

Gender Differences in Biological Function in Young Adulthood: An Intragenerational Perspective



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Sex-gender differences in health are a function of social and biological factors and their interplay over the life course. A large body of research documents sex-gender as a determinant of health behavior and outcomes. Far less scholarship examines how these differences are reflected in physiologic function in young adulthood. Using nationally representative, longitudinal data from the National Longitudinal Study of Adolescent to Adult Health, we examine the relationship between gender and biological function in young adulthood. We also examine the contribution of social and economic circumstances in childhood and early adulthood to gender differences in health. The findings reveal strong gender differences in information and immune function, which are robust to the inclusion of many indicators of the social environment.

Keywords: gender, biomarkers, life course

Sex and gender differences in health and longevity are well established (Read and Gorman 2010; Rieker, Bird, and Lang 2010). Female advantages in survival and life expectancy exist across time and place (Austad 2006). The factors that contribute to these patterns are multifaceted, but sex and gender differences in patterns of acute and chronic illnesses are pervasive (Crimmins et al. 2010). In the United States, for example, women report higher rates of chronic illness, and men report higher rates of acute illness. Further, men are more likely to die from cardiovascular disease and cancer at younger ages and to be more vulnerable to

illness and death once gender-differentiated social and behavioral factors are taken into account (Case and Paxson 2005). Nonetheless, although these patterns provide a logic, the underlying mechanisms remain elusive.

Sex and gender differences in health and longevity are understood to be a function of social and biological factors, and their interplay over the life course (Bird and Rieker 2002; Institute of Medicine 2001; Yang and Kozloski 2011; Short, Yang, and Jenkins 2013). Notably, scholarship on sex and gender-differentiated aging tends to focus on middle and late adulthood, when ill health and death are more fre-

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quent, rather than on the early adulthood period, a time that is formative in shaping long-term health trajectories (Harris 2010). Moreover, although a large body of research documents the importance of sex and gender as a determinant of health behavior and health outcomes, far less scholarship examines how these differences are reflected in physiologic function, especially in young adulthood. Yet, physiologic function is an important mediator through which social experiences, such as health behaviors, may “get under the skin” with consequence for later life health.

A life course perspective leads us to expect that physiologic function in young adulthood is shaped by circumstances during childhood (Hayward and Gorman 2004; Heckman 2006; Jackson 2010; McDade et al. 2014). The adoption of behaviors begins early in life, when morbidity risk is low but youth decide whether to initiate behaviors that “track” into adulthood, such as smoking, exercise, and health-seeking behavior (for example, Chen and Kandel 1995). Research on socioeconomic status and health has long adopted a life course perspective, but research on gender and health has less systematically incorporated early life experiences, even though many health behaviors and social circumstances are shaped by gender. One result is that we know little about whether we should attribute differences in health between men and women to experiences during adulthood, to the cumulative and persistent effect of earlier behaviors and circumstances, or to other factors.

Using nationally representative, longitudinal data from the National Longitudinal Study of Adolescent to Adult Health (Add Health), we have two goals. First, we examine the relationship between gender and biological function in young adulthood, considering biomarkers of inflammation and immunosuppression, two markers associated with later life morbidities. Second, we examine the contribution of social and economic circumstances in childhood and early adulthood to the size of any gender differences in health, as indicated by levels of Epstein-Barr virus (EBV) and C-reactive protein (CRP). Examining health within a cohort of young men and women in early adulthood will reveal the emergence of gender differences in

both biological and clinical markers of health, as well as differences both within and across women and men according to profiles of childhood and adulthood environments. Understanding the intragenerational process by which gender-specific variation in health is produced will inform our broader understanding of the emergence of gender differences in health over the life course.

BACKGROUND

Sex and gender differences in health vary across contexts and change over time. On average, for example, U.S. women live 4.8 years longer than men, down from a peak gap of 7.8 years in the late 1970s (Arias 2014). Explanations for the gender gap as well as changes the gap over time have focused on health behaviors among adults, revealed variation in the size of gender differences across educational groups, and documented differences in the social and economic pressures both men and women face, underscoring the role of social experiences in shaping sex and gender patterns of health (Arias 2014; Berkman 2012; Meara, Richards, and Cutler 2008; Montez and Zajacova 2013; Pampel 2001).

At the same time, sex-specific variation in biology has promoted biologically anchored explanations for differences in health and survival (Bird and Rieker 2002; Short, Yang, and Jenkins 2013). These include differences tied to immune competence, which is related to reproductive biology and possibly shaped by hormones; differences in insulin-like growth factor 1, signaling, and oxidative stress production; and differences in genomes that stem from the presence of X and Y chromosomes, and processes such as cell mosaicism (Fish 2008).

Previous research indicates that disease exposure and susceptibility, immune response, and markers of inflammation vary with sex and gender (Lleo et al. 2008; Markle and Fish 2014; Ordaz and Luna 2012). The prevalence, onset, and severity of autoimmune diseases and allergic diseases are one example. Generally, women are two to three times as likely to develop an autoimmune disease as men, but disease severity can be worse in men (Shames 2002). Research points to a role for “sex” steroids and hypothalamic-pituitary hormones in

immune system development and immune response, but the associated mechanisms and models are not well specified (Fish 2008; Shames 2002). Evidence that puberty, pregnancy, and hormonal contraceptive use are associated with variation in immune system response is well documented (Shames 2002). For example, asthma severity is affected by menstruation, pregnancy, and menopause (Shames 2002). Likewise, inflammation varies by sex and gender, with patterns indicating differences in prevalence, severity, and onset that are not well understood (Fish 2008). Overall, data from the National Health and Nutrition Survey (NHANES) indicates that inflammation is higher among women than men, the difference diminishing with age, and especially so in late adulthood (Yang and Kozloski 2011). Notably, numerous exposures shape levels, and sometimes in gender-differentiated ways. Existing research, based on clinic and population samples, claims adiposity, smoking, and sleep is related to inflammation levels differently in women and men (Cartier et al. 2009; Le-Ha et al. 2014; Miller et al. 2009).

A Life Course Perspective

Despite a robust body of research documenting the association between gender and health among adults, limited scholarship to date examines gender differences in physiologic function at a relatively young age. Evidence is clear that young women are more likely than young men to demonstrate high levels of CRP (Ishii et al. 2012; Shanahan, Freeman, and Bauldry 2014), and that gender-specific HPA-axis reactivity is observed in adolescence (Oldehinkel and Bouma 2011). We know less, however, about the extent of gender differences in young adults' biologic function across the full distribution of risk, or about the intragenerational process by which these patterns are produced. The behaviors and health outcomes we observe among adult men and women are generated from behaviors and exposures occurring earlier in life, given ample evidence that health behaviors, as well as biomarkers of health, track from adolescence into adulthood (for example, Chen and Kandel 1995; McDade, Williams, and Snodgrass 2007).

Studying biomarkers of health is particu-

larly useful among young adults because these measures offer insight into future disease risk among a population in which clinically defined disease is low, but predictive power for future disease prevalence is high. Life course theory emphasizes the possibility that circumstances across ages may have differing and combined effects on childhood and adulthood outcomes, pointing to the importance of the timing, duration and stability of a circumstance (Ben-Schlomo and Kuh 2002; Ferraro and Shippee 2009; Schoon et al. 2002). This perspective enables consideration of whether we should attribute differences between men and women to experiences during adulthood, to the cumulative and persistent effect of earlier behaviors and circumstances, or to something else.

Circumstances During Adulthood

Ample evidence documents a strong relationship between individuals' social and economic environments and health, whether health is self-reported or defined by particular acute, chronic or disabling conditions. A growing literature also documents the ways in which health affects social and economic processes over the life course (Kitagawa and Hauser 1973; Jackson 2015; Lynch 2003; Marmot 2001; Moore and Hayward 1990; Morenoff 2003; Palloni 2006). The gender gap in life expectancy is particularly pronounced among women with less education; relatedly, some dimensions of socioeconomic status (SES), such as education and employment, are increasing across historical time in their predictive power for health, suggesting that circumstances during adulthood remain strongly linked to health (Kunst et al. 2005; Lynch 2003; Meara, Richards and Cutler 2008; Montez and Zajacova 2013; Smith et al. 2000). Studies that bridge two literatures—those on the socioeconomic determinants of health and on links among biological markers and morbidity-mortality—have also yielded important findings about the role of the social environment in predicting physiologic functioning among adults (see, for example, McDade et al. 2014). Poverty is associated with elevated levels of C-reactive protein, an inflammatory marker related to cardiovascular disease (Alley et al. 2006; Kanjilal et al. 2006), as well as elevated blood pressure and chole-

terol (Karamangla et al. 2005). In addition, data from national health surveys, such as NHANES, or disease-specific surveys, such as CARDIA (Coronary Artery Risk Development in Young Adults), show how biological measures of health are unequally distributed by race-ethnicity and socioeconomic indices.

Circumstances During Childhood and Adolescence

Considering adulthood circumstances exclusively, however, only partly accounts for the environments that may result in gender differences in health. Circumstances during childhood have far-reaching effects and are key to understanding outcomes during adulthood (Hayward and Gorman 2004; Heckman 2006; Jackson 2010). Evidence is compelling that the socioeconomic “gradient” in health has origins in childhood—education is strongly related to health even before birth, and that relationship grows with age (Case, Lubotsky, and Paxson 2002; Finch 2003). Similarly, behavior and mortality during adulthood are strongly predicted by childhood socioeconomic status (Duncan, Ziol-Guest, and Kalil 2010; Hayward and Gorman 2004). Evidence is accumulating that childhood socioeconomic status leads to differences in physiologic functioning, such as EBV and CRP, before adulthood (Dowd et al. 2014; McDade, Stallings, and Worthman 2000). Whether youth decide to adopt particular behaviors also depends in part on their family socioeconomic circumstances (Duncan, Ziol-Guest, and Kalil 2010). Youth in low-resource families are more likely to be exposed to unhealthy behaviors and living circumstances, less likely to pursue higher education, and more likely to experience financial instability and unemployment (Duncan, Ziol-Guest, and Kalil 2010; Wagmiller et al. 2006).

Socioeconomic Pathways Linking Gender to Biological Function

Given evidence that socioeconomic circumstances during childhood, adolescence and adulthood are strongly related to health and biological function, biological risk should be greater among both men and women who experience low-resource environments during childhood and the transition to adulthood. To

the extent that socioeconomic circumstances, or the meaning of such differences, at different points in the early life course are gender differentiated, gender differences in biological function may be explained in part by differential exposures and experiences. If women are especially likely to experience persistently low or declining SES environments through early adulthood, greater or cumulative exposure to stressful and unhealthy circumstances might contribute to gender differences in biological function.

Evidence of gender differences in exposures and socioeconomic circumstances that are plausibly linked to health is stronger among adults than among children and adolescents. For example, in the United States, adult women on average earn lower wages than men and have lower incomes (Jacobson 2016). Occupational segregation leads women and men to experience different workplace environments (Read and Gorman 2010). Social integration and behaviors such as smoking and exercising vary by gender (Pampel 2001; Umberson, Crosnoe, and Reczek 2010; Yang et al. 2013). Men and women experience differences in family responsibilities and different forms of gender-specific harassment, both of which also shape health (Bianchi, Robinson, and Milke 2006; Rieker, Bird, and Lang 2010). Many of these patterns implicate social structures that shape health through constraining choices in gender-differentiated ways (Rieker, Bird, and Lang 2010).

Although evidence to date is scant, youth circumstances can also differ by gender, and such differences may well vary with family resources (Williams 2002). A growing literature, for example, demonstrates a more favorable environment for girls with respect to the development and sanctioning of social and behavioral skills. Girls are less likely than boys to exhibit aggressive behavior, as indicated by externalizing behaviors, and some evidence suggests that they are less likely than boys to be penalized in school for the same problematic behaviors (DiPrete and Jennings 2012; Entwisle et al. 2007; Farkas 1990). At the same time, increasing social pressure to conform to gender expectations may reduce girls' perceived or real opportunities as they enter adolescence, and may contribute to higher levels of depression

among girls (Nolen-Hoeksema 2001; Pomeranz et al. 2002). In sum, experiences vary in gendered ways over the life course and may reasonably be related to health and aging (Short, Yang, and Jenkins 2013). Our goal in this article is to investigate whether gender differences in physiological function are evident among young adults, and to describe the social and demographic correlates over the life course associated with these patterns.

DATA

Our analyses are based on data from waves one through four of the National Longitudinal Study of Adolescent to Adult Health, a longitudinal study of adolescents' health and its determinants. The first wave of this nationally representative, school-based sample of about twenty thousand adolescents was conducted in 1994 and 1995, when students were in grades seven through twelve, ranging in age from eleven to twenty-one; mean age at the first wave is sixteen. Information was gathered from schools, adolescents, and parents. Data collection has resulted in three subsequent waves to date: one to two years after baseline (1996), seven years after baseline (2001), and fourteen years after baseline (2008). In line with the survey's goal of understanding the transition into adulthood, information is collected from respondents about their health, relationships, educational experiences, and labor market participation. At wave one, information about family background was also collected from parents. Add Health data are useful for this research in that they offer detailed information on socioeconomic background, behaviors, health, and social and economic transitions during the period of the life course when behaviors are initiated and key transitions are made.

MEASURES

Biomarkers are key to this analysis—measures of inflammatory and immune functioning, available at wave four. We examine high-sensitivity C-reactive protein (hsCRP) and EBV antibodies. CRP is a commonly used indicator of inflammation that is highly correlated with cardiovascular disease, type 2 diabetes, and mortality (Fahdi et al. 2003; Ridker et al. 1998).

EBV is a commonly used measure of immune function. Although EBV is extremely common, with approximately 90 percent of the human population estimated to be infected (Dowd, Palermo, Brite, et al. 2013), the virus is usually latent unless reactivated. Higher levels of EBV antibodies indicate greater difficulty in the immune system's ability to regulate the virus, and have been linked to diseases such as cancer, lupus, and multiple sclerosis (Esen et al. 2012; Hsu and Glaser 2000; James and Robertson 2012; Levin et al. 2010; Thompson and Kurzrock 2004). EBV antibody levels indicate reduced cell-mediated immune function and are associated with social stressors (McDade, Stallings, and Worthman 2000).

In the Add Health, both CRP and EBV were measured via dried blood spot samples, and results are reported as milligrams per liter for CRP, and as arbitrary units per milliliter for EBV. Both measures have been used in previous research with Add Health data (for example, Everett et al. 2014; McDade et al. 2014). We represent CRP and EBV using logged measures in final analyses. We considered representing CRP and EBV using sex-standardized measures, but did not find sufficient rationale in the existing literature for adopting this approach. In particular, CRP and EBV are not uniformly patterned by sex across populations. Further, a given level of CRP or EBV does not translate consistently into a clear level of risk across populations. Explanations for sex and gender differences in both patterns and risks note the complicated interaction of social, behavioral, and biological factors in producing patterns. Taken together, our overall approach is guided by a desire to impose as few assumptions as possible on the measurement of CRP and EBV. Thus, in our analyses, rather than computing risk thresholds that may impose arbitrary cutoffs between individuals, or constructing sex-specific measures, we examine variation across the full distribution of each measure, without sex-specific standardization, using quantile regression techniques.

We measure several sociodemographic variables, some of which are of primary interest and all of which are potentially correlated with both gender and biologic function. *Gender* is a dichotomous variable based on respondents'

identification as either male or female (reference category). *Race-ethnicity* distinguishes among those who identify as non-Hispanic and white (reference category), non-Hispanic and black, Hispanic, Asian, and other (including Native Americans). Respondents who identify within multiple racial-ethnic categories are represented by the one that they report to best reflect their identity. *Nativity status* distinguishes among first-generation (immigrant, reference), second-generation (one or both immigrant parents), and third-generation (both parents U.S.-born) youth.

Parental education is measured categorically at wave 1 through responding parents' reports: less than high school (reference), high school completion, some college, and college diploma or higher. In the vast majority of cases (94 percent) the responding parent is the mother. We take the natural log of *family income* at wave 1; income coefficients can therefore be interpreted as a percentage change. *Parental marital status* is also measured at wave 1 and separates those whose parents are currently married from all other union statuses (reference). We use these three measures as our focal indicators of socioeconomic circumstances in childhood and adolescence.

We also control for measures of *parental obesity* and respondent *birthweight*. Parental obesity captures both genetic and environmental sources of weight and height. We combine mothers' and fathers' reports at wave one into a measure indicating whether either parent has ever been told by a doctor that he or she is obese. Birthweight is a continuous measure indicating respondents' weight (in pounds) at birth. Finally, we include a continuous measure of *age*.

At waves three and four, we measure several important indicators of respondents' social and economic environments. To capture socioeconomic status in early adulthood, we measure respondents' *educational attainment* by wave four, differentiating some college and college-plus relative to those with a high school education or less. We measure *household income* at wave four with a continuous measure that uses the midpoint of each income band. A binary measure indicates whether respondents have *ever been married* at wave four. We

measure current *employment status*, as well as a measure indicating how often respondents have had to reduce work hours in the past year because of *family responsibilities*. Finally, we measure the *number of live births* resulting from pregnancies up to and including wave four.

We also control for several *health behaviors*, including an indicator of daily smoking (yes-no) and physical activity (yes-no) at wave four. We control for obesity (yes-no) at wave four—results are not sensitive to using a three-category measure of normal weight, overweight or obesity, or to measuring BMI-overweight at wave three. Finally, we control for *pregnancy status* at wave 4, and test the sensitivity of the results to excluding women who are currently pregnant; we include all women because the results do not change.

In all analyses, we examine the sensitivity of our results to the inclusion of measures of morbidity to understand the extent to which our biomarkers are affected by preexisting and contemporaneous health. These include several wave three and four measures of health: self-rated health (on a five-point scale ranging from excellent to poor) at wave three, and self-reports of a physician diagnosis of high blood pressure (waves three and four), high cholesterol (wave three), and asthma (wave three). We control for recent acute illness and infection, for whether respondents have an inflammatory disease, and for respondents' anti-inflammatory and immune-suppression medication use at wave four. Finally, we control for women's use of hormonal contraception at wave four.

ANALYSIS

We use quantile regression to consider variation in the effects of gender across the observed distributions of CRP/EBV. Typical regression approaches identify group differences in the mean of the dependent variable. Although this approach is often a sensible one, it assumes that the determinants of the center of a distribution are the same as those at the extremes. This assumption may lead researchers to obscure important variation among respondents, or to unnecessarily exclude cases in an effort to avoid undue influence from extreme observations. In the case of CRP levels, for example,

respondents with very high levels of CRP (typically considered to be above 10 mg/L) are often excluded from analyses because of the possibility that these levels reflect acute (rather than chronic or systemic) inflammation (O'Connor et al. 2009). At the same time, however, some evidence indicates both that high CRP levels not only are a proxy for recent illness, but also more strongly predict future cardiovascular and mortality risk than low levels, and that women are more likely to have very high CRP levels (for example, Shanahan, Freeman, and Bauldry 2014). These findings suggest that understanding variation in biological function across the distribution is equally important to understanding average variation. Quantile regression estimates conditional differences in the median and other quantiles (10th, 25th, 50th, 75th, and 90th percentile) using least-absolute-value estimation. Differences between men and women are estimated at each percentile, given their other characteristics. For models with covariates other than gender, the gender coefficient can be interpreted as differences in CRP and EBV levels between men and women, for those with otherwise identical characteristics on observed covariates. We consider differences among women across the distribution, and between men and women at and across percentiles.

We begin with a baseline model that only includes gender, and successively incorporate childhood or adolescent and then early adulthood characteristics. In addition to identifying gender differences in inflammation and immune function, this strategy allows us to examine the contribution of social and economic environments at different points of the early life course to gender differences observed in adulthood. We compare quantile regression estimates to those from ordinary least squares (OLS) regression models that incorporate the same childhood or adolescent and early adulthood factors.

RESULTS

We begin by describing gender differences in biological function in bivariate and multivariate perspective. Next, we examine variation across the distribution and describe several sensitivity analyses.

Establishing Differences in Biological Function

Table 1 presents weighted characteristics of the analytic sample by gender. Male and female respondents have highly equivalent sociodemographic characteristics during childhood or adolescence. About two-thirds of each group identify as non-Hispanic white, and about 15 percent of respondents are from immigrant families, either immigrants themselves or the child of an immigrant parent (16 percent of men, 15 percent of women). Almost 25 percent of respondents lived in families with a college-educated parent (23 percent of males, 22 percent of females). By early adulthood, women are more likely to have received a college degree (34 percent of women versus 28 percent of men). Women are also more likely to have ever been married by wave 4 (55 percent of women versus 45 percent of men).

Male and female respondents also have generally similar early health environments, with almost 25 percent of respondents having an obese parent at baseline, and with mean birthweight about 7.5 pounds (7.75 for men, 7.47 for women). By wave 3, young women are slightly more likely to have been diagnosed by a doctor with high blood pressure and asthma. By wave 4, men and women are equally likely to be obese (37 percent and 36 percent), and women are slightly less likely to be regular smokers than men (22 percent versus 28 percent). Despite their shared environments with respect to parental circumstances and sociodemographic characteristics, women have higher measured CRP and EBV levels at wave 4. Mean CRP is 3.3 for men and 6.3 for women, and mean EBV is 138.4 for men and 164.2 for women. These descriptive patterns suggest that, by early adulthood (approximately age thirty), women in Add Health demonstrate more inflammation and lower immune function than their male peers.

Table 2 presents the results from an OLS regression of logged CRP and EBV (respectively) on gender and the other covariates. Net of respondents' childhood or adolescent and early adulthood social, economic, and health environments, men and women have significantly different levels of inflammation and immune function in young adulthood. Levels of CRP in

Table 1. Weighted Characteristics of Sample by Gender, Add Health

	Male (49%)	Female (51%)
Race-ethnicity		
Non-Hispanic white (reference)	69	69
Hispanic	12	11
Black	14	16
Asian	4	3
Other	1	1
Nativity		
First generation (reference)	6	5
Second generation	10	10
Third-plus generation	84	85
Parental education, wave 1		
Less than high school (ref.)	16	17
High school	32	33
Some college	29	28
College or more	23	22
Logged family income, wave 1	3.16	3.12
Parents married, wave 1	73	73
Age (years)	30	29.8
Education, wave 4		
High school or less	40	31
Some college	32	35
College or more	28	34
Ever married, wave 4	45	55
Employed part or full-time, wave 4	86	67
Work hours affected by family duties, wave 4		
Frequently	3	9
Sometimes	15	20
Rarely	23	19
Never	60	52
Health		
CRP (mg/L)	3.3	6.2
EBV (AU/L)	138.4	164.2
Parent obese, wave 1	23	24
Birthweight	7.75	7.47
Smoke daily, wave 4	28	22
Exercise daily, wave 4	13	17
Obese, wave 4	37	36
Self-rated health, wave 3	1.9	2.1
Asthma, wave 3	15	18
High blood pressure, wave 3	5	7
High cholesterol, wave 3	4	4
N	6,451	6,715

Source: Authors' compilation based on Add Health, Waves 1-4.

Table 2. OLS Regression of Logged CRP (mg/L) and EBV (AU/L), Add Health

	Log CRP	Log EBV
Male	-0.524*** (-21.58)	-0.190*** (-14.52)
Race-ethnicity		
Hispanic	0.131** (3.25)	0.0786*** (3.62)
Black	-0.00982 (-0.31)	0.200*** (11.61)
Asian	-0.313*** (-5.69)	0.0448 (1.52)
Other race	-0.129 (-1.01)	0.133 (1.93)
Nativity		
Second generation	0.00164 (0.03)	0.0504 (1.74)
Third-plus generation	0.0520 (0.96)	0.0525 (1.80)
Parental education		
High school	-0.0591 (-1.60)	-0.0190 (-0.95)
Some college	-0.115** (-3.03)	-0.0266 (-1.30)
College or more	-0.139*** (-3.34)	-0.0718** (-3.21)
Logged family income	-0.00104 (-0.18)	0.0109*** (3.48)
Parent obese	0.137*** (4.90)	0.0228 (1.51)
Birth weight	-0.0328*** (-4.00)	0.00212 (0.48)
Parents married	-0.0116 (-0.43)	-0.0192 (-1.31)
Age (wave 1)	-0.000455 (-0.07)	0.0118** (3.16)
Obese, wave 4	0.157*** (21.63)	0 (0.03)
Smoke daily, wave 4	-0.0325 (-1.10)	0.0569*** (3.56)
Exercise daily, wave 4	0.182*** (5.66)	-0.0111 (-0.64)
Ever married, wave 4	0.0772** (3.15)	0.0450*** (3.41)
Some college, wave 4	-0.0583* (-2.03)	-0.0321* (-2.07)
College or more, wave 4	-0.133*** (-4.38)	-0.0159 (-0.98)
Household income, wave 4	8.81e-08 (1.73)	2.89e-08 (1.05)
Employed part or full time	-0.0828* (-2.09)	-0.0204 (-0.98)
Work or family conflict	-0.00185 (-0.11)	-0.0045 (-0.49)
Constant	0.908*** (6.24)	4.569*** (58.38)
N	13,166	13,238

Source: Authors' compilation based on Add Health, Waves 1-4.

Note: Models also control for pregnancy status at wave 4, and parity at wave 4.

* $p < .05$; ** $p < .01$; *** $p < .001$

men are more than 50 percent lower than among women, and levels of EBV 19 percent lower. Higher levels of CRP in women is consistent with other examinations at of the Add Health, as well as Multiethnic Study of Atherosclerosis and NHANES (Lakoski et al. 2006; Yang and Kozloski 2011), but this pattern is not universal across populations (Oksuzyan et al. 2015). Similarly, EBV is higher in women than men in analyses of the NHANES (ages six through nineteen), and a national population-based sample in Taiwan (Chen et al. 2015; Dowd, Palermo, Chyu, et al. 2013), and in heterosexual respondents in the Add Health. However, analyses of gay and bisexual respondents in the Add Health find little evidence of difference between men and women (Everett et al. 2014).

Consistent with previous research, the findings also reveal higher CRP levels among Hispanic respondents. Compared with non-Hispanic whites, CRP levels are 13 percent higher among Hispanics; EBV levels are 8 percent higher among Hispanics and 20 percent higher among blacks. Asian respondents have CRP levels that are 31 percent lower than among non-Hispanic whites, on average, but no difference is observed for EBV levels. After controlling for socioeconomic factors, no nativity-based differences in CRP and EBV are significant. Education shows an educational gradient in CRP and EBV, whereby respondents with college-educated parents have 13 percent lower CRP levels and 7 percent lower EBV levels, on average, than their peers from the most poorly educated families. By early adulthood, respondents who attain a college education experience a similarly lower risk of inflammation, with a 13 percent difference between college-educated respondents than their peers with a high school education or less. There is also a higher risk associated with marriage, on average, whereby respondents who have ever been married by wave four have 7 percent higher CRP levels and 5 percent higher EBV levels than their peers.

Finally, respondents' family health environments and early health characteristics are associated with inflammation—respondents with an obese parent have 13 percent higher CRP levels and 3 percent higher EBV levels, on aver-

age; each additional pound at birth is associated with a 3 percent decrease in CRP in adulthood. In general, the covariates are more consistently and strongly associated with inflammation than with immune function, though there is certainly evidence of meaningful gender differences in immune function.

Gender Differences in Biological Function Across the Distribution

The results from OLS analyses establish gender differences in biological function at the center of the distribution. This approach establishes a useful benchmark against which to examine variation across the distribution of CRP and EBV. Given evidence that young adult women are more likely than men to have very high levels of CRP, and that it is common to exclude such cases from analyses despite their strong relationship with future morbidity and mortality, it is important to understand the degree of risk at extreme values of the distribution.

Table 3 presents the results from the quantile regression of CRP on gender and the other covariates. The results show that, for those with identical observed characteristics, the size of gender differences in inflammation increases significantly across the distribution. Model 1, which shows the bivariate relationship between gender and CRP, shows that men at the 10th percentile have CRP levels that are approximately 17 percent lower than women, and that this difference rises to 36 percent at the 25th percentile, 60 percent at the median, 66 percent at the 75th percentile, and 69 percent at the 90th percentile. Coefficient equality tests across percentiles show that these differences are significant at the 0.01 level. Results for EBV, presented in table 4, are somewhat different. There is no evidence of a monotonic gradient—instead, gender differences are most pronounced in the middle of the distribution, with women at the median having EBV levels about 21 percent lower than their similar male peers. The size of the gender difference is 16 percent at the 10th percentile, and 16 percent at the 95th percentile.

Exploring the contribution of childhood or adolescent and early adulthood factors in explaining gender differences in CRP and EBV reveals robust differences. After controlling for

Table 3. Quantile Regression of Logged CRP (mg/L) on Gender, Add Health

	Percentile					
	0.1			0.25		
	(1)	(2)	(3)	(1)	(2)	(3)
Male	-0.190*** (-6.95)	-0.173*** (-4.14)	-0.214*** (-4.27)	-0.358*** (-16.39)	-0.354*** (-11.25)	-0.444*** (-11.28)
Race or ethnicity						
Hispanic		0.201*** (3.53)	0.133* (1.99)		0.269*** (3.87)	0.127* (2.31)
Black		-0.0725 (-1.74)	-0.137** (-2.68)		0.00530 (0.13)	-0.0627 (-1.18)
Asian		-0.329*** (-3.71)	-0.303*** (-4.72)		-0.319*** (-4.44)	-0.338*** (-3.94)
Other race		0.118 (0.94)	0.123 (0.57)		0.379* (2.03)	0.0509 (0.31)
Nativity						
Second generation		-0.0190 (-0.31)	-0.0468 (-0.60)		0.179** (2.89)	0.0335 (0.40)
Third-plus generation		0.0620 (1.20)	0.0199 (0.26)		0.223*** (3.29)	0.0706 (0.93)
Parental education						
High school		-0.104* (-2.09)	-0.0800 (-1.15)		-0.189*** (-5.29)	-0.0904 (-1.59)
Some college		-0.148** (-2.97)	-0.143* (-2.07)		-0.284*** (-4.26)	-0.152** (-3.02)
College or more		-0.263*** (-4.98)	-0.147* (-1.99)		-0.387*** (-7.94)	-0.136 (-1.84)
Logged family income		-0.00207 (-0.34)	-0.00780 (-0.96)		0.000830 (0.15)	-0.00465 (-0.47)
Parents married		-0.0485 (-1.38)	-0.0127 (-0.29)		-0.0246 (-0.74)	-0.0483 (-1.29)
Ever married, wave 4			0.0520 (1.55)			0.0658* (2.02)
Some college, wave 4			0.0478 (0.98)			-0.0243 (-0.54)
College or more, wave 4			-0.162** (-3.17)			-0.204*** (-4.93)
Household income, wave 4			7.88e-08 (0.85)			6.70e-08 (0.67)
Employed part or full time			-0.0421 (-0.69)			-0.172* (-2.19)
Work or family conflict			-0.0229 (-0.60)			0.0212 (-0.8)
Constant	-0.949*** (-45.07)	1.174*** (-5.17)	-0.778*** (-3.53)	-0.0471* (-2.07)	-0.413 (-1.94)	0.0558 (0.25)
N	13,600	13,600	11,359			

Source: Authors' compilation based on Add Health, Waves 1-4.

Note: Models also control for measures listed in table 1, pregnancy status at wave 4, and parity at wave 4. T statistics in parentheses.

* $p < .05$; ** $p < .01$; *** $p < .001$

Percentile								
0.5			0.75			0.9		
(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)
-0.616*** (-27.20)	-0.573*** (-20.27)	-0.606*** (-17.44)	-0.674*** (-21.08)	-0.645*** (-21.10)	-0.663*** (-19.12)	-0.695*** (-24.94)	-0.694*** (-21.75)	-0.695*** (-12.49)
	0.247*** (5.00)	0.131* (2.17)		0.215*** (4.58)	0.162*** (3.52)		0.193*** (3.68)	0.159* (2.07)
	0.110** (2.60)	0.0177 (0.50)		0.154*** (3.75)	0.0430 (0.79)		0.192*** (6.26)	0.0865 (1.81)
	-0.272*** (-4.95)	-0.304*** (-3.86)		-0.182** (-2.76)	-0.266** (-3.02)		-0.204 (-1.82)	-0.334*** (-4.63)
	0.402* (2.40)	0.268 (1.82)		0.505*** (5.81)	0.532*** (4.53)		0.605*** (4.71)	0.369* (2.04)
	0.159* (2.46)	-0.0163 (-0.23)		0.145** (2.65)	-0.0121 (-0.17)		0.0648 (1.01)	-0.0670 (-0.77)
	0.213*** (4.06)	-0.0289 (-0.49)		0.244*** (5.99)	0.101 (1.25)		0.184* (2.02)	0.0166 (0.17)
	-0.148*** (-3.67)	-0.0573 (-1.28)		-0.119** (-2.76)	-0.0583 (-1.48)		-0.0812 (-1.59)	-0.0142 (-0.29)
	-0.271*** (-5.29)	-0.151** (-3.14)		-0.240*** (-4.35)	-0.128** (-2.77)		-0.133 (-1.87)	-0.0215 (-0.31)
	-0.403*** (-11.04)	-0.149* (-2.56)		-0.332*** (-6.40)	-0.144*** (-3.60)		-0.229*** (-3.44)	-0.0558 (-0.75)
	-0.000617 (-0.08)	0.00101 (0.13)		-0.00670 (-0.93)	-0.000577 (-0.06)		0.00247 (0.33)	0.0112 (1.24)
	-0.0345 (-1.19)	-0.0129 (-0.46)		-0.00435 (-0.12)	0.00748 (0.31)		-0.00452 (-0.13)	0.0116 (0.35)
		0.103** (2.87)			0.0706* (2.12)			0.0614* (1.98)
		-0.0583 (-1.83)			-0.0819* (-2.36)			-0.112 (-1.86)
		-0.157*** (-4.01)			-0.0931** (-2.62)			-0.0954 (-1.74)
		0.000000141* (2.16)			0.000000172** (2.89)			1.78e-08 (0.30)
		-0.0859 (-1.36)			-0.117 (-1.84)			-0.0190 (-0.25)
		0.00342 (0.12)			-0.017 (-0.64)			-0.0198 (-0.62)
1.066*** (45.27)	0.762*** (5.32)	0.903*** (4.73)	2.003*** (105.83)	1.650*** (7.21)	2.051*** (10.17)	2.729*** (116.94)	2.428*** (8.47)	2.615*** (10.81)

Table 4. Quantile Regression of Logged EBV (AU/L) on Gender, Add Health

	Percentile					
	0.1			0.25		
	(1)	(2)	(3)	(1)	(2)	(3)
Male	-0.164*** (-5.26)	-0.176*** (-7.54)	-0.180*** (-6.17)	-0.209*** (-7.58)	-0.199*** (-12.01)	-0.197*** (-11.45)
Race-ethnicity						
Hispanic		0.161*** (4.22)	0.197*** (6.67)		0.0929*** (4.73)	0.124*** (4.38)
Black		0.219*** (8.04)	0.248*** (7.38)		0.244*** (9.74)	0.254*** (10.93)
Asian		0.145* (2.51)	0.203*** (4.38)		0.0229 (0.79)	0.0688 (1.84)
Other race		0.0994 (1.46)	0.0997 (1.64)		-0.000739 (-0.01)	-0.0895 (-1.01)
Nativity						
Second generation		0.0152 (0.26)	0.0563 (1.31)		0.0819 (1.89)	0.0792 (1.91)
Third-plus generation		0.0534 (1.09)	0.105* (2.34)		0.0879* (2.47)	0.113* (2.50)
Parental education						
High school		0.0127 (0.40)	-0.00129 (-0.03)		-0.0333 (-1.14)	-0.0156 (-0.52)
Some college		0.0268 (0.86)	0.0269 (0.65)		-0.0351 (-1.25)	-0.0153 (-0.46)
College or more		-0.0867** (-2.85)	-0.0795 (-1.93)		-0.104** (-3.27)	-0.0832* (-2.00)
Logged family income		0.0133* (2.00)	0.0170 (1.77)		0.0148*** (4.44)	0.0149** (2.63)
Parents married		-0.0134 (-0.51)	0.00155 (0.05)		-0.0172 (-0.75)	-0.0165 (-0.69)
Ever married, wave 4			0.127*** (4.20)			0.0781*** (3.31)
Some college, wave 4			0.00445 (0.19)			-0.0113 (-0.74)
College or more, wave 4			-0.0451* (-1.96)			-0.00775 (-0.33)
Household income, wave 4			9.29e-08 (1.83)			8.82e-08* (2.10)
Employed part or full time			-0.0139 (-0.33)			0.0185 (0.64)
Work or family conflict			-0.0128 (-0.78)			0.0129 (0.60)
Constant	3.970*** (169.27)	3.302*** (24.46)	3.217*** (24.21)	4.443*** (237.19)	3.916*** (34.20)	3.935*** (38.71)
N	13,679	13,679	11,428			

Source: Authors' compilation based on Add Health, Waves 1-4.

Note: Models also control for measures listed in table 1, pregnancy status at wave 4, and parity at wave 4.

T statistics in parentheses.

* $p < .05$; ** $p < .01$; *** $p < .001$

Percentile								
0.5			0.75			0.9		
(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)
-0.211*** (-9.06)	-0.207*** (-11.80)	-0.218*** (-11.45)	-0.169*** (-8.36)	-0.172*** (-12.30)	-0.184*** (-15.31)	-0.158*** (-9.32)	-0.135*** (-7.65)	-0.157*** (-12.13)
	0.0361 (1.25)	0.0620 (1.87)		0.0736* (2.17)	0.0576* (1.98)		0.0220 (0.68)	0.00382 (0.11)
	0.213*** (15.19)	0.198*** (9.77)		0.182*** (11.09)	0.156*** (6.92)		0.197*** (9.86)	0.175*** (5.09)
	-0.0264 (-0.59)	0.00500 (0.09)		0.0205 (0.59)	0.0110 (0.24)		0.0536 (1.40)	0.0350 (1.27)
	0.0798 (1.03)	0.0536 (0.73)		-0.0190 (-0.48)	0.00478 (0.07)		-0.0655 (-0.90)	-0.0160 (-0.26)
	0.0531 (1.31)	0.0792 (1.53)		0.0644 (1.80)	0.0607 (1.23)		0.0142 (0.48)	-0.0315 (-0.65)
	0.0445 (1.45)	0.0922 (1.71)		0.0417 (1.46)	0.0306 (0.75)		-0.00744 (-0.25)	-0.0520 (-1.03)
	-0.0370 (-1.48)	-0.0287 (-0.92)		0.0133 (0.66)	0.00392 (0.23)		-0.0263 (-1.26)	-0.0255 (-0.88)
	-0.0379 (-1.82)	-0.0367 (-1.22)		-0.00181 (-0.10)	-0.00905 (-0.42)		-0.0567* (-2.17)	-0.0499* (-1.97)
	-0.0978*** (-4.25)	-0.0899** (-2.79)		-0.0593* (-2.42)	-0.0589* (-2.26)		-0.102** (-2.79)	-0.0869*** (-3.57)
	0.00607 (1.46)	0.00755 (1.50)		0.00133 (0.26)	0.00530 (0.86)		0.00659* (2.55)	0.00852 (1.95)
	0.00113 (0.06)	-0.00721 (-0.31)		-0.00540 (-0.25)	-0.0361* (-2.25)		-0.0266 (-1.26)	-0.0433* (-2.24)
		0.0317 (1.79)			0.0189 (0.95)			0.00252 (0.15)
		-0.0437 (-1.62)			-0.0492* (-2.27)			-0.0388 (-1.70)
		0.0123 (0.63)			-0.00428 (-0.20)			-0.0151 (-0.74)
		2.73e-08 (0.98)			1.74e-08 (0.50)			1.96e-08 (-0.48)
		0.0067 (0.34)			-0.052* (-2.27)			-0.0486 (-1.40)
		-0.0017 (-0.20)			-0.0245* (-2.50)			-0.0036 (-0.27)
4.956*** (520.53)	4.593*** (41.91)	4.526*** (37.75)	5.389*** (482.49)	5.080*** (49.22)	5.167*** (56.21)	5.753*** (443.19)	5.615*** (48.30)	5.736*** (49.86)

demographic factors and for variation in earlier-life circumstances such as parental education, family income, and family structure, the magnitude and significance of gender differences in both biomarkers are essentially identical. Similarly, controlling for variation in early adulthood characteristics—health behaviors, educational attainment, and family formation—does not reduce, and in some cases slightly increases, gender differences. This pattern is similar across both outcomes and across the distribution, suggesting that variation in the socioeconomic environments of young adult men and women is not large enough—at least among the overall population—to contribute meaningfully to pronounced differences in biological function.

Tables 3 and 4 also show the extent to which the relationship between other social categories and biomarkers is explained by different circumstances over the early life course, and whether patterns vary across the distribution. Although differences between Hispanic and non-Hispanic white respondents, and between Asians and non-Hispanic whites, are similarly robust to the inclusion of socioeconomic and health measures, black-white differences in CRP (but not EBV) are more substantially reduced by the inclusion of early adulthood circumstances, especially at higher points in the distribution. Early adulthood characteristics also go further in explaining the relationship between high parental education (college or more) and inflammation than for immune function. Finally, differences are pronounced in inflammation according to respondents' education, college-educated respondents having lower levels of inflammation (but not immune function), especially at lower levels of the distribution.

Figures 1 and 2 visualize gender differences in CRP and EBV, respectively. The x-axis on each graph is the percentile and the y-axis is the coefficient size. Each graph, therefore, shows the degree of gender differences in CRP or EBV at different percentiles. The line with shading around it is the gender coefficient from the quantile regression (from the full model, model 3) at a particular percentile and the gray shading graphs the 90 percent confidence interval. The dark horizontal line shows the OLS regres-

sion coefficient from the CRP or EBV regression, respectively. The figures confirm the findings displayed in tables 3 and 4, and show that gender differences at the extremes of the distribution vary from those at the mean. This variation is especially pronounced for the case of inflammation. That these differences are so large at a young age is striking, given evidence that those with high levels of CRP have significantly higher risk for cardiovascular disease and premature mortality.

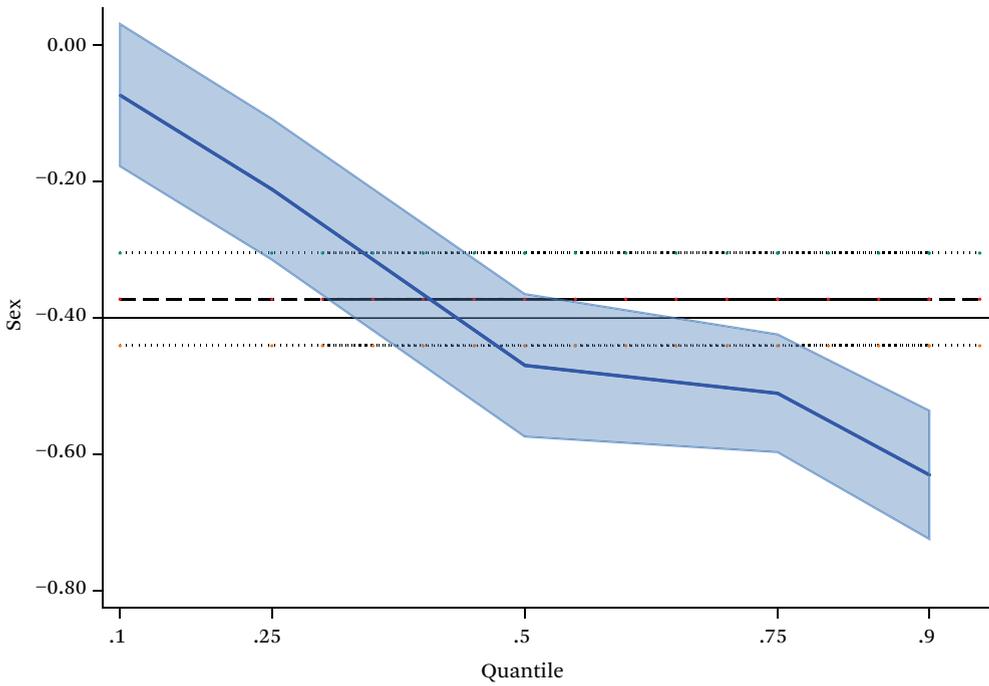
Sensitivity Analyses

We conduct several additional analyses to test the sensitivity of our results to sample restrictions. First, we limit CRP analyses to those with levels at or below 10 mg/L. Because results are nearly identical, we retain these observations in final models. Second, we limit EBV analyses to those who are seropositive, in order to prevent seronegative respondents from biasing coefficients, given evidence that seropositivity is predicted by sociodemographic factors. We use a method established by Jennifer Dowd, Tai Palermo, Laura Chyu, and their colleagues of establishing seronegativity as the bottom 10 percent of continuous EBV antibody values (2013). Coefficients are nearly identical, so we retain seronegative respondents. Finally, we estimate sibling fixed effects models to better control for childhood family environments among siblings. These models, which are identified from siblings who differ in gender, also show a significant gender gap in inflammation and immune function. These results suggest that results from quantile regression analyses are not unduly biased by unmeasured childhood circumstances that are shared by siblings.

DISCUSSION

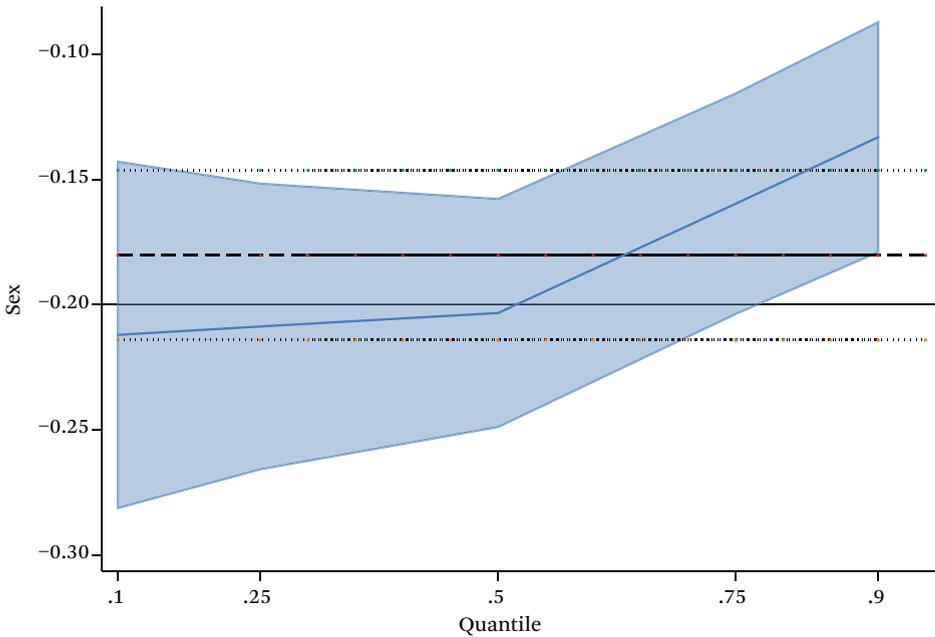
We use nationally representative, longitudinal data from the National Longitudinal Study of Adolescent Health to better understand the relationship between gender and biological function in young adulthood, and to begin to consider the extent to which social and economic circumstances in childhood, adolescence and young adulthood contribute to differences between men and women. We focus on biologic function—specifically, inflammation and immune function—because their pres-

Figure 1. Gender Coefficient for CRP Across Distribution



Source: Authors' compilation based on Add Health, Waves 1–4.

Figure 2. Gender Coefficient for EBV Across Distribution



Source: Authors' compilation based on Add Health, Waves 1–4.

dictive power for future disease risk is high, and because research on gender differences in physiologic risk at young ages is scant.

Results from OLS and quantile regression reveal strong gender differences in both inflammation and immune function among this sample of young adults. Although OLS results offer a useful benchmark, examining variation across the distribution reveals that women have disproportionately higher inflammation and lower immune function relative to their male peers at the top of the distribution. In the case of immune function, gender differences are most pronounced in the middle of the distribution, though women have also lower immune function than their male peers at both extremes. These results suggest that gender differences in physiologic function appear relatively early in the life course, and that substantial differences exist in the early part of individuals' adult trajectories. Considering the contribution of childhood or adolescent and early adulthood circumstances—including demographic factors, family socioeconomic background, health behaviors, and respondents' family formation and socioeconomic attainment—to the magnitude and significance of gender differences yields little evidence of a strong explanatory role for these factors.

Our findings, particularly for the robustness of gender differences in biological function to the inclusion of many indicators of the social environment in our model, inform the debate about how gender differences in health over the life course reflect biological and social variation. Biological explanations for gender differences in health and mortality are well established, as is evidence for variation in the social and economic environments of young adult men and women. By measuring the individual and family-level circumstances and behaviors that may give rise to gender differences in health, our analysis provides a more comprehensive test of the importance of social and economic factors over the life course in explaining gender differences observed in early adulthood. The stability of gender differences in the presence of these factors suggests, at minimum, that variation in the social and economic environments of young men and women in the United States—at least among the overall pop-

ulation—is not large enough to produce the observed pronounced differences in biological function by early adulthood.

Although we do not find evidence for a primary role of early life social environments in explaining gender differences in CRP and EBV in the overall population, we caution against concluding that observed gender differences in these biomarkers are only a function of biology. First, such a conclusion would be inconsistent with evidence from other settings that indicates no mean difference CRP or EBV by sex or gender. Second, our measures of social and economic environments are limited. For example, we are not able to measure cumulative environments as comprehensively as we would like, given that parent-reported data are only available during the first wave. Add Health data do permit us to test a more comprehensive life course model than has been the case in previous research, but it is possible that a more cumulative measure of socioeconomic status—one that captures early-childhood SES in addition to adolescent and young adulthood SES, for example—would yield a larger contribution of the social environment to gender differences in young adulthood. For this to be the case, however, girls and boys would need to experience significantly different childhood environments, and we find little evidence of meaningful variation with our available data (see table 1). Moreover, the results are robust to the inclusion of sibling fixed effects, which capture unobserved, shared family circumstances of male and female respondents. The results are also insensitive to alternative coding and measurement strategies for CRP and EBV, to the exclusion of potential outlier respondents (for example, pregnant women and recently ill respondents), and to controls for health conditions and medication use. Third, this analysis focuses on additive effects and not interactions. While such an approach is a necessary first step, real-world complexity suggests that effects of social environments might be masked if biomarker levels are a function of interactive relationships.

Further, although higher levels of CRP and EBV are associated with risk of chronic disease, including cancer, cardiovascular and metabolic disease, whether a given level in these biomark-

ers translates into similar risk for men and women is difficult to discern from current studies. Existing research is based on different samples and populations, investigates different health outcomes, and produces results that show both equivalent and different risks (Cushman et al. 2005; Han et al 2002; Pai et al. 2004; Yamada et al. 2001). Further, existing practice guidelines do not offer guidance. For example, in 2003, a statement from the Centers for Disease Control and Prevention and the American Heart Association identified over 3 mg/L of CRP as high risk for cardiovascular disease for both men and women, but also noted the need for additional high-quality evidence (Pearson et al. 2003). In short, we emphasize that though we document difference in physiologic function by gender, and patterns and correlates of such difference, the implications for differences in risk are unclear. In future research it will be useful to examine a more comprehensive life course model that includes the critical and sensitive early childhood years. In addition, it will be useful to consider gender variation in the effects of social and economic circumstances during adulthood on biological function. Variation in patterns of family formation between men and women, such as the high prevalence of single motherhood in the United States, mean that adult men and women experience different social stressors that may condition the influence of gender on biological function. Conversely, contemporary patterns of higher educational attainment among women may offset some of the physiologic and health disadvantages associated with low socioeconomic status. Further, additional work on the implications of differences in physiologic function for risk is needed. Examining health within a cohort of young men and women in early adulthood advances our understanding of the intra-generational predictors of pronounced gender differences in biological function, highlights the robustness of those differences, and describes patterns that can be further explored.

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