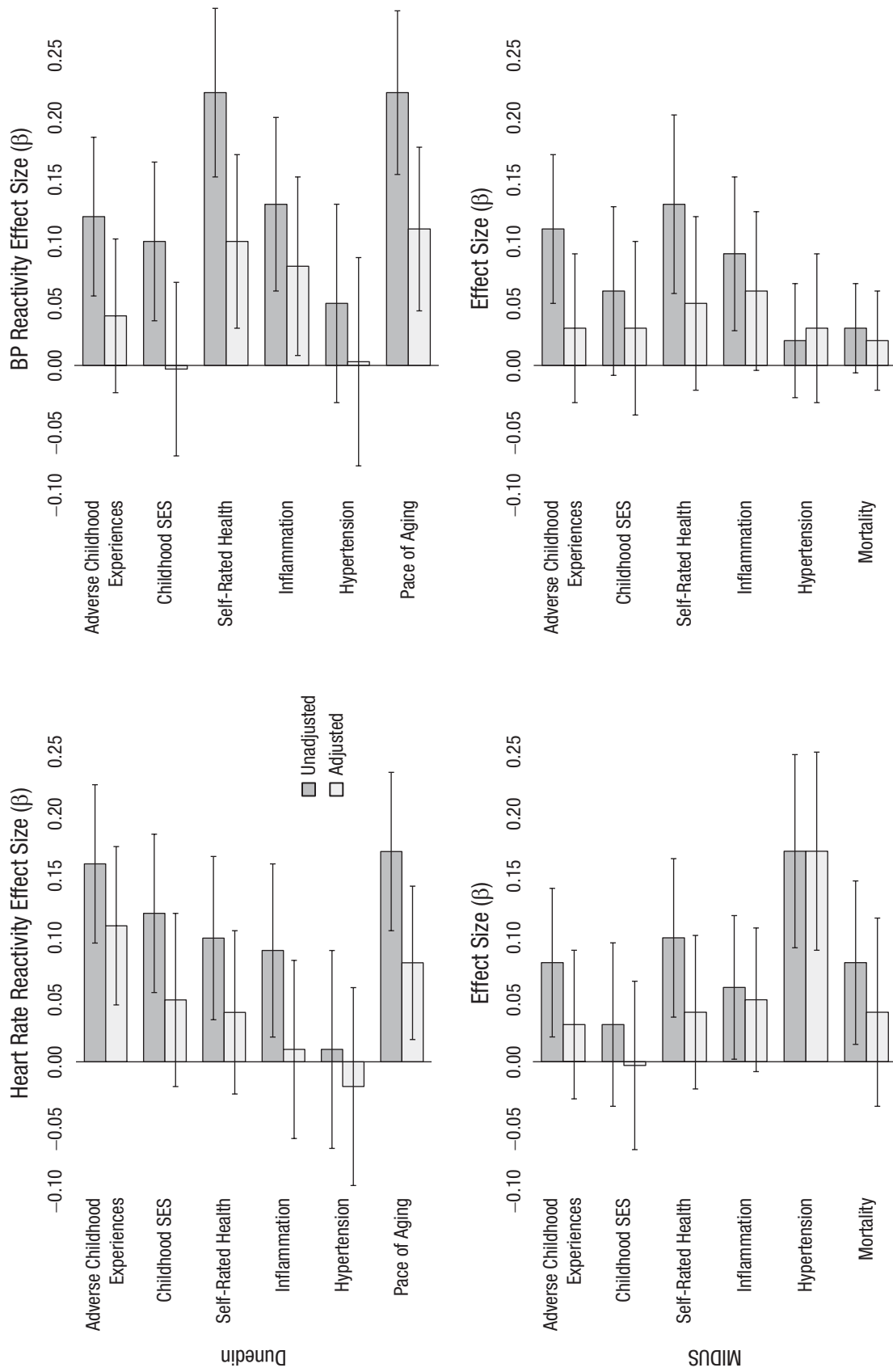


Fig. 2. Associations of childhood predictors and health outcomes with unadjusted cardiovascular reactivity and with cardiovascular reactivity adjusted for conscientiousness, cognitive ability, and depression.

ACEs to poorer self-rated health and greater inflammation via lower blood pressure reactivity. Greater adversity in childhood was also indirectly associated with faster biological aging in Dunedin via lower cardiovascular reactivity. Second, our models showed that cardiovascular reactivity was associated with individual differences in conscientiousness, cognitive ability, and depressive symptoms and that including these constructs as covariates in our models partially attenuated the association of cardiovascular reactivity with childhood adversity and health. The results suggest these individual differences account for the observed results, at least in part.

How should these results be understood? We propose two interpretations. First, these results may reflect the specific tasks and methods used to assess cardiovascular reactivity in the Dunedin and MIDUS studies. Both studies used cognitive stressors that were administered through computerized prompts, and both assessed participants who had completed a number of previous study assessments. Social evaluation plays an important role in eliciting cardiovascular reactivity (Bosch et al., 2009; Smith et al., 1997), and the computerized administration of these tasks may have presented participants with less social evaluation than is present when an experimenter asks participants to complete math questions. Second, participants in both cohorts had long-standing familiarity with research studies. They may have habituated to the experience of participating in study tasks, which may have resulted in lower levels of social evaluation during the Stroop task and math stressors than might be the case among people participating in a research study for the first time.

Reactivity to these computerized, cognitive stressor tasks may instead have varied systematically according to participants' individual differences. For example, people with higher levels of intelligence and/or who are more conscientious may have placed more value on performing well on these tasks and evidenced higher reactivity as a result. Likewise, more depressed participants may engage less fully in such tasks. The associations between reactivity and the individual-difference variables provide some evidence for this conclusion. This interpretation would suggest that the types of stressors used in future studies of cardiovascular reactivity should be carefully chosen. Although cognitive stressors such as the Stroop task or mental arithmetic task are assumed to be stressful for all people, they could evoke differences in cardiovascular activity because of individual differences in how people appraise or respond to such stressors—which may be correlated with the relevant predictors, outcomes, or both in any given study—resulting in spurious findings.

Note that the current study assessed these covariates at a similar point in time as cardiovascular reactivity and that alternative explanations are also possible. For example, rather than reducing engagement to result in blunting, these individual differences might also be consequences of blunted cardiovascular reactivity, or both could be affected by an unmeasured third variable. Future studies would benefit from also using stressful tasks that do not appear to reflect individual differences strongly, such as stressors like the cold pressor (Brindle et al., 2017). Alternatively, tasks that are high in emotional salience, such as a trauma recall or anger recall, might result in more consistent task engagement. Additional empirical study using multiple methods of eliciting reactivity (e.g., cognitive, physiological, and emotionally salient stressful tasks) would be needed to provide evidence to this end.

A second interpretation is that the life-course conceptualization of cardiovascular reactivity linking early adversity and later health through greater reactivity needs to be revised. Our results found associations in the opposite direction of what was expected on the basis of extending the cardiovascular-reactivity hypothesis. Greater cardiovascular reactivity being associated with better health stands in opposition to research across a variety of contexts, tasks, and populations (see Chida & Steptoe, 2010; Jennings et al., 2004; Manuck, 1994; Treiber et al., 2003). However, much of the research linking cardiovascular reactivity with poorer health is focused on hypertension and disease states. For example, although in a 2010 meta-analysis, Chida and Steptoe found an overall association between cardiovascular reactivity and cardiovascular risk status ($r = .09$), this effect was restricted to prospective associations with higher blood pressure—there was a null association when examining the pooled correlation between reactivity and subclinical markers of atherosclerosis. Note that we did not observe a consistent association between cardiovascular reactivity and hypertension in the Dunedin and MIDUS cohorts. Some studies actually have found that greater reactivity was associated with improved markers of disease (Heponiemi et al., 2007). More recent studies have found similar associations; for example, people with higher diastolic blood pressure reactivity had lower risk for subsequent early mortality (Kupper et al., 2015).

The emerging research from the past decade and a half examining blunted cardiovascular reactivity and health (Phillips, 2016) provides important context to these results. Over the past decade, low levels of cardiovascular reactivity (i.e., a blunted response) to cognitive stressors have been linked to poorer health outcomes (Allen, 2013; Carroll et al., 2009; Carroll, Phillips, et al.,

2012), leading to the theory that particularly low levels of reactivity—in addition to higher reactivity—may be detrimental to health because of associated dysregulation in behavioral motivation (Carroll et al., 2017; Ginty et al., 2013; Phillips et al., 2013). Within this model, one could expect that both high and low reactivity would be associated with poorer health. However, we did not find evidence of nonlinearity in our results—participants with the highest cardiovascular reactivity were the healthiest and vice versa.

Studies of posttraumatic stress disorder and developmental psychopathology have also found that trauma and childhood adversity are associated with lower cardiovascular reactivity (Ginty et al., 2017; Lin et al., 2014; Murali & Chen, 2005; Voellmin et al., 2015), which may be relevant to understanding the mixed evidence linking cardiovascular reactivity and health broadly. Such work theorizes that adversity and trauma, particularly severe or chronic exposure to such stressors, can result in dysregulation of the stress response system and a “freezing” response to stress (Lin et al., 2014). A freezing response would result in decreased cardiovascular reactivity when faced with threatening tasks (Heleniak et al., 2016). The current results could support a blunted cardiovascular-reactivity hypothesis and suggest that people who experience early life adversity develop lower later cardiovascular reactivity as a result of changes to the stress-response system. These lower levels of cardiovascular reactivity, in turn, could be associated with poorer health by indexing individual differences in behavioral motivation (Carroll et al., 2017). The specific mechanisms linking motivated behavior to health were not explicitly tested in this study, however, which limits the degree to which these findings support such theoretical speculation.

This study has notable strengths. The samples are large compared with many studies of cardiovascular reactivity and include a variety of measures that are not available in some of the larger, population-based cohort studies. In addition, we were able to replicate the majority of our results across two independent cohorts using multiple measures of cardiovascular reactivity to the same cognitive stressors, providing evidence that the results are not simply due to Type I error. These strengths speak to the reliability of the results but speak less to which interpretation of the results is more appropriate. Determining the best interpretation would require, at minimum, examining associations between childhood adversity, cardiovascular reactivity, and health outcomes using reactivity to other types of stressful tasks, particularly because people with blunted responses to cognitive stressors do not evidence this same blunting to cold pressor and exercise tasks (Brindle et al., 2017). Regardless of the correct interpretation,

publication of these results is particularly important given concerns about file-drawer effects in the study of cardiovascular reactivity. As stated by Allen (2013), “Indeed, one wonders how many ‘file cabinet’ data sets are out there that were not published or even submitted for publication due to not getting the expected exaggerated reactivity to be associated with a variable of interest” (p. 98). Of the 313 publications using MIDUS biomarker data from 2005 to 2020, to our knowledge, only three studies (Coyle et al., 2020; Creaven et al., 2020; Lin et al., 2014) used cardiovascular reactivity. Likewise, the Dunedin longitudinal study (Poulton et al., 2015) collected cardiovascular reactivity at age 32 years, but—as described in the introduction—has not published any empirical articles on these data to date.

This study also has limitations. First, although both cohort studies included longitudinal assessments, neither included multiple assessments of cardiovascular reactivity. Assessments of cardiovascular reactivity in childhood would be necessary to test if reactivity had been blunted from a previously normal level among people who experienced adversity in childhood. Future studies would benefit from multiple measurements of cardiovascular reactivity, which might provide temporal ordering when examining its associations with childhood adversity, individual-difference variables (e.g., conscientiousness, depressive symptoms, cognitive ability), and health. Second, both studies relied on cognitive stressors to assess cardiovascular reactivity, and it is unknown whether different stressors (e.g., a cold pressor task) would yield different results. Third, there was significant overlap in the constructs measured in the two cohorts, but differences in the timing and type of measures of childhood adversity (e.g., retrospective compared with prospective) or health outcomes (e.g., mortality compared with biological aging) limited our ability to replicate findings across cohorts.

Conclusion

This study used two longitudinal cohorts to investigate the association between early life adversity, cardiovascular reactivity, and health outcomes. The results were in the opposite direction to what would be predicted according to the traditional cardiovascular-reactivity hypothesis. Early life adversity was associated with lower levels of later cardiovascular reactivity, and lower cardiovascular reactivity was associated with worse health outcomes in adulthood. These results were somewhat attenuated when controlling for individual differences in personality, intelligence, and depression that may have affected participants’ reactions to the stressors used in the study. These results highlight the

importance of accounting for individual differences when assessing cardiovascular reactivity using computerized cognitive stressors. In addition, the results may support links between early life adversity, a lower cardiovascular response to stress, and poorer health.

Transparency

Action Editor: Kenneth J. Sher

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Author Contributions

K. J. Bourassa, T. E. Moffitt, and A. Caspi developed the study concept, with feedback from all authors. T. E. Moffitt, H. Harrington, R. Poulton, S. Ramrakha, and A. Caspi planned and managed the study design and data collection for the Dunedin study. K. J. Bourassa conducted the data analysis and interpretation, with substantive feedback from T. E. Moffitt, A. Caspi, and R. Houts. K. J. Bourassa drafted the manuscript with substantive feedback from T. E. Moffitt and A. Caspi. R. Houts conducted the statistical reproducibility check of the results. H. Harrington, R. Houts, R. Poulton, and S. Ramrakha provided additional feedback on the final manuscript, and all of the authors approved the final manuscript for submission.

Declaration of Conflicting Interests


The author(s) declared that there were no conflicts of interest with respect to the authorship or the publication of this article.

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Supplemental Material

Additional supporting information can be found at <http://journals.sagepub.com/doi/suppl/10.1177/2167702621993900>

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