



Original Investigation | Psychiatry

# Socioemotional and Executive Control Mismatch in Adolescence and Risks for Initiating Drinking

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## Abstract

**IMPORTANCE** Maturation imbalance between the expression of heightened socioemotional drivers and immature executive control creates a period of vulnerability to risk-taking behaviors during adolescence. Developmental imbalance of the brain functional activity coupled with these behaviors may explain the propensity to develop heavy drinking.

**OBJECTIVE** To compare the longitudinal trajectories of socioemotion and executive control in non-heavy drinkers with those of heavy drinkers before and after drinking onset.

**DESIGN, SETTING, AND PARTICIPANTS** This cohort study was conducted from January 13, 2013, to January 15, 2022, among participants from the National Consortium on Alcohol and Neurodevelopment in Adolescence cohort with resting-state functional magnetic resonance imaging and neuropsychological assessment data. All participants were individuals who abstained from drinking or drank very little at the baseline visit. The statistical analysis was conducted between October 11, 2024, and July 3, 2025.

**EXPOSURES** Categorical drinking levels that summarized self-reported alcohol consumption in the past year.

**MAIN OUTCOMES AND MEASURES** Brain-behavioral system z scores that reflected the association of selective brain connectivity patterns with selected behaviors were calculated. Higher scores indicated higher reactivity strength of socioemotion or executive control.

**RESULTS** A total of 3076 visits among 633 participants (mean [SD] age at baseline, 15.7 [2.6] years; 318 female [50.2%]) were analyzed. A total of 238 participants (37.6%) who initiated heavy drinking during the study showed significantly elevated brain-behavior scores (score increase, 0.18; 95% CI, 0.08-0.28) in the socioemotional system prior to drinking onset, in contrast to their developmental trajectories of the executive control deviation occurring in both systems (score increase, 0.44; 95% CI, 0.33-0.54), with the effect size in the socioemotional system significantly greater ( $z = 3.51$ ;  $P < .001$ ) compared with before initiating drinking.

**CONCLUSIONS AND RELEVANCE** These findings suggest that asynchronized maturation of socioemotion- and executive function-coupled brain connectivity in adolescence may be a risk factor for heavy drinking onset in emerging adulthood. In turn, drinking was associated with deficits in the executive control system and exacerbated alteration in the socioemotional system. Recognition of these early-age developmental discrepancies could aid in averting alcohol use disorder in adulthood.

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## Key Points

**Question** What is the association of imbalances between socioemotional and executive control brain-behavior systems in adolescence with risk for heavy alcohol drinking in emerging adulthood?

**Findings** In this cohort study of 633 participants aged 12 to 21 years at baseline, faster development of the socioemotional system was associated with drinking onset in individuals who initiated heavy drinking compared with those who refrained from initiating heavy drinking. Heavy drinking was associated with deviation in the executive control system and exacerbated socioemotional dysregulation.

**Meaning** These findings suggest that maturational imbalance may be a risk factor for heavy drinking onset, which in turn may pose risks for developing alcohol-induced damage of both brain-behavior systems.

## + Supplemental content

Author affiliations and article information are listed at the end of this article.

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## Introduction

Adolescence is a period marked by emotionally driven choices and heightened impulsivity, often occurring before the full maturation of behavioral control mechanisms.<sup>1,2</sup> This imbalance stems from brain-behavior systems that may develop at different rates. The socioemotional system becomes highly active and responsive to social rewards, peer influence, and emotional stimuli, often motivating impulsive actions.<sup>3,4</sup> Its reactivity peaks in midadolescence and begins to stabilize in late adolescence to early adulthood as brain connectivity matures.<sup>5</sup> Contemporaneously, the executive control system, which supports decision-making, planning, and impulse regulation,<sup>6,7</sup> matures more gradually compared with the socioemotional system, with supportive brain structural and functional changes continuing substantially into the mid-20s.<sup>8,9</sup> The asynchronized development of this dual system, that is, maturational imbalance, may be a cognitive mechanism that heightens risk taking,<sup>2,5</sup> including heavy alcohol drinking, observed during adolescence and young adulthood.

Recent neuroimaging research has shed light on the effects of heavy alcohol consumption on the adolescent brain. Longitudinal structural magnetic resonance imaging (MRI) studies have revealed that heavy alcohol consumption during adolescence is associated with abnormally fast reductions of gray matter volume in regions such as the prefrontal cortex<sup>10-12</sup> and hippocampus and with compromised microstructural integrity in the frontal commissural tracts.<sup>13,14</sup> Functional imaging has shown altered neural activity in emotion,<sup>15</sup> default-mode,<sup>16</sup> and sensorimotor networks<sup>17</sup> linked to impaired neuropsychological functions such as decision-making, impulse control, emotional dysregulation, and reward processing.<sup>18-21</sup> Despite widely replicated neuroimaging findings, there remains an incomplete and incomprehensive understanding of how imbalanced neurodevelopment of the dual system contributes to drinking onset and how alcohol, in turn, dysregulates socioemotional function and dampens executive functional control that should otherwise provide protection against engagement in risky behavior. Addressing this lacuna requires identifying neural systems linked to the 2 behavior domains of socioemotion and executive control, estimating their concurrent developmental trajectories, and comparing normal trajectories with those of alcohol drinkers both before and after initiating heavy drinking.

Based on 9-year longitudinal resting-state functional MRI and neuropsychological testing data collected by the National Consortium on Alcohol and Neurodevelopment in Adolescence (NCANDA) project,<sup>22</sup> the current study used a data-driven machine learning approach (eFigure 1 in [Supplement 1](#)) to discern developmental trajectories of the 2 brain-behavioral domains linked to socioemotion and executive control. By comparing trajectories among individuals who drink heavily (hereafter referred to as heavy drinkers) and those who do not (hereafter referred to as non-heavy drinkers) before and after drinking onset, we aimed to provide direct neurobiological evidence to support the conceptual hypothesis that heavy drinking behavior would emerge when socioemotional development outpaces the maturation of the executive control system. In turn, we investigated whether heavy drinking is associated with disrupted executive control functions and exacerbated socioemotional dysregulation.

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## Methods

### Participants

This cohort study used NCANDA data from participants aged 12 to 21 years at baseline recruited across 5 collection sites in the US from January 13, 2013, to January 15, 2022, and assessed annually for psychobiological measures. The institutional review boards of each site approved data collection and use. Adult participants and parents of minor participants provided written informed consent before participation in the study. Minor participants provided assent before participation. This study followed the Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) reporting guideline.

Based on self-reported alcohol use history released throughout the first 9 years of the study, drinking levels of participants were defined based on the youth-adjusted Cahalan score on a scale of 0 to 3.<sup>10,23</sup> Heavy drinkers (Cahalan = 2 or 3) ranged from moderate frequency (eg, 2 times per month) with high-quantity consumption (eg, 3-4 drinks on average and >4 drinks maximum) to higher frequency (eg,  $\geq 1$  time per week) with moderate-quantity consumption (eg, 2-3 drinks on average and >4 drinks maximum). The current analysis focused on participants who abstained from drinking or drank very little (Cahalan = 0) at their baseline visits, had no brain structural anomaly,<sup>24</sup> and had usable resting-state functional MRI scans during the study period. Given the right-tail skew of the age distribution (eFigure 2 in Supplement 1), we confined the maximum age to 26 years to avoid bias from the limited data points at older ages.

### MRI Data

The structural and resting-state brain data of all NCANDA participants were preprocessed using the publicly available NCANDA pipeline,<sup>17</sup> which included motion correction, outlier detection, detrending, physiologic noise removal, and temporal and spatial smoothing. The pipeline produced the mean blood oxygen level-dependent signal from 109 brain regions defined by the SRI24 Atlas,<sup>25</sup> including 53 bilateral regions and 3 cerebellum vermis regions. To reduce dimensionality, we computed a Pearson correlation coefficient of the blood oxygen level-dependent signal between each pair of regions and averaged the correlation values of the left and right hemispheres,<sup>26</sup> resulting in 1378 functional connectivity measures for each participant visit. Following previous work,<sup>13</sup> we adjusted the potential confounding effects of socioeconomic status (defined by the maximum years of education of either parent), data collection site, and self-reported race (Asian, Black, White, other [American Indian, Pacific Islander, multiracial]) from each connectivity measure by a linear mixed-effects regression and normalized the residualized measures to z scores. These variables were based on prior findings of the NCANDA dataset that they are associated with neuroimaging and neuropsychological data and may confound the analysis.<sup>27,28</sup> Finally, we used principal component analysis to reduce the dimensionality from 1378 to 384 to preserve 90% of the variance in the data based on prior practice.<sup>29,30</sup>

### Neuropsychological Measurements

Measures were from 17 types of neuropsychological test batteries (eg, Penn Computerized Neurobehavioral Test Battery,<sup>31</sup> delay discounting,<sup>32</sup> grooved pegboard<sup>33</sup>) or behavioral questionnaires (eg, Urgency-Premeditation-Perseverance-Sensation Seeking-Positive Urgency Impulsive Behavior Scale,<sup>34</sup> Behavior Rating Inventory of Executive Function,<sup>35</sup> Alcohol Expectancy Questionnaire<sup>36</sup>). Each test battery or questionnaire was categorized into 1 of the 2 domains based on their primary intended use noted in the literature. For a test battery or a questionnaire that contained many subscale measures, we kept only a few summary scores to reduce the overall dimensionality (eg, of the 53 recorded subscales of the Stroop test,<sup>37</sup> only the mean and SD of the response time over all correct trials were kept). Next, we removed measures that were missing for more than half of the participants and removed participant visits in which more than half of the neuropsychological measures were missing. This process resulted in 82 neuropsychological measures assigned to the executive functioning system and 44 measures assigned to the socioemotional system (eTables 1 and 2 in Supplement 1). For each measurement, a linear mixed-effects model regressed out socioeconomic status, data collection site, and race<sup>13</sup> and normalized the residualized measures to z scores. Finally, missing data were imputed by a k-nearest neighbor imputation algorithm, as previously described.<sup>19,38</sup>

### Statistical Analysis

#### Canonical Correlation Analysis

A canonical correlation analysis (CCA)<sup>29,39</sup> was conducted for each system separately (eFigure 1A in Supplement 1) to identify the high-dimensional covariate pattern between the 384 functional

principal component analysis scores and neuropsychological measures across participant visits. The CCA was evaluated by a 10-fold participant-level cross validation. During training, CCA derived a set of canonical components, each represented by a pair of linear transformations of functional and neuropsychological measures that were highly correlated (a brain-behavioral mapping) (eFigure 1B in Supplement 1). The correlation value between the 2 transformed variables in each component was then computed on the testing fold and averaged over the 10 folds after cross validation. Statistical significance of the correlation values was determined by permutation tests and corrected by the number of neuropsychological measures in each system based on Bonferroni correction. The threshold for significance was set at  $P \leq .05$ .

Next, we identified which individual neuropsychological measures contributed most to brain-behavior mapping. For a significant component, we averaged the canonical loading (eMethods in Supplement 1) of each neuropsychological measure over the 10 folds and used a permutation test to identify measures with significantly large canonical loadings. Finally, for each participant visit, the overall reactivity strength of a component (hereafter referred to as a brain-behavior score) was computed as the average between brain functional and neuropsychological canonical variables (ie, linear transformations of functional and neuropsychological measures) (eMethods in Supplement 1) after z score transformation.

### Association Between Heavy Drinking and Dual-System Development

For each significant component, we compared the developmental trajectory of the brain-behavior score of non-heavy drinkers (Cahalan = 0 or 1) to the trajectory of heavy drinkers before and after drinking onset (eFigure 1B in Supplement 1). To examine whether there was developmental deviation predated drinking onset, a mixed-effects model with participant-specific intercepts was used to test the difference in brain-behavior scores between the 2 drinking groups, with age and sex as covariates. For the heavy drinkers, only participant visits prior to drinking onset were included in this regression. A stepwise approach<sup>40</sup> was used to determine the highest-order polynomial term associated with age to account for the potential nonlinear developmental trajectory of a component. Statistical significance of the drinking effect was corrected by the number of significant components in CCA based on Bonferroni correction. Then, the same mixed-effects model was repeated by replacing the prior to drinking visits of the heavy drinkers with visits after drinking initiation. For each component showing a significant group difference, 3 exploratory analyses were performed. First, we added a sex-by-group or site-by-group interaction term to the mixed-effects model to test whether the group difference differed by sex or data collection site. Second, to test whether heavy drinking had a dose-response association with brain-behavior scores, the mixed-effects model was applied to heavy drinkers by replacing the drinking group variable with a continuous alcohol consumption variable (average number of drinks in the past month or number of binge drinking episodes in the past month). Third, we applied the mixed-effects model to test the alcohol effect on the trajectory of each individual neuropsychological measure with significant canonical loading. The statistical analysis was conducted between October 11, 2024, and July 3, 2025, using MATLAB, version R2022b (The MathWorks, Inc).

## Results

A total of 3076 visits from 633 participants (mean [SD] age at baseline, 15.7 [2.6] years; 318 female [50.2%] and 315 male [49.8%]; 60 self-reporting as Asian [9.5%], 80 as Black [12.6%], 466 as White [73.6%], and 37 as other [5.8%] race) were included in the analysis (Table). A total of 238 participants (37.6%) initiated heavy drinking before age 26 years. The remaining 395 participants (62.4%) who did not initiate heavy drinking were considered non-heavy drinkers.

### Socioemotional System

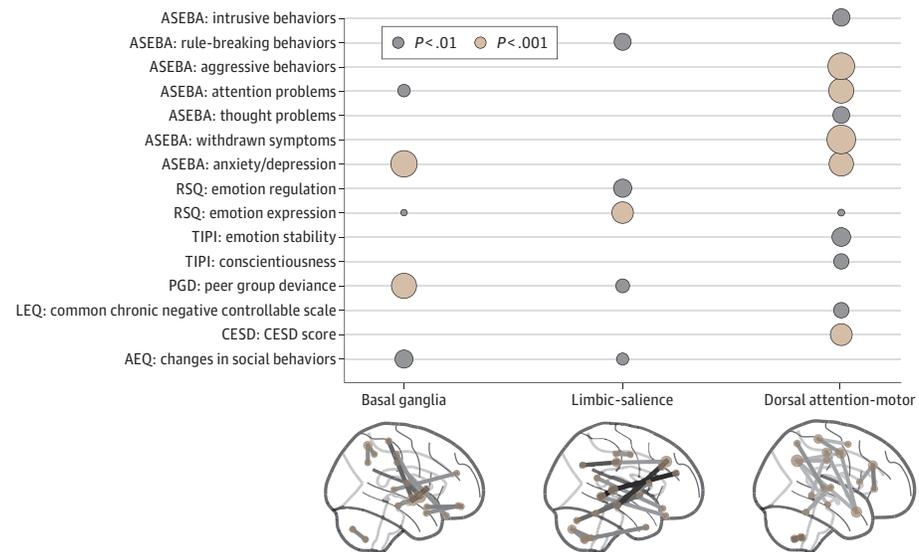
The CCA specific to the socioemotional system yielded 3 significant components ( $P < .001$  by permutation test) (eFigure 3A in Supplement 1). These components were consistently identified based on different data processing methods (eFigures 4 and 5 in Supplement 1). Each component encoded a pattern of brain-behavior coupling between a functional brain network and a constellation of neuropsychological measures (Figure 1). Based on functional connectivities and neuropsychological measures with high canonical loadings (Figure 1; eTable 3 in Supplement 1), the first socioemotional component showed an association between functional connectivity in basal ganglia and symptoms of anxiety and depression and peer group deviance score. The second

Table. Demographics of NCANDA Participants During Follow-Up

Characteristic	Participants, No. (%)	
	Non-heavy drinkers (n = 395)	Heavy drinkers (n = 238)
Sex		
Female	217 (54.9)	101 (42.4)
Male	178 (45.1)	137 (57.6)
Age at baseline, mean (SD), y	15.7 (2.6)	16.0 (2.0)
No. of visits, mean (SD)	4.4 (1.7)	5.3 (1.6)
Family drinking history <sup>a</sup>	35 (12.8)	24 (12.4)
Race		
Asian	34 (8.6)	16 (6.7)
Black	64 (16.2)	16 (6.7)
White	273 (69.1)	193 (81.1)
Other <sup>b</sup>	24 (6.1)	13 (5.5)
Site		
University of Pittsburgh	62 (15.7)	25 (10.5)
SRI International	82 (20.7)	36 (15.1)
Duke	88 (22.3)	47 (19.7)
Oregon Health & Science University	71 (18.0)	57 (24.0)
University of California, San Diego	92 (23.3)	73 (30.7)
Socioeconomic status, mean (SD) <sup>c</sup>	16.6 (2.5)	17.1 (2.4)

<sup>a</sup> Number of participants who had (1) at least 1 biological parent with substantial problems indicative of an alcohol disorder; (2) 2 or more biological grandparents with substantial problems indicative of an alcohol disorder; or (3) 1 or more biological grandparents and 2 or more other biological second-degree relatives (eg, aunt, uncle) with substantial problems indicative of an alcohol or other drug disorder.  
<sup>b</sup> Other race included American Indian, Pacific Islander, and multiracial.  
<sup>c</sup> Defined as maximum years of education of either parent.

Figure 1. Significant Brain-Behavior Components in the Socioemotional Domain and Top 1% Functional Connectivity With Highest Canonical Loadings



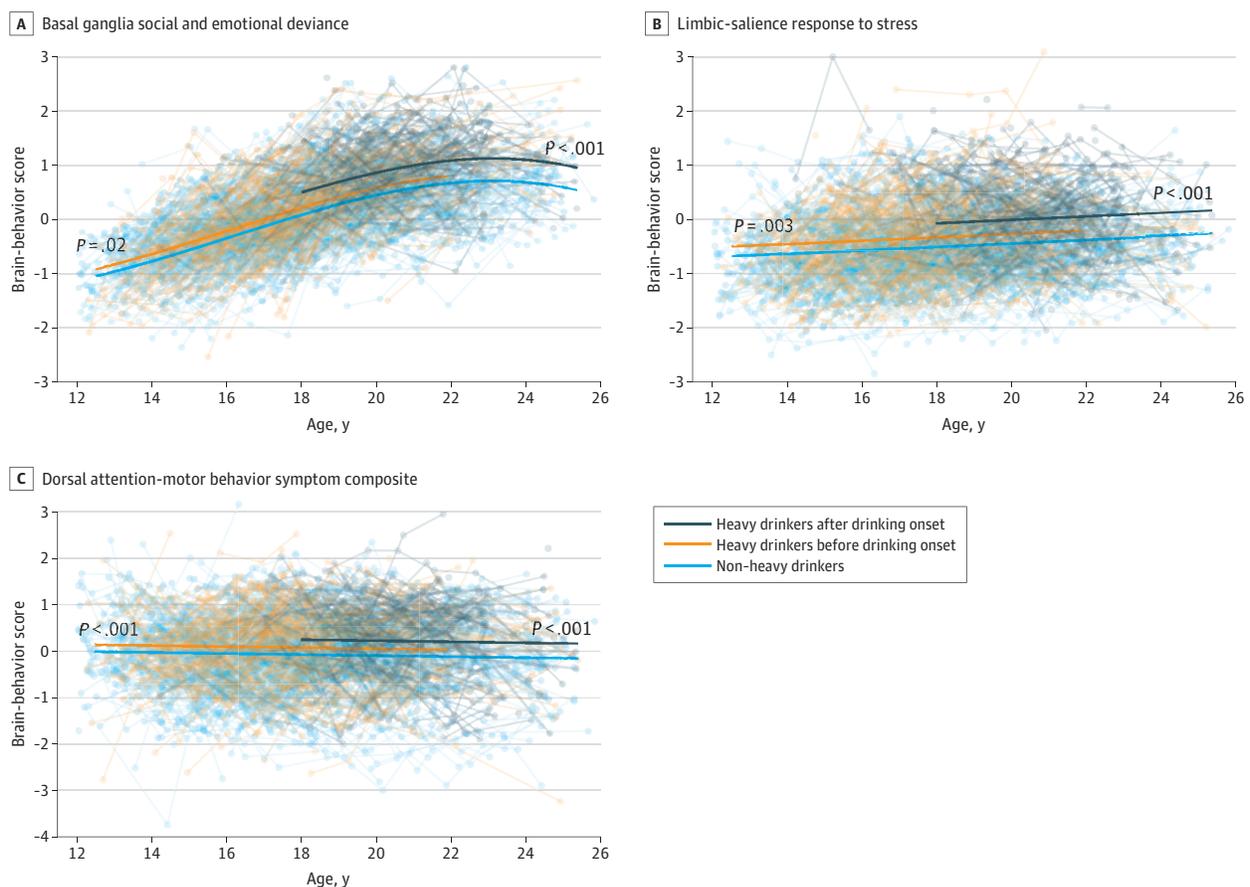
Each column displays the corresponding neuropsychological measures with significant loadings. Measures with nonsignificant loadings for all 3 components were omitted. AEQ indicates Alcohol Expectancy Questionnaire; ASEBA, Achenbach System of Empirically Based Assessment; CESD, The Center for Epidemiologic Studies Depression; LEQ, Life Event Questionnaire; PGD, peer group deviance; RSQ, Response to Stress Questionnaire; TIPI, Ten Item Personality Measure.

component showed an association between functional connectivity in limbic regions and the salience network and subscales of the Response to Stress Questionnaire.<sup>41</sup> The third component showed an association between functional connectivity in dorsal attention and motor networks and subscales of the Childhood Behavior Checklist defined by the Achenbach System of Empirically Based Assessment.<sup>42</sup>

Compared with non-heavy drinkers, the first brain-behavior score (associated with basal ganglia and anxious and depressive symptoms) showed a significant elevation among heavy drinkers prior to drinking onset (score increase, 0.13; 95% CI, 0.02-0.22;  $P = .02$ ) (Figure 2A). The second brain-behavior score (associated with the limbic-salience network and Response to Stress Questionnaire subscales) also showed a significant elevation (score increase, 0.18; 95% CI, 0.08-0.28;  $P = .003$ ) (Figure 2B). After heavy drinking onset, all 3 brain-behavior scores showed a significant association with heavy drinking (Figure 2A-C). The largest elevation was observed in the second brain-behavior score (score increase, 0.44; 95% CI, 0.33-0.54;  $P < .001$ ). Critically, in all 3 components, the effect size of the group difference was significantly larger after drinking onset than before onset, with the highest increase observed in the first component ( $z = 3.51$ ;  $P < .001$ ). The association between brain-behavior scores and heavy drinking in the 3 components endured after testing differences in trajectories of functional canonical variables alone or neuropsychological canonical variables alone (eTable 4 in Supplement 1).

The significant associations persisted based on different ways of grouping participants into drinkers vs nondrinkers (eTable 5 in Supplement 1); were not biased by the greater percentage of male and White participants in the heavy drinking group (eFigure 6 in Supplement 1); and endured

Figure 2. Developmental Trajectories of Brain-Behavior Scores From 3 Socioemotional Components



The highest order of the polynomial of age was determined by stepwise regression.  $P$  values correspond to significant developmental deviation with respect to the non-heavy drinkers.

adjustment for data collection site, race, and socioeconomic status (eTable 6 in Supplement 1).

Exploratory analyses revealed that brain-behavior scores did not show a dose response after drinking onset, and there was no significant alcohol-site interaction both before and after drinking onset. However, a significant alcohol-sex interaction emerged, with female heavy drinkers showing a significantly larger elevation in the first brain-behavior score compared with male heavy drinkers ( $P = .014$ , Bonferroni corrected) (eFigure 7 in Supplement 1) prior to drinking onset. This interaction was no longer significant after drinking onset.

Finally, we analyzed which individual neuropsychological measures of the socioemotional system drove the group differences in brain-behavior scores. Testing the association between heavy drinking and measures with significant loadings (Figure 1; eFigure 8 in Supplement 1) revealed that heavy drinking was associated with higher peer group deviance, more changes in social behavior, more rule-breaking behaviors, more intrusive behaviors, and lower withdrawn depression symptom scores (eFigure 9 in Supplement 1). Other than intrusive behavior, the scores showed significant associations both before and after drinking onset, with the effect sizes enlarged following drinking onset ( $t_{2516} > 4.1$ ;  $P < .001$ ).

### Executive Control System

The CCA specific to the executive control system yielded 4 significant components ( $P < .001$  by permutation test) (eFigures 4 and 5 in Supplement 1). To ensure that this result was not biased by the larger number of executive control measures (82 vs 44 in the socioemotional system), we reduced the 82 executive control measures to 44 measures by principal component analysis or random sampling and repeated the CCA, which confirmed that all 4 significant components were preserved (eFigure 10 in Supplement 1). Based on the top functional connectivities with the highest canonical loadings (Figure 3; eTable 3 in Supplement 1), the first and the third components captured the functional integration between the sensorimotor network and the cerebellum and visual networks, whereas the second and fourth components captured widespread functional connectivity in the frontal and temporal regions. On examination of significant neuropsychological measures driving the brain-behavior correlation (Figure 3; eFigure 8 in Supplement 1), the first component summarized a wide variety of test scores in the computerized test battery,<sup>31</sup> the second component was driven by sensation-seeking behavior,<sup>34</sup> the third component was driven by visual object learning performance, and the fourth component was driven by Rey-Osterrieth complex figure<sup>43</sup> and face memory test performance.

Testing the group difference in the trajectories of brain-behavior scores revealed that compared with non-heavy drinkers, none of the 4 identified components showed significant developmental deviation among the drinkers before they initiated heavy drinking (Figure 4A-D). After drinking onset, however, a significant drinking association emerged for the second brain-behavior score of the executive control system (linked to the fronto-temporo-visual network and sensation-seeking behavior) (score increase, 0.21; 95% CI, 0.11-0.32;  $P < .001$ ) (Figure 4B), in which heavy drinkers showed elevated scores during their heavy drinking visits. This group difference did not significantly interact with sex or site but showed dose responses: After adjusting for age, brain-behavior scores of the heavy drinkers significantly correlated with the average number of drinks ( $t_{472} = 2.7$ ;  $P = .048$ ) and number of binges in the past month ( $t_{344} = 2.8$ ;  $P = .04$ ) (eFigure 11 in Supplement 1).

Finally, applying the same mixed-effects model to regress each of the significant neuropsychological measures of that component revealed that heavy drinkers had significantly higher sensation-seeking scores than non-heavy drinkers ( $P < .001$ ) (eFigure 9F in Supplement 1) both before and after drinking onset and poorer emotion control (score increase, 0.18; 95% CI, 0.04-0.32;  $P$  for trend = .01) (eFigure 9G in Supplement 1) only after drinking onset.

### Discussion

In this cohort study based on 9-year longitudinal neuroimaging, neuropsychological, and behavioral data of the NCANDA cohort, our data-driven approach yielded neurobiological evidence supporting

the association of dual-system asynchronous development with adolescent drinking problems. Heightened reactivity in the socioemotional system during adolescence was found to hijack the regulatory ability of the executive control system, heightening the risk of initiating heavy alcohol consumption. Heavy drinking, in turn, was associated with deficits in the executive control system and exacerbated alteration in the socioemotional system during late adolescence and young adulthood. This pattern may represent a precursor of developing addiction, another cyclical process

Figure 3. Significant Brain-Behavior Components in the Executive Control Domain and Top 1% Functional Connectivity With Highest Canonical Loadings



Each column displays the corresponding neuropsychological measures with significant loadings. Measures with nonsignificant loadings for all 4 components were omitted. BRIEF indicates Behavior Rating Inventory of Executive Function; CNP, Penn Computerized Neurobehavioral Battery; GPEG, grooved pegboard; PASAT, Paced

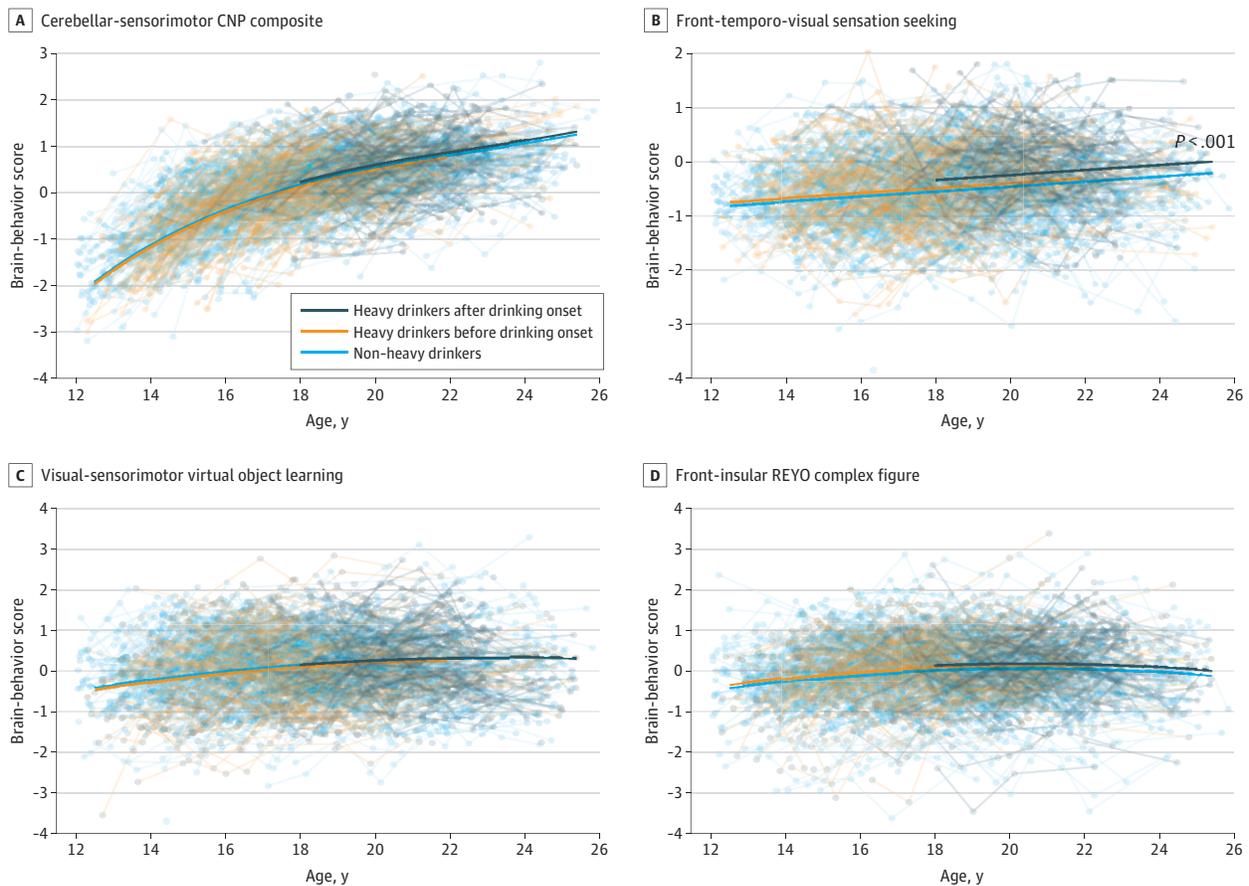
Auditory Serial Addition Test; REYO, Rey-Osterrieth Complex Figure Test; UPPS-P, Urgency-Premeditation-Perseverance-Sensation Seeking-Positive Urgency Impulsive Behavior Scale.

(ie, the Koob model)<sup>44,45</sup> driven by neurobiological changes in similar functional systems responsible for reward processing, emotion regulation, and executive control.

The maturational imbalance model is a testable hypothesis that has arisen from developmental cognitive science.<sup>1,2,5</sup> Several variants of the model exist that diverge in their specific formulations on the shape of developmental trajectory and timing of the maturational peak of brain systems.<sup>46-48</sup> To characterize trajectories, research often has relied on only a few neuropsychological measures, which might not capture the full complexity of a brain-behavior system. Our data-driven analysis enriched the characterization by coupling whole-brain functional connectivity with a comprehensive array of neuropsychological measures. Results showed that the socioemotional component (Figure 2A) was characterized by an inverted U shape that plateaued after age 20 years and declined thereafter, whereas the executive control component (Figure 4A) continued to develop into young adulthood. Such distinct developmental patterns between the 2 systems raise the possibility of maturational imbalance.

A major debate poses whether it is possible to measure maturational imbalance objectively.<sup>49</sup> Indeed, the dual systems were segregated in their functional networks and primary roles,<sup>49</sup> and neuropsychological measures associated with the 2 systems were often not commensurate. Our analysis detected imbalance by comparing the developmental trajectory of each system with its respective norm. As shown herein, the onset of heavy drinking was associated with elevated brain-behavior strength only in the socioemotional system. Such higher socioemotional reactivity

Figure 4. Developmental Trajectories of Brain-Behavior Scores From 4 Executive Control Components



The highest order of the polynomial of age was determined by stepwise regression. *P* values correspond to significant developmental deviation with respect to the non-heavy drinkers. CNP indicates Penn Computerized Neurobehavioral Battery; REYO, Rey-Osterrieth Complex Figure Test.

outpaced the regulatory capacity of the executive control system, which followed a typical, slower developmental trajectory without significant deviation.

Regarding the predrinking deviation in the socioemotional system, widespread associations with functional networks and neuropsychological measures were observed in 3 neural systems (Figure 1). In particular, the dorsal attention network is crucial for voluntary attention control,<sup>4,50</sup> supporting emotional self-regulation, social attention, and response and resilience to stressors.<sup>51-55</sup> On the other hand, the basal ganglia and limbic systems are critical dopamine pathways involving the nucleus accumbens<sup>56,57</sup> and play a role in processing rewards, emotions, and social cues.<sup>58-61</sup> These functional associations underpinned impaired neuropsychological behaviors in social functioning and mental health problems<sup>62</sup> (Figure 2A-C; eFigure 9A-E in [Supplement 1](#)), increasing the risk of heavy drinking onset, especially in female participants. This finding was in line with a previous analysis of NCANDA participants showing that 27 mental health measures during high school better predicted heavy drinking onset during college among young women compared with young men.<sup>19</sup>

In contrast to the socioemotional system, the association between heavy drinking and brain-behavior scores in the executive control system was primarily observed after drinking onset (Figure 4B). The associated component involved a constellation of neuropsychological measures led by sensation seeking and functional connectivity in regions commonly identified in default mode and sensorimotor networks. Notably, these findings replicated and extended a prior analysis of the first 3 years of longitudinal NCANDA data,<sup>17</sup> which also showed that heavy drinkers exhibited greater within-network connectivity in the sensorimotor network and 2 other motor networks with more alcohol consumption, a dose response that persisted in the current analysis. Concurrently, the impaired executive control system was less able to regulate the exacerbated stress and negative emotional states.

### Limitations

This study had some limitations. First, our categorization of neuropsychological measures into 2 systems was subjective and required fine-tuning. Many measures cut across both systems; for example, the emotion recognition test in WebCNB,<sup>31,63</sup> delay discounting, and impulsivity are indicators of both cognitive control and socioemotion.<sup>49</sup> Additionally, although we showed that principal component analysis and bilateral averaging of functional connectivity could improve canonical correlation between brain and behavior measures, this practice might obscure findings on unilateral functional networks associated with lateralized executive tasks. Finally, our analysis did not consider the influence of genetic variation and fine-grained environmental factors, such as urbanicity and pollution, on maturational imbalance.

### Conclusions

In this cohort study of NCANDA participants, we used a data-driven approach to analyze 9-year longitudinal neuroimaging, neuropsychological, and behavioral data to find direct biological evidence to support the role of dual-system asynchrony in adolescent drinking behaviors. The findings suggest that heightened socioemotional reactivity occurring during adolescence may overwhelm executive control, increasing the risk of heavy alcohol use, which, in turn, may pose risks for developing alcohol-induced damage of both systems. This cycle could enable and reinforce persistent alcohol use, thereby increasing vulnerability to developing alcohol use disorder.

### ARTICLE INFORMATION

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**Author Contributions:** Dr Zhao had full access to all of the data in the study and takes responsibility for the integrity of the data and the accuracy of the data analysis.

*Concept and design:* Zhao, Milecki, Pohl.

*Acquisition, analysis, or interpretation of data:* All authors.

*Drafting of the manuscript:* Zhao, Sullivan, Pohl.

*Critical review of the manuscript for important intellectual content:* All authors.

*Statistical analysis:* Zhao, Milecki, Pohl.

*Obtained funding:* Zhao, Pfefferbaum, Sullivan, Pohl.

*Administrative, technical, or material support:* Zhao, Brumback, Pfefferbaum, Pohl.

*Supervision:* Kuceyeski, Sullivan, Pohl.

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**Data Sharing Statement:** See [Supplement 2](#).

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#### SUPPLEMENT 1.

##### eMethods.

**eFigure 1.** Schematic Overview of Analysis Workflow

**eFigure 2.** Distribution of Age of All Participant Visits

**eTable 1.** Neuropsychological Measures of Socioemotional System

**eTable 2.** Neuropsychological Measures of the Executive Control System

**eFigure 3.** Two CCA Models Used to Find Correlated Patterns Between Resting-State Functional Connectivity and Neuropsychological Measures

**eFigure 4.** Distribution of Regression Slope of Linear Mixed-Effects Models

**eFigure 5.** Distribution of Canonical Correlation Values Across Processing Methods of Resting-State Functional Data

**eTable 3.** Top 5 Functional Connectivities (Between Pairs of Regions) With Highest Canonical Loadings for Each Significant Canonical Component in eFigure 3.

**eTable 4.** P Values of the Group Differences Between Heavy Drinkers and Non-heavy Drinkers Based on Different Regression Targets

**eTable 5.** The Same LME Models in eTable 4 Applied to Different Groupings of Participants Into Drinkers (Cahalan Score > 0) and Nondrinkers (Cahalan Score = 0)

**eTable 6.** Results of LME With Additional Covariates of Race, Site, and Socioeconomic Status

**eFigure 6.** Distribution of P Values (Uncorrected) Associated With the Alcohol Effects When Running Mixed-Effects Models on Randomly Sampled Cohorts

**eFigure 7.** Significant Sex-Alcohol Interaction in the First Socioemotional Component

**eFigure 8.** Null Distribution of Canonical Loadings for Individual Neuropsychological Measures of the Socioemotional and Executive Control Systems

**eFigure 9.** Distribution of Top Neuropsychological Measures Defining the Canonical Components That Showed Significant Alcohol Effects

**eFigure 10.** Distributions of Canonical Correlations

**eFigure 11.** Brain-Behavior Score of the Second Component of the Executive Control System

##### eReferences.

#### SUPPLEMENT 2.

##### Data Sharing Statement