

Socioeconomic status moderates age-related differences in the brain's functional network organization and anatomy across the adult lifespan

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An individual's environmental surroundings interact with the development and maturation of their brain. An important aspect of an individual's environment is his or her socioeconomic status (SES), which estimates access to material resources and social prestige. Previous characterizations of the relation between SES and the brain have primarily focused on earlier or later epochs of the lifespan (i.e., childhood, older age). We broaden this work to examine the relationship between SES and the brain across a wide range of human adulthood (20–89 years), including individuals from the less studied middle-age range. SES, defined by education attainment and occupational socioeconomic characteristics, moderates previously reported age-related differences in the brain's functional network organization and whole-brain cortical structure. Across middle age (35–64 years), lower SES is associated with reduced resting-state system segregation (a measure of effective functional network organization). A similar but less robust relationship exists between SES and age with respect to brain anatomy: Lower SES is associated with reduced cortical gray matter thickness in middle age. Conversely, younger and older adulthood do not exhibit consistent SES-related difference in the brain measures. The SES–brain relationships persist after controlling for measures of physical and mental health, cognitive ability, and participant demographics. Critically, an individual's childhood SES cannot account for the relationship between their current SES and functional network organization. These findings provide evidence that SES relates to the brain's functional network organization and anatomy across adult middle age, and that higher SES may be a protective factor against age-related brain decline.

socioeconomic status | aging | lifespan | resting-state networks | cortical thickness

Socioeconomic status (SES) is a multifaceted construct that not only represents social standing but also plays a prominent role in shaping our environment by defining our access to healthcare, nutrition, and enrichment (1, 2). To this end, an individual's SES is strongly associated with his or her well-being, including physical health (3, 4), mental health (5, 6), neurodevelopment (7, 8), cognitive ability (4), and the structure and function of his or her brain (for review, see ref. 9).

Examination of the relationship between SES and the brain has primarily focused on the earlier or later stages of the lifespan, which are considered sensitive periods that are typified by significant changes in brain anatomy and function (10). Throughout childhood and adolescence, lower SES has been shown to negatively impact neurodevelopment (for review, see ref. 9), wherein children from lower SES backgrounds exhibit smaller brain volume (e.g., ref. 11) and altered brain function (e.g., refs. 8 and 12). While SES-related brain differences are pronounced in cases of poverty (e.g., refs. 13–15), they also exist across the SES continuum, providing evidence for a SES gradient relationship to measures of the brain and associated cognition (e.g., refs. 8 and 11). On the other end of the age spectrum, studies of older adults have demonstrated a relationship between SES, brain structure,

and health. SES differences are associated with differences in regional and global brain structure (16–20), neuropathology (21), and incidence of dementia (22). While the direction of some of these relationships in older adults varies across studies, it has been hypothesized that higher SES may serve as a “reserve” or protective factor, delaying or minimizing age-related brain deterioration (16, 23). Notably, this work in older adults aligns with studies in animals (rodents and nonhuman primates; refs. 24 and 25) as well as humans (26–28) that have demonstrated how numerous lifestyle and environment factors, such as exercise and environmental stimulation, can maintain or even enhance different aspects of brain structure and function over the course of aging.

It is possible that SES–brain relationships are most pronounced when the brain is most vulnerable: during the earlier and later stages of life. Alternatively, an individual's SES may represent one of a number of life course factors that continually influences brain and cognition over time (29–31), either serving to amplify or limit progressive changes in brain structure and function over age (32–36). Preliminary evidence for this latter hypothesis exists but is incomplete; while a relationship between SES and the brain in both younger adulthood (37, 38) and middle-age adulthood (39–41) have been reported, these studies have typically focused on different brain measures representing separate features of the

Significance

An individual's socioeconomic status (SES) is a central feature of their environmental surroundings and has been shown to relate to the development and maturation of their brain in childhood. Here, we demonstrate that an individual's present (adult) SES relates to their brain function and anatomy across a broad range of middle-age adulthood. In middle-aged adults (35–64 years), lower SES individuals exhibit less organized functional brain networks and reduced cortical thickness compared with higher SES individuals. These relationships cannot be fully explained by differences in health, demographics, or cognition. Additionally, childhood SES does not explain the relation between SES and brain network organization. These observations provide support for a powerful relationship between the environment and the brain that is evident in adult middle age.

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brain (e.g., functional activation during specific tasks, regional brain anatomy), and few studies have simultaneously examined SES–brain relationships across different segments of age. Obtaining a deep understanding of whether and how SES is differentially associated with the brain across adulthood thus requires piecing together results from independent studies that have used distinct measures of SES (e.g., education, income, community disadvantage, constructs of multiple variables) and different measures of brain function and anatomy. This approach has fallen short in producing a clear conclusion (for a recent review, see ref. 9) and limits our ability to understand both the nature and extent of SES's association with brain function and anatomy over distinct portions of the adult lifespan.

In the present study, we evaluate whether SES relates to previously established differences in brain function (33, 42, 43) and anatomy (44, 45) across a wide range of the adult lifespan (age, 20–89 y). Given that SES represents a surrogate measure for diverse domains of life (e.g., education, income, occupational social prestige; ref. 1), it is likely that it relates to the function and structure of multiple brain regions that are distributed across distinct brain systems (e.g., see ref. 9). However, it is possible that different parts of the brain will be differently affected at distinct phases of adulthood, making it difficult to isolate and compare these effects across age. Accordingly, we focused our efforts on examining relationships between SES and “global” measures of an individual's brain function and structure, as opposed to properties of specific brain regions. Measures of brain function and anatomy were operationalized using neuroimaging-based quantification of functional network organization and cortical gray matter thickness, respectively.

To measure brain function, we leveraged recent methodological developments, which have permitted quantification and evaluation of the brain's large-scale functional network organization (46, 47). A measure of brain network organization, system segregation, was defined using formal network models applied to functional interactions between areas of the brain as measured at rest [resting-state functional correlations (RSFCs) (48, 49); Fig. 1 *A* and *B*]. RSFC patterns are malleable over both shorter and longer timescales (i.e., development and aging) in relation to the changing statistics of the environment (47), thus making these functional brain signals a useful target for understanding differences in an individual's SES and life experiences. Importantly,

examining the brain at rest minimizes the influence of differences in task performance that might relate to differences in SES (e.g., refs. 50 and 51). The RSFCs of brain areas are organized into a large-scale brain network that consists of multiple segregated subnetworks or modules, many of which correspond to known functionally dissociable brain systems (e.g., the visual system, the default system, the frontal-parietal control system; ref. 52). The segregation of these brain systems reflects an important organizational feature of an individual's brain network and how it functions (53). We and others have demonstrated that RSFC system segregation decreases with increasing age (for review, see ref. 53) and relates to the task-related functional activity of brain areas (54). Reduced brain system segregation is also associated with poorer cognitive ability across age (33), and training studies have shown that increases in system segregation are associated with improvement in cognitive ability (55).

Following the prior work that has demonstrated relationships between SES and brain anatomy in both childhood (56–59) and adulthood (17, 37, 40, 60), we also evaluated whether SES moderates age-associated differences in brain anatomy. Measurements of brain anatomy were estimated from T1-weighted images of each individual's brain (Fig. 1 *C* and *D*). Automated image segmentation of structural MRI scans is capable of deriving high-precision measurements of an individual's neuroanatomy (e.g., thickness, surface area, volume). This method has been externally validated with both manual tracing and histological analyses of postmortem brains (61–63) to reveal strong associations with cellular neuroanatomy (e.g., neuronal count, cell density, cortical thickness; refs. 64–66). In keeping with this, image-based estimation of gray matter thickness values has been shown to correlate strongly with adult aging (45) and early diagnoses of dementia (67).

Accumulating evidence has demonstrated how both RSFC system segregation and cortical gray matter thickness each exhibit systematic differences across adult age. It is unclear whether and how an individual's environment may be related to differences in these important global measures of brain function and anatomy during adulthood. Here, we tested the hypothesis that SES moderates age-related differences of functional network organization and brain anatomy across a wide age sampling of community-dwelling adult individuals. We provide evidence that an individual's SES relates to their functional network organization and brain structure during their adulthood.

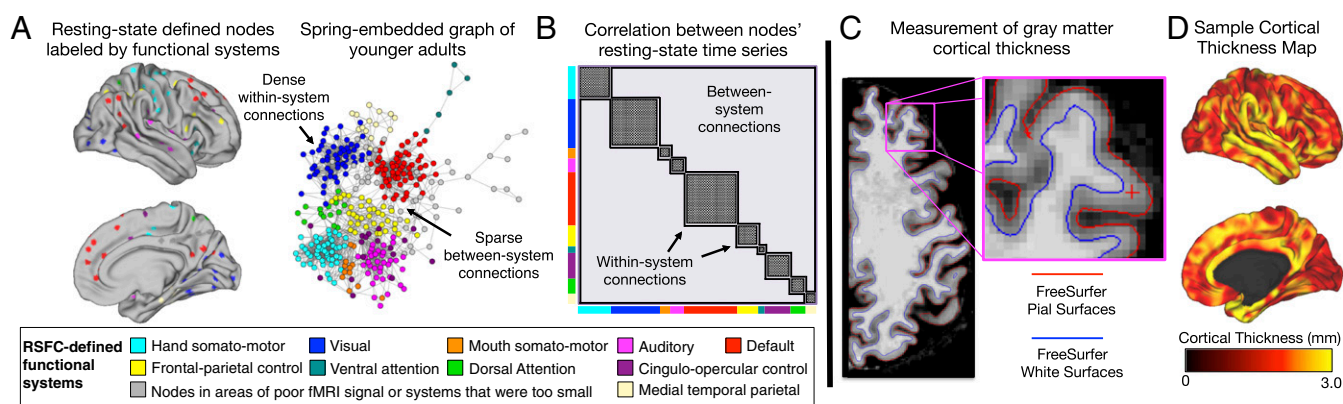


Fig. 1. Functional network and structural measures of the brain. (*A*) Functional network organization was measured using resting-state network analysis. The set of nodes is depicted on a “midthickness” brain surface (*Left*) and with a spring-embedded graph (*Right*). The spring-embedded graph depicts the network organization of the mean younger adult (20–34 y) brain network, where nodes in the same functional system are more connected with each other (i.e., closer in distance) than with nodes from other systems. The nodes shown in the figure are colored by younger adults' functional system assignments. (*B*) The mean resting-state fMRI time series of each node was extracted to form a node-to-node cross-correlation data matrix. Edges between nodes of the same brain system are within-system connections (shaded), whereas edges between nodes of different systems (e.g., connection from a default system node to a frontal-parietal control system node) are between-system connections (not shaded). For each participant's resting-state data matrix, system segregation is calculated using the mean within-system and mean between-system connection strength (see *Materials and Methods* for additional details). (*C*) Cortical gray matter thickness was measured as the distance between the pial (CSF–gray matter boundary; red) and white (gray matter–white matter boundary; blue) surfaces of the brain. (Magnification: 3.87 \times .) (*D*) An example of a single participant's cortical thickness map is shown on a midthickness surface rendering of the right hemisphere.

Results

SES Moderates Age-Related Differences in Functional Network Organization. Participants' ages ranged from 20 to 89 y ($n = 304$), with relatively dense sampling across each of the decades (Table 1). All participants were cognitively normal [Mini-Mental State Examination (MMSE) ≥ 26]. SES was defined from a combination of participants' education years and their occupational socioeconomic index defined for their present occupation (68) [see *Materials and Methods* and additional details in *SI Appendix, Supplemental Results 1.1*, for comparison of the SES measure to other relevant measures (e.g., weighted household income, subjective social standing)]. For retired participants, occupational socioeconomic index was defined from their pre-retirement occupation. Although education level differed across age, with older age being associated with fewer years of formal education, the derived measure of SES was comparable across age [SES by age correlation: $r_{(303)} = -0.06$, $P = 0.271$; see Table 1 for descriptive statistics within specific age ranges].

We evaluated whether SES relates to RSFC brain system segregation as a function of age. A general linear model was conducted to test for the main effects of SES, age, and their interaction on RSFC brain system segregation (controlling for measures of head motion in the scanner). Both SES and age were treated as continuous variables. The main effect of SES was marginally significant in predicting brain system segregation [$\beta = 0.32$, $t_{(299)} = 1.83$, $P = 0.069$], and the main effect of age was significant [$\beta = -0.43$, $t_{(299)} = -7.61$, $P < 0.001$]. Notably, a significant interaction between SES and age was observed in brain system segregation [$\beta = -0.0003$, $t_{(299)} = -1.98$, $P = 0.049$].

To clearly interpret the SES by age interaction on brain system segregation, participants were divided into four independent age groups defined across adulthood [young adults (YA), 20–34 y; middle-early adults (ME), 35–49 y; middle-late adults (ML), 50–64 y; older adults (OA), 65–89 y]. The age groups coincided with the age-based divisions that were used to derive the functional systems of the brain networks, an approach that was also important for minimizing any possible age-related biases when estimating brain system segregation (33, 54) (see *Materials and Methods* for details). This analysis confirmed that SES (included as a continuous variable) and age group exhibited a significant interaction on system segregation [$F_{(3,295)} = 3.16$, $P = 0.025$, $\eta_p^2 = 0.03$].

Additional post hoc analyses were conducted to unpack the SES relationships further. Overall, the SES by age group interaction was characterized by higher SES being associated with greater brain system segregation in the middle-age segments (35–64 y) but not the younger or older age segments. The SES by age interaction is summarized in Fig. 2A, where SES is also stratified into higher and lower groups defined by the median SES of the entire sample. As depicted in the figure, higher SES was associated with greater brain system segregation in the earlier portion of middle age (ME, $t_{(41)} = 2.15$, $P = 0.038$, 95% CI = [0.002, 0.070]), with a similar trend in the later portion of middle age (ML, $t_{(83)} = 1.62$, $P = 0.109$, 95% CI = [−0.006, 0.060]). Significant SES-related differences were not present for system segregation in the youngest and oldest age groups (values of $t < 1.05$, values of $P > 0.296$, 95% CIs cross zero). We note that while these post hoc tests do not survive correction for multiple comparisons, they served to better understand the nature of the differing relationships that drove the SES by age interaction [see *SI Appendix, Supplemental Results 1.2–1.3*, for additional analyses including a broader range of middle age that matches previous work (40, 41)].

In addition, examination of Fig. 2A also suggests that brain system segregation was comparable between lower SES adults at earlier middle age and higher SES adults at later middle age. Direct comparisons of these adults confirmed the observation: brain system segregation did not differ between lower SES ME (35–49 y) and higher SES ML (50–64 y: M (SD) = 0.50 (0.05) vs. 0.49 (0.07), respectively; $t_{(60)} = 0.79$, $P = 0.434$, 95% CI = [−0.020, 0.046]). This comparison suggests that middle-aged adults of lower SES may exhibit earlier signs of brain aging; however, the observations should be interpreted with caution given the cross-sectional study design.

SES Moderates Differences in Brain Anatomy Across Age Groups. A parallel set of analyses examined the relationship between SES and age on mean cortical thickness [adjusting for head size and controlling for measures of head motion in the scanner (69)]. Although the main effect of age was significant for mean cortical thickness [$\beta = -0.58$, $t_{(299)} = -12.15$, $P < 0.001$], neither the main effect of SES [$\beta = -0.08$, $t_{(299)} = 0.52$, $P = 0.604$] nor the SES by age interaction was significant for mean cortical thickness [$\beta = -0.00007$, $t_{(299)} = -0.61$, $P = 0.545$]. However, when age was

Table 1. Demographic information

Variable	Age groups				P value
	Younger (20–34 y)	Middle early (35–49 y)	Middle late (50–64 y)	Older (65–89 y)	
N	44	43	85	132	NA
% Female	59%	63%	66%	55%	NS
% Minority	34%	21%	12%	4%	<0.001***
SES (SD)	0.18 (1.30)	0.21 (1.21)	−0.11 (1.19)	−0.05 (1.22)	NS
Education years (SD)	16.52 (2.46)	16.29 (2.26)	15.65 (2.13)	15.50 (2.29)	0.028*
Occupational socioeconomic index (SD)	45.73 (12.96)	47.44 (13.02)	45.30 (12.99)	47.20 (11.79)	NS
MMSE (SD)	28.41 (1.23)	28.67 (1.13)	28.39 (1.13)	28.11 (1.21)	0.035*
BMI	25.20 (4.82)	24.99 (3.91)	26.54 (4.00)	25.80 (3.64)	NS
SF-36 PCS	56.74 (4.88)	56.99 (4.10)	55.38 (6.93)	53.05 (5.77)	<0.001***
% with chronic physical health issues	23%	14%	20%	31%	NS
% Hypertension	0%	2%	24%	32%	<0.001***
% Never smoker	64%	67%	73%	56%	NS
Alcohol per week (SD)	2.88 (3.73)	2.42 (2.74)	2.31 (3.84)	2.77 (3.75)	NS
Head motion (mean FD; SD)	0.12 (0.03)	0.14 (0.04)	0.16 (0.06)	0.18 (0.06)	<0.001***

Statistical tests between age groups were conducted with χ^2 test for distributions of gender, minority (self-reporting as nonwhite/Caucasian), participants with chronic physical health issues, participants with hypertension, and smoker status; ANOVAs were used for SES, education years, occupational socioeconomic index, Mini-Mental State Examination (MMSE), SF-36 Physical Component Score (PCS), alcohol consumption (drinks per week), and in-scanner head motion during resting state (BOLD fMRI) scan [mean frame displacement (FD)]. Asterisks denote statistical significance (* $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$); NS, not significant; NA, statistical test was not performed.

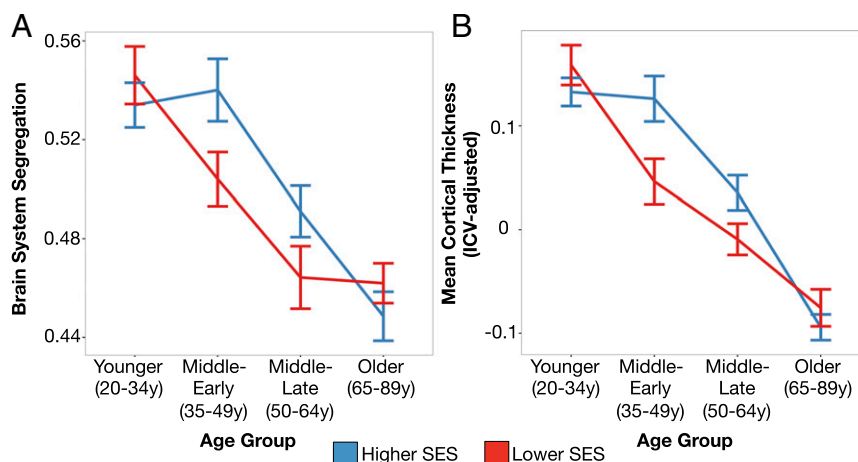


Fig. 2. Lower SES adults exhibit reduced segregation of their resting-state functional brain networks and lower mean cortical thickness in middle-age adulthood. For each age group, brain system segregation (A) and mean cortical thickness (B) are plotted for higher and lower SES (stratified using a median split across the entire participant sample; error bars depict standard error of the mean). Higher SES is associated with greater brain system segregation and mean cortical thickness in middle-age groups (ME, 35–49 y; ML, 50–64 y). Primary statistical models were completed using general linear modeling, where SES was modeled continuously (see *SES Moderates Age-Related Differences in Functional Network Organization* and *SES Moderates Differences in Brain Anatomy Across Age Groups* for details).

examined using the categorical divisions described above, a significant interaction emerged between SES and age group on mean cortical thickness [$F_{(3,295)} = 2.67$, $P = 0.048$, $\eta_p^2 = 0.03$].

Consistent with the functional network observations, the SES and age group interaction was characterized by higher SES being associated with greater mean cortical thickness in the middle-age segments (35–64 y; Fig. 2B). Post hoc comparisons demonstrated that higher SES was associated with greater gray matter thickness in middle age (ME, $t_{(41)} = 2.58$, $P = 0.014$, 95% CI = [0.017, 0.142]; ML, $t_{(83)} = 1.98$, $P = 0.050$, 95% CI = [0.000, 0.090]), but not in younger or older age groups (values of $t < 1.12$, values of $P > 0.271$, 95% CIs cross zero). Also similar to the observations in brain system segregation, cortical thickness was comparable between lower SES adults at early middle age and higher SES adults at late middle age [M (SD) = 0.05 (0.10) vs. 0.04 (0.11), respectively; $t_{(60)} = 0.38$, $P = 0.703$, 95% CI = [−0.046, 0.067]; also see *SI Appendix, Supplemental Results 1.2 and 1.3*, for additional related analyses].

SES by Age Interactions on Global Brain Measures Are Largely Maintained After Controlling for Participant Demographics, Health, or Cognitive Ability. A broad collection of life factors varies in relation to an individual's SES (1, 4, 9), increasing age (70, 71), or both (e.g., refs. 72 and 73). It is critical to examine whether individual differences in these variables might explain any of the described brain observations. Independent statistical models were constructed to include participants' demographics, measures of physical health, measures of mental health, and cognitive ability, while simultaneously controlling for head motion in evaluating the SES by age group interactions on both system segregation and cortical thickness (see Table 2 for a summary).

Controlling for gender and whether or not the participant was classified as a racial minority (i.e., self-reporting as nonwhite/Caucasian) revealed a marginally significant interaction between SES and age group (SES treated as a continuous variable) on brain system segregation [$F_{(3,284)} = 2.47$, $P = 0.062$, $\eta_p^2 = 0.03$] and a significant interaction on cortical thickness [$F_{(3,284)} = 2.81$, $P = 0.040$, $\eta_p^2 = 0.03$]. A statistical model incorporating differences in physical health included a broad range of available variables [i.e., SF-36 Physical Component Score (PCS), body mass index (BMI), hypertension, smoking status, alcohol consumption, and chronic physical health issues; see *Materials and Methods* and *SI Appendix, Supplemental Methods*, for descrip-

tion of each covariate]. The SES by age group interaction remained significant for brain system segregation [$F_{(3,288)} = 3.11$, $P = 0.027$, $\eta_p^2 = 0.03$] and attenuated to marginal significance for cortical thickness [$F_{(3,288)} = 2.26$, $P = 0.082$, $\eta_p^2 = 0.02$]. Finally, controlling for two mental health variables, depressive symptoms and subjective well-being, also did not qualitatively alter the SES by age group interaction on brain system segregation [$F_{(3,292)} = 3.15$, $P = 0.026$, $\eta_p^2 = 0.03$] or cortical thickness [$F_{(3,292)} = 2.81$, $P = 0.040$, $\eta_p^2 = 0.03$].

Broad domains of cognitive ability (e.g., episodic memory, fluid intelligence) vary across age (70, 74), SES (even in middle-age adulthood, the time window where we observed the strongest SES–brain effect; ref. 75), and in relation to brain anatomy (36, 76) and brain network organization (53). In light of these observations, it is important to determine whether the present SES–brain findings apply to individuals across levels of cognitive ability. Aggregate measures representing both long-term episodic memory and fluid processing (intelligence) were incorporated as covariates in statistical models testing for an interaction between SES and age group on brain anatomy and functional network organization. Controlling for episodic memory, the interaction remained significant for brain system segregation [$F_{(3,294)} = 3.41$, $P = 0.018$, $\eta_p^2 = 0.03$] and cortical thickness [$F_{(3,294)} = 2.67$, $P = 0.048$, $\eta_p^2 = 0.03$]; controlling for fluid intelligence, the SES by age group interaction remained significant for brain system segregation [$F_{(3,294)} = 2.97$, $P = 0.032$, $\eta_p^2 = 0.03$] and was marginally significant for cortical thickness [$F_{(3,294)} = 2.23$, $P = 0.085$, $\eta_p^2 = 0.02$].

Individual Differences in Childhood SES Do Not Explain SES-Related Brain Network Organization Differences in Adulthood. Given the strong influence of childhood SES on brain maturation and function (i.e., shown in children and adults; refs. 5, 20, 57, 59, and 77), it is possible that the presently observed SES-related brain differences are a consequence of preestablished brain differences carrying forward from childhood. Information related to childhood SES (parental education) was available for a subsample of our study participants ($n = 168$). Childhood SES was defined as the highest educational degree completed by either parent (*Materials and Methods*). The childhood-SES measure was significantly associated with an individual's present (“adult”) SES [$F_{(6,161)} = 2.94$, $P = 0.010$, $\eta_p^2 = 0.10$], and also age group ($\chi^2 = 43.78$, $P < 0.001$), where lower present SES and

Table 2. Covariate analysis summary

Covariates	SES × age group (<i>n</i> = 304)				SES × age group controlling for childhood SES (<i>n</i> = 168)			
	Brain system segregation		Cortical thickness		Brain system segregation		Cortical thickness	
	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>
Head motion only	3.16	0.025*	2.67	0.048*	2.85	0.039*	0.65	NS
Demographic + head motion	2.47	0.062 ⁺	2.81	0.040*	2.79	0.042*	0.97	NS
Physical health + head motion	3.11	0.027*	2.26	0.082 ⁺	2.58	0.056 ⁺	0.78	NS
Mental health + head motion	3.15	0.026*	2.81	0.040*	2.59	0.055 ⁺	1.22	NS
Fluid intelligence + head motion	2.97	0.032*	2.23	0.085 ⁺	2.73	0.046*	0.56	NS
Long-term episodic memory + head motion	3.41	0.018*	2.67	0.048*	2.98	0.033*	0.68	NS

Summary of SES by age group interactions on brain system segregation and mean cortical thickness: controlling for multiple sets of covariates (left), controlling for covariates and childhood SES in a subsample of participants who had childhood-SES information available (right). Head motion refers to estimated measures of participants' in-scanner head motion (mean FD); demographic variables include sex and race; physical health variables include BMI, SF-36 PCS, hypertension, smoker status, alcohol consumption, and chronic health issues; mental health variables include depressive symptoms and life satisfaction; fluid intelligence and episodic memory are factor scores generated from multiple tasks (see *SI Appendix, Supplemental Methods*). Asterisks denote statistical significance, **P* < 0.05; crosses denote marginal effects, ⁺*P* < 0.10; NS, not significant.

older age were associated with having parents with less education (*SI Appendix, Table S1*).

Controlling for childhood SES, the interaction between the participants' present SES and age group remained significant for brain system segregation [$F_{(3,153)} = 2.85$, $P = 0.039$, $\eta_p^2 = 0.05$]. Furthermore, this SES by age group interaction on segregation largely remained significant, or showed very minor attenuation in effects, when controlling for childhood SES in conjunction with the numerous sets of covariates (Table 2 and *SI Appendix, Supplemental Results 1.4*). The interaction between SES and age group was not significant for cortical thickness when controlling for childhood SES [$F_{(3,153)} = 0.65$, $P = 0.583$, $\eta_p^2 = 0.01$], and the interaction remained nonsignificant when controlling for additional covariates (Table 2).

It is possible that there exists a cumulative effect of an individual's childhood SES and their present SES on their functional network organization and brain anatomy, as a function of age. When modeled together with age group, the main effect of childhood SES [values of $F_{(6,145)} \leq 1.76$, values of $P > 0.111$] and its interaction with age group did not significantly predict either brain measure [values of $F_{(12,145)} \leq 1.13$, values of $P > 0.337$]. Additionally, no significant interactions were observed between childhood SES, present SES, and age group for either brain measure [values of $F_{(9,126)} < 0.51$, values of $P > 0.863$]. We note that a high collinearity was observed between childhood SES and age group, possibly influencing the observations involving childhood SES with respect to age (*SI Appendix, Supplemental Results 1.5*).

Discussion

SES moderated previously reported age-related differences in large-scale resting-state functional network organization. A similar but less robust interaction was evident between SES and age on brain anatomy. SES was operationalized as a combination of educational attainment and occupational socioeconomic characteristics, factors that have been previously shown to relate to health and well-being outcomes (1). Middle-aged participants with lower SES exhibited reduced segregation of the systems in their large-scale functional brain networks and thinner mean cortical gray matter, compared with higher SES individuals in an equivalent age range. Critically, the relationships between SES and both brain measures were largely present while controlling for differences in participant demographics, physical and mental health, and cognitive ability. A measure of childhood SES was available for a subset of participants and was included in statistical models: the interactions between an individual's present

(adult) SES and age persisted while controlling for childhood SES on brain system segregation. By measuring features of brain function and anatomy within individuals who have been densely sampled across the adult lifespan, these collective observations provide support for a compelling relationship between the socioeconomic aspects of the environment and the brain that is evident in middle-age adulthood.

SES Relates to Global Measures of Brain Function and Anatomy. In addition to its adult lifespan focus, a unique aspect of the present work is the inclusion of a measure of large-scale functional network organization. Studies of SES and brain function have primarily examined task-related functional activity in specific regions of the brain and under specific task demands (8, 78, 79). Due to the specificity of these observations, it has been difficult to gain a broader understanding of SES–brain function results and their potential impact on cognitive ability. We hypothesized that SES should be associated with brain function across distributed and widespread regions, as SES represents broad environmental and lifestyle variation in a person's life. Accordingly, we examined the relationship between SES and a brain measure that captures the brain's large-scale functional network interactions (for additional follow-up analysis focusing on more specific system distinctions, see *SI Appendix, Supplemental Results 1.6*). The segregated organization of brain systems (i.e., subnetworks) observed in healthy young adults facilitates functional specialization and efficient communication across systems (80). Multiple studies have shown that brain systems exhibit less segregation with increasing age in healthy adults (33, 42, 43). Reduction in segregation is related to poorer cognitive performance (33) and differences in brain function (54) in older adults. In the present study, we found that SES moderates the relationship between age and brain system segregation. Specifically, lower SES individuals exhibited reduced system segregation in middle-age adulthood (35–64 y), compared with higher SES individuals. Critically, this relationship remained significant after controlling for cognitive abilities related to aging and system segregation.

A less segregated functional brain network reflects a “blurring” of distinct and modular systems of the brain. One of us (G.S.W.) has previously speculated that reduced brain system segregation may characterize or result in diminished resilience against focal brain damage (53). When brain systems are segregated, the impact of localized damage is more likely to be confined within specific brain systems (81–83), whereas comparable damage to a less segregated network should result in a failure of function that propagates across multiple systems. This idea echoes previously described “brain reserve” concepts (23) and

individuals with lower SES may systematically experience greater amount of continual stress related to their environment in comparison with higher SES individuals resulting in allostatic load (for review, see ref. 103), a factor known to result from chronic dysregulation of physiological systems (e.g., neuroendocrine, immune, and autonomic nervous function; refs. 104 and 105) and impact brain anatomy (e.g., refs. 106 and 107). While the brain observations reported here would look identical under either of these scenarios, this latter possibility suggests an entirely different mechanism to the presently reported SES–brain relationships.

Relevant evidence has been provided from studies that have examined the relationship between neighborhood-level factors and brain anatomy. For example, a recent study of middle-aged adults (30–54 y) revealed a relationship between community-level socioeconomic factors and cortical anatomy (40). Reduced cortical volume and greater community disadvantage was mediated by cardiovascular risk and neuroendocrine function. This finding aligned with an earlier study of middle-age adulthood (35–64 y) demonstrating that the relationship between neighborhood level deprivation and regional anatomical differences (cortical thickness in supramarginal gyrus, superior temporal gyrus, and lateral sulcus; described as “Wernicke’s area,” and its right homolog) was mediated by an inflammation factor (a cardiometabolic risk marker) (41). These studies reveal how the relationship between socioeconomic factors and brain anatomy may be mediated by both neuroendocrine and cardiometabolic pathways. While the present study controlled for measures of physical health in each individual (i.e., BMI, hypertension), other physiological risk factors that were not surveyed may be found to mediate the relationship between an individual’s SES and their brain.

Although the design of the current study could not distinguish the possible difference between “protective” higher SES vs. “harmful” lower SES, the results demonstrate that SES is an important factor to account for when investigating age-related differences in the brain. Furthermore, as the stratification of SES in our study is not defined by a specific threshold (e.g., the poverty line), our results suggest that the relationship between SES and the brain in middle-aged adults may exhibit a gradient pattern where individuals with relatively higher SES reap environmental benefits compared with those with lower SES, similar to SES gradients in health and cognition (1, 108). Altogether, SES may relate to the aging brain in multiple ways. Whereas engaging and resourceful environments associated with higher SES may provide a buffer (or delay) against aging, inadequate health conditions associated with lower SES environments (e.g., exposure to toxins, poorer nutrition) together with continual stress may accelerate the aging process (e.g., refs. 30, 100, and 106) and/or make the brain more vulnerable to damage.

Limitations and Future Directions. It is important to acknowledge that while considerable effort was placed to recruit participants with lower SES, the necessary exclusion criteria of the study can still result in a relatively selective sample that prohibits examination of a broader range of SES due to the relation between SES and certain exclusion criteria (e.g., BMI, psychiatric disorders; refs. 109 and 110).

In addition, it is possible that some of the brain differences observed across participants in middle age relate to their environment during childhood and adolescence. We attempted to control for this possible source of influence using a subsample of participants with childhood-SES data. Childhood SES (when modeled together with age) did not exhibit a significant relationship with either brain measure, and there was an absence of any three-way interactions between an individual’s childhood SES, their present SES, and age on either brain measure. However, given the limitations in both the childhood-SES measure and number of participants with available childhood-SES data, additional work is needed to further understand the pos-

sible cumulative and interactive relationships between childhood SES and present SES on the brain.

Importantly, the cross-sectional design of this study prohibits conclusions to be made regarding age-related changes in brain function and anatomy within an individual as a function of their SES. In keeping with this, and as noted earlier, we cannot rule out the possibility of cohort-specific effects. Cross-cohort comparisons enable the examination of age-related differences across a broad segment of ages, but these differences in age are also inherently tied to differences in the sociocultural environments under which the individuals were born and raised (e.g., propensity for SES mobility, wartime, times of broad economic hardship). Relatedly, the correlational aspects of analysis also prohibit conclusions about SES–brain directionality. For example, it is entirely possible that differences in brain functional organization or anatomy earlier in the lifespan modify the SES of middle-aged or older participants. The SES observations reported here motivate future studies examining brain aging using longitudinal designs to investigate whether SES moderates the rate and direction of brain change across the adult lifespan and, critically, how these changes relate to changes in cognition.

Given the relation between SES and both brain measures during middle age, a promising goal for future studies is to dissect and even manipulate the features that are represented by SES and its relation to the brain in adults. As SES represents a complex and multifaceted construct, it is extremely difficult to manipulate as a whole (e.g., ref. 111). However, there are numerous opportunities to influence certain aspects of an individual’s environment that may relate to the features differentiated by SES. These interventions include programs designed to enhance social and cognitive engagement (98, 112), increase active learning (100), reduce stress (113), and improve physical fitness (28, 97). While many of these interventional studies have targeted older adults, our results suggest that expanding the focus to middle age is an important goal.

Conclusion

An individual’s SES, defined by educational attainment and occupational socioeconomic characteristics, relates to their functional network organization and brain anatomy across broad segments of life. Compared with middle-aged adults from higher SES, lower SES adults exhibited signs of both functional and structural brain aging earlier in adulthood. Prevailing work has highlighted the impact that SES has on brain and cognition in both childhood but also advanced age. We provide evidence that there exists a powerful relationship between an individual’s present environment and their brain well beyond these specific segments of life.

Materials and Methods

Participants. Participants were recruited from the Dallas–Fort Worth community and provided written consent before participating. Study procedures were reviewed and approved by the Institutional Review Boards at the University of Texas at Dallas and the University of Texas Southwestern Medical Center. Participants were part of the Dallas Lifespan Brain Study (DLBS), a study designed to examine the effects of healthy aging on brain and cognition. The current study includes data from participants that completed both an anatomical and a resting-state fMRI scan, and reported their occupation/occupation-before-retirement in their demographic survey ($n = 359$; age range, 20–89 y; mean age, 59.79; SD, 16.89).

The entire participant sample shares the exclusion criteria typically employed in studies involving MRI: BMI of >35 , loss of consciousness of >10 min, radiation/chemotherapy in the last 5 years, various diseases (e.g., epilepsy, lupus, stroke, multiple sclerosis, Parkinson’s, Alzheimer’s) or major psychiatric disorders [e.g., manic-depressive (bipolar) disorder, schizophrenia, depression (including if individuals were in remission without treatment for >6 mo), attention deficit or learning disorder (including if the disorder was resolved by high school graduation)], electroshock therapy for depression, brain surgery, excessive alcohol or caffeine consumption (self-report), blood pressure of $>160/90$, unprescribed/illegal drug use, use of sedatives, use of benzodiazepines, use of antipsychotics, coronary bypass, and an MMSE (114) score of ≤ 25 . Despite these restrictions, an important feature of the DLBS is that it includes recruitment of participants who more

broadly represent the education level and health condition of middle-aged and older adults in the general population. The recruitment of this subset of participants targeted individuals that were 50 y and older with lower educational attainment (i.e., no college degree; ≤ 14 y of formal education); exceptions were made for individuals with education of 15 or more years, if they also have a chronic health condition (e.g., hypertension, diabetes, cancer, or heart surgery).

Brain imaging data were subjected to substantial quality control to remove participants with poor data quality or excessive head movement (see below); 304 participants remained in the final study sample. Participants were separated into four age groups for the identification of age group-specific functional brain systems (see below), and the statistical analyses used to further understand the continuous SES by age interaction. Each age group contained 15 y until the typical cutoff age for older adults that corresponds to the average age of retirement in the United States (65 y): YA, 20–34 y; ME, 35–49 y; ML, 50–64 y; OA, 65–89 y. See Table 1 for detailed demographic information per age group.

SES was assessed, on average, within 3 wk of the MRI scan (mean time, 22 d; range, 1–205 d; 236 out of 304 participants were tested within 30 d). Accordingly, all participants' anatomical and functional neuroimaging data were collected in close proximity to the measurement of SES-related variables (education and occupation or occupation before retirement).

Experimental Design and Data Acquisition. The DLBS consists of multiple data acquisition sessions of cognitive and neuropsychological testing, MRI scanning, as well as take-home surveys on demographic and psychosocial data. All functional and anatomical MRI scans were acquired on a Phillips Achieva 3T scanner. Brief descriptions of acquisition and processing of structural and functional images are provided below; additional details are available in [SI Appendix, Supplemental Methods](#).

Structural Imaging Acquisition and Preprocessing. For each participant, a T1-weighted magnetization-prepared rapid-acquisition gradient echo structural image was obtained (repetition time, 8.1 ms; echo time, 3.7 ms; flip angle, 12°; field of view, 204 × 256 mm; 160 slices with 1 × 1 × 1-mm voxels). The volumetric image was used to construct the participant's cortical surface, which was then deformed to the fsaverage surface using FreeSurfer, version 5.3 (115, 116). The left and right fsaverage surfaces were then registered to a hybrid left-right fsaverage atlas (fs_LR; ref. 117).

Resting-State fMRI Acquisition and Preprocessing. Participants completed an eye-open fixation resting-state blood oxygen level-dependent (BOLD) scan. The functional images went through standard fMRI preprocessing to reduce artifacts (e.g., slice-timing correction, realignment). Then, RSFC preprocessing was used to reduce spurious variance unlikely to reflect neuronal activity: (i) demean and detrending; (ii) multiple regression of the BOLD data to remove variance related to the whole-brain signal, ventricular signal, white matter signal, their derivatives, and the "Friston24" motion regressors (118); and (iii) bandpass filtering (0.009–0.08 Hz). Last, motion-contaminated resting-state fMRI volumes [i.e., if frame displacement (FD) > 0.3 mm] were flagged ("scrubbing"), discarded, and interpolated for subsequent nuisance regression and bandpass filtering (119). Preprocessed resting-state data were registered to the fs_LR (32 k) left and right hemisphere surfaces for analysis (116).

Brain Graph Construction and Brain System Assignments.

Nodes and edges definition. RSFC brain graphs were constructed using a modified set of published nodes applied to the surface-mapped resting-state data (Fig. 1A) (33). The cross-correlation of each node's mean time course was incorporated into a node-to-node correlation matrix with Fisher's z-transformed correlation values for each participant (Fig. 1B; for details, see [SI Appendix, Supplemental Methods](#)). Due to possible artifactual negative correlations (120, 121) introduced by a necessary processing step used to ensure removal of motion-related artifacts (119, 122), negative z values were excluded from the data matrix in accordance with past studies (33, 123).

Age group-specific functional brain system assignments. To eliminate biases introduced by using younger adult-defined brain systems (123), which may provide a better fit for the younger compared with the older adults, large-scale functional brain systems were identified in each age group by applying Infomap community detection (124) to bootstrapped, thresholded mean matrices of each age group (2–10% edge density, 1,000 iterations). The final system assignment was based on the most common (modal) assignment across 2–10% edge density (see assignments for younger adults in Fig. 1A). The community assignments were labeled based on their overlap with a set of published RSFC functional systems (123). While age was treated as a continuous variable in the primary analyses, follow-up analyses treated age

as a categorical variable to maintain the same level of granularity used in calculating system segregation from age group-specific system assignments.

Measures.

Brain measures.

Measure of functional network organization: System segregation score. System segregation (33) takes the differences in mean within-system and mean between-system correlations (Fig. 1B) as a proportion of mean within-system correlation, as noted in the following formula:

$$\text{System Segregation} = \frac{\bar{Z}_w - \bar{Z}_b}{\bar{Z}_w},$$

where \bar{Z}_w is the mean Fisher's z-transformed Pearson's correlation (z) across nodes within the same system, and \bar{Z}_b is the mean z between nodes of one system to all nodes in other systems. Values along the diagonal of the matrix were not included. Importantly, the measure of system segregation retains the weight of all positive edges in a graph, allowing weak connections to contribute to the characterization of system interactions.

Measure of anatomical brain structure: Cortical thickness. Cortical thickness was measured for each participant by estimating the distance between the pial and white matter surfaces at every brain vertex generated by FreeSurfer following quality-control procedures, which included manual editing to correct for brain segmentation errors (Fig. 1C and D and ref. 69). The mean whole-brain cortical thickness was calculated by averaging across the left and right hemispheres' average cortical thickness values. All statistical analyses were conducted on mean cortical thickness that was adjusted for intracranial volume (ICV) (obtained from automated FreeSurfer output): ICV-adjusted cortical thickness.

SES. Using a combination of estimated years of education and occupational socioeconomic index (68), an SES score was derived for each participant (125). Years of education and occupational socioeconomic index were significantly correlated [$r_{(302)} = 0.49$, $P < 0.001$]. Principal component analysis was used to extract a common factor between the two variables, with the first factor score representing SES (75% variance explained by the first factor).

Education attainment. Years of education were estimated by degree completion (e.g., high school/GED, college, graduate school) plus the extra years of education beyond the completed degree that were self-reported. Extra education years were capped to be one less than the next possible degree. For example, a participant with 5 additional years after a high school degree (estimated to take 12 y) but never completing college would have an estimated 15 y of education (1 y less than completion of college, which is estimated to take 16 y). The maximum years of education was capped at 22 (e.g., PhD, MD).

Occupational socioeconomic characteristics. Participants' self-reported occupation (current or preretirement) was matched to a corresponding US census occupation code, and then assigned a gender-specific socioeconomic index score derived from predicted prestige score (i.e., composite score based on estimated occupational wages, occupational education, and a wage–occupation–prestige index; ref. 68). While this index is somewhat subjective, among our participants with additional SES measures ($n = 168$), it is correlated with weighted household income [$r_{(166)} = 0.27$, $P < 0.001$] and the MacArthur's scale of subjective socioeconomic status (126) [$r_{(166)} = -0.22$, $P = 0.004$; see [SI Appendix, Supplemental Results 1.1](#), for more details]. If an occupation was not listed, the code from the most related job was used. Consensus of coding for nonlisted occupations was established between two coders (M.Y.C. and J.N.). Since full-time student and homemaker status do not provide a clear approximation of occupation-related income and prestige, participants listing their occupation as student ($n = 19$) or homemaker ($n = 5$) were not included in the current study.

Childhood SES. For a subset of participants with available information ($n = 168$), parental education was used to determine their SES during childhood. Information regarding parental occupation was not available. For analyses controlling for childhood SES as a covariate, participants' childhood SES was defined by the highest degree either parent completed, coded as a seven-level categorical variable ([SI Appendix, Table S1](#)).

Measures of demographics, health, and cognitive ability. An extensive set of measures was used as covariates to control for participants' demographics, health (physical and mental), and cognition. For each set of variables, statistical models were created to examine the relationship between SES and age with respect to each of the brain measures while controlling for the corresponding covariates of interest. A summary of each set of covariates is provided in [SI Appendix, Supplemental Methods](#).

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