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Title: Forecasting future prevalence and gender differences in binge drinking among young adults through 2040

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36 **Abstract**

37

38 Background

39 Binge drinking among adolescents and young adults has changed over time, but patterns differ by age and
40 gender. Identifying high-risk groups to target future reductions remains a public health priority.

41 Forecasting methods can provide a better understanding of variation and determinants of future binge
42 drinking prevalence.

43

44 Methods

45 We implemented regression-based forecasting models to estimate the prevalence and gender differences
46 in binge drinking among cohort groups of U.S. young adults, ages 18, 23-24, and 29-30 through 2040.

47 Forecasting models were adjusted for covariates accounting for changes in demographic, Big-5 social
48 roles (e.g., residential independence), and drinking norms and related substance use, to understand the
49 drivers of forecasted binge drinking estimates.

50

51 Results

52 From the last observed cohort group (years varied by age) through 2040, unadjusted binge drinking
53 prevalence was forecasted to decrease from 26% (95% CI: 20, 33%) (2011-15) to 11% (95% CI: 4, 27%)
54 at age 18, decrease from 38% (95% CI: 30, 45%) (2006-2010) to 34% (95% CI: 18, 55%) at ages 23/24,
55 and increase from 32% (95% CI: 25, 40%) (2001-2005) to 35% (95% CI: 16, 59%) at ages 29/30.

56 Gender-stratified forecasts show a continuation in the narrowing of binge drinking prevalence between
57 young men and women, though the magnitude of narrowing differs by age. Estimated trends were
58 partially explained by changing drinking norms and other substance use, though these indirect effects
59 explained less of the total trend as age increased.

60

61 Conclusions

62 Understanding how covariates influence binge drinking trends can guide public health policies to
63 effectively leverage the most important determinants of future binge drinking and reduce the harm caused
64 by binge drinking from adolescence to adulthood.

65

66 **Key words:** binge drinking, gender differences, adolescents, young adults, forecasting

67

68 Introduction

69

70 Binge drinking is the consumption of a large amount of alcohol over a short period, typically defined as
71 five or more drinks in a row, and is implicated in more than half of alcohol-attributable deaths in the US
72 (Naimi et al., 2003), through both acute and chronic health conditions (Chikritzhs et al., 2001). Reducing
73 the prevalence of binge drinking is a significant public health goal, and efforts to do so must be guided by
74 an understanding of the population-level variation and determinants of binge drinking over time.

75

76 There is evidence of substantial variation in binge drinking prevalence over time, though patterns differ
77 by age. Among high school seniors, binge drinking has decreased across cohorts from 1976-2019 (Miech
78 et al., 2020; Schulenberg et al., 2020). Among young adults, binge drinking trends have been uneven
79 across cohorts. Decreasing trends have been consistent among 19-20 year olds especially since 2005,
80 while among those age 21-30, binge drinking increased through about 2006-2010, then leveled off in
81 more recent years (Schulenberg et al., 2020). Gender differences in alcohol use have also been
82 diminishing. Adolescent girls are now as likely as boys to initiate alcohol consumption and binge drink
83 (Cheng et al., 2016; Cheng and Anthony, 2017; Miech et al., 2020), and increases in binge drinking
84 prevalence have been greater in young adult women than men (Cheng and Anthony, 2017; Patrick et al.,
85 2019; Slade et al., 2016; White et al., 2015). Research to identify causes of these changes is limited, but
86 trends appear to be related to changing disapproval of alcohol (Keyes et al., 2012), as well as gender-
87 specific changes in traditional gender roles (Seedat et al., 2009), attitudes towards drinking (Kuntsche et
88 al., 2011) and the social contexts of drinking (Holmila and Raitasalo, 2005).

89

90 Together, trends suggest binge drinking should remain an important focus for future public health
91 priorities. To galvanize support and optimize the resources needed to meet this priority, future levels of
92 binge drinking can be estimated using forecasting models, which predict dynamic changes in health
93 outcomes under prespecified conditions (detailed summaries of forecasting methods can be found here
94 (Soyiri and Reidpath, 2012)). Forecasting models have been used to understand the future burden of other
95 health conditions under current policies and conditions, such as infectious disease (Choi et al., 2016;
96 Chretien et al., 2014), cancer (Bray and Møller, 2006), injuries (Ladrón de Guevara et al., 2004;
97 O'Connor, 2005), and obesity (Robinson et al., 2013); however, they are underutilized in substance use
98 research. One recent study forecasted the prevalence of alcohol-related hospital admissions, but estimates
99 were limited to 2021 (de Vocht et al., 2017).

100

101 Also, most forecasting applications focus solely on variation by observed age, period, and cohort patterns,
102 without consideration of other known determinants of observed rates. Historical trends in binge drinking
103 are influenced by numerous factors, including parental socio-economic factors in adolescence (Lemstra et
104 al., 2008; Patrick et al., 2012), alcohol norms and friends' alcohol use (Keyes et al., 2012), use of
105 cigarettes and marijuana (Bobo and Husten, 2000; Midanik et al., 2007; Weitzman and Chen, 2005), and
106 the fulfilment of young adult social roles in the transition to adulthood (Jager et al., 2015). These 'Big 5'
107 social roles (i.e., attending college, finding employment, residential independence, getting married, and
108 having children) reflect the historical context of labor force and social structures and the normative
109 expectations faced during this period of life (Settersten Jr, 2007), in ways that are associated with binge
110 drinking (Bachman et al., 1997). Finally, binge drinking prevalence may vary according to the
111 demographic composition of the population. Incorporating information from these covariates, including
112 how they have changed over time, serves two key purposes. First, it informs more accurate forecasting
113 models, overall and for key population groups, and second, it highlights important modifiable targets to
114 reduce future binge drinking levels. A key purpose of forecasting is to guide effective public health
115 policies and understanding how covariates influence binge drinking trends can be used to target those
116 policies to effectively address the most important determinants of future binge drinking.

117
118 The current study developed forecasting models to estimate binge drinking prevalence and gender
119 differences in cohorts of young adults from 2016 through 2040, and to understand the role of social and
120 demographic determinants of binge drinking in forecasted estimates.

121 122 Materials and Methods

123 124 Sample

125
126 The Monitoring the Future (MTF) study includes nationally representative samples of approximately
127 15,000 high school seniors (12th grade) surveyed annually since 1976 (Miech et al., 2020). From the
128 annual survey, 2450 students are randomly selected for longitudinal follow-up, with oversampling for
129 students who report drug use (Schulenberg et al., 2020). Those selected begin follow-up assessments
130 either one (modal age 19) or two (modal age 20) years later, and are followed biennially thereafter
131 through modal age 29/30 (Schulenberg et al., 2020). An Institutional Review Board of University of
132 Michigan approved the study.

133

134 Respondents were grouped by cohort and age, in order to estimate prevalence across cohorts stratified by
135 age. Cohorts were defined based on the year that respondents were seniors in high school and grouped in
136 5-year intervals. Observed cohort groups ranged from 1976-1980 to 2011-2015; forecasted cohorts
137 continued to 2036-2040. Age was defined over the study period as the modal age(s) of respondents at:
138 baseline (age 18), third follow-up (ages 23-24), and sixth follow-up (ages 29-30). Ages were selected to
139 broadly represent the beginning, middle, and end of the transition to adulthood (Waters et al., 2019).

140
141 Because of the longitudinal study design, the most recent observed cohort group (and the first forecasted
142 group) differed by age. For ages 23-24, the first cohort group was surveyed in 1981 and the most recent
143 observed cohort group was surveyed in 2006-2010, so forecasts begin with the 2011-2015 cohort group.
144 At ages 29-30, the first cohort group was interviewed in 1986-1990 and the most recent observed cohort
145 group was 2001-2005, and forecasts begin with the 2006-2010 cohort group (see Figure 1). The observed
146 analytic sample sizes across all groups comprised 97,812 respondents at age 18, 85,559 respondents at
147 age 23-24, and 73,298 respondents at age 29-30.

148 149 Variables

150
151 Binge drinking was defined at each wave as any versus none, based on their response to the question,
152 “How many times have you had five or more drinks in a row over the past two weeks?”.

153
154 To better understand determinants of binge drinking trends, we compared observed binge drinking trends
155 without vs. with adjustment for three sets of covariates. Covariates were selected based on a priori
156 associations with binge drinking and evidence of variation over time, and included: 1) baseline socio-
157 demographics, including: sex (male, female), high school GPA (9=A (93-100) 8=A- (90-92) 7=B+ (87-
158 89) 6=B (83-86) 5=B- (80-82) 4=C+ (77-79) 3=C (73-76) 2=C- (70-72) 1=D (69 or below)), father’s and
159 mother’s highest reported education (<HS degree, HS degree with/without some college, college degree
160 or more), race/ethnicity (Non-Hispanic Black, Non-Hispanic White, Hispanic, Other (including multiple
161 races)); 2) (binary) young adult Big-5 social roles (attend two-/ four-year college full-time, residential
162 independence, have children, married, work full-time); and 3) drinking norms and substance use
163 (disapprove of having 5 or more drinks on the weekend (1: Don't Disapprove – 3: Strongly Disapprove),
164 how many of your friends drink (1: None – 5: All), perceived risk of 5 or more drinks on the weekend (1:
165 No Risk – 4: Great Risk), use of marijuana and tobacco (past-year marijuana use (yes/no), past-year
166 cigarette use (yes/no)). Covariates were lagged by one year to establish temporality.

167

168 Baseline socio-demographics were recorded at age 18. Among big-5 social roles, college attendance,
169 residential independence, and working full-time were only included among the 23-24 and 29-30 age
170 groups. For all other variables (i.e., have children, married, drinking attitudes and marijuana/cigarette
171 use), responses varied at each age.

172
173 Attrition and missing data

174
175 Three variables were missing more than 10% of possible responses at age 18: marijuana use (17%),
176 perceived risk of binge drinking (41%), and binge drinking disapproval (52%) (see Supplementary Table
177 1). There were two main sources of missing data, study attrition and planned missingness.

178
179 To account for attrition, all models included attrition weights, calculated as the inverse of the probability
180 of participation at each age group (i.e., 23-24, 29-30), based on the following baseline characteristics:
181 gender, race/ethnicity, college plans, truancy, high school grades, number of parents in the home,
182 religiosity, parental education, alcohol use, cigarette use, marijuana use, other illicit drug use, region,
183 cohort, and sampling weight (correcting for over-sampling of age 18 substance users).

184
185 Planned missingness arose due to the MTF study design. To reduce the survey participation burden,
186 certain survey questions are only administered to one of six randomly assigned subsamples (i.e., forms),
187 in addition to a core set of questions. This planned missingness study design feature resulted in some data
188 that were missing completely at random. To maximize the study sample size, data were multiply imputed
189 across forms. Where data are assumed to be missing completely at random, this approach has been shown
190 to be a valid method to reduce Type II error rates (Little and Rhemtulla, 2013; Noble and Nakagawa,
191 2018; Rhemtulla and Little, 2012; Wood et al., 2019), even when up to 90% of data are missing (Madley-
192 Dowd et al., 2019). Twenty models were imputed using chained equations, based on all observed
193 exposure, covariate, and outcome data, and combined with corrected standard errors (Rubin, 2004).
194 Covariate distributions did not vary between unimputed and imputed datasets (see Supplementary Table
195 1).

196
197 Analysis

198 We utilized a linear regression-based approach to build forecasting models in a series of eight steps. All
199 steps were completed separately for each age group, and ages 23-24 and 29-30 included age 18 values as
200 additional covariates to utilize the longitudinal data. We describe each step as applied to one covariate
201 (GPA) for clarity.

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Model fitting and validation

Prior to building the forecasting model, we identified the best fitting model as a combination of the a priori specified covariates. Model fit was assessed using likelihood ratio tests of nested models, sequentially adding demographic, big 5, and alcohol norms/other substance use covariates. To examine the validity of the forecasting model, we estimated the accuracy of the model in predicting observed binge drinking prevalence. To do this we removed the observed binge drinking data for the three most recently observed cohort groups (e.g., 2001-2005, 2006-2010, and 2011-2015 for age 18), then forecasted binge drinking prevalence using multiple imputation based on the best fitting regression model. We compared the predicted vs. observed binge drinking values for these cohorts. The results are shown in Supplementary Table 2. The best-fitting prediction model included all covariates, for which the Area Under the Curve (AUC) ranged from 0.79-0.80 for each age group, indicating good accuracy.

Build a forecasting model with observed covariates

First, we visually assessed variation across cohort group in each covariate to determine the functional form of change (e.g., no change, linear increase/decrease, non-linear). Covariates were standardized based on deviation of the within-cohort mean from the total sample mean (i.e., grand mean). While the values of these variables are not interpretable, they facilitate the visualization of trends over time and comparison between variables. Subsequent steps utilized unstandardized variables, so that model estimates would be interpretable. With the specified functional form, we estimated the magnitude of change over time by regressing each covariate on cohort. For example, the unstandardized cohort mean GPA increased from 5.58 to 6.52 (on a scale from 1-9), across observed cohort groups. The linear regression estimate was 0.15 (SE=0.003).

Second, the cohort-level covariate means were extrapolated based on the form of change across cohort group (see step 1) and the previous group mean, starting with the baseline cohort (2001-05, 2006-10, 2011-15, depending on the age group). For example, the average GPA has been linearly increasing by 0.15 points per cohort group, and the baseline (2011-15) mean was 6.52, so the 2016-20 mean was $6.52 + 0.15 = 6.67$. If a variable did not meaningfully change across cohort, the baseline value was carried forward. For binary variables, the covariate means were assigned on the logit scale.

235 Third, using these extrapolated means, we simulated each individual's covariate values in the forecasted
236 cohorts. Each forecasted cohort group included 12,200 individuals (i.e., the average size of observed
237 cohort groups). Individual covariate values were simulated from a distribution with the cohort mean (see
238 step 2), and the standard deviation of the baseline cohort group. For example, the 2016-20 cohort GPA
239 was simulated from $X \sim N(6.67, 1.93)$. Skew and kurtosis measures indicated that continuous variables
240 were approximately normally distributed and were thus simulated from a normal distribution; binary
241 variables were simulated from a binomial distribution.

242

243 Fourth, a column for binge drinking status was added to the simulated dataset, with all values set to
244 missing, and merged the simulated and observed datasets.

245

246 *Multiply impute binge drinking in forecasted cohorts*

247

248 Fifth, we pooled the multiply imputed datasets with corrected standard errors (Rubin, 2004) to estimate
249 binge drinking prevalence in forecasted cohort groups, converting the log odds to prevalence. We imputed
250 20 datasets using chained equations, combined with corrected standard errors, averaging coefficient
251 vectors, variance-covariance matrices, and adding a non-negative correction to variance-covariance
252 matrices inversely proportional to the predictive ability of the imputation models, effectively widening
253 confidence intervals where missing data values are poorly predicted by observed data (Pigott, 2009). To
254 reflect the uncertainty around the forecasted point estimates, the model residuals were adjusted under the
255 assumption of uncorrelated residuals, using the formula $\hat{\sigma}_h = \hat{\sigma}\sqrt{h}$, where $\hat{\sigma}_h$ is the standard deviation of
256 the h -step forecast distribution, and $\hat{\sigma}$ is the residual standard deviation (Hyndman and Athanasopoulos,
257 2018).

258

259 Sixth, we added sequential covariate sets to estimate binge drinking trends accounting for concurrent
260 patterns in: a) demographics; b) (a and) Big-5 social roles; and c) (a, b, and) drinking norms/substance
261 use. This approach was to understand what might explain variation in binge drinking trends and was
262 analogous to a decomposition approach to estimate distinct mediation pathways, rather than a
263 confounding elimination strategy. The unadjusted estimates refer to the total cohort group trends (i.e., the
264 effect of cohort on binge drinking through all pathways), whereas the covariate-adjusted estimates refer to
265 the effect of cohort trends in binge drinking, not due to the model covariates (i.e., the controlled direct
266 effect). We also calculated the relative difference in binge drinking prevalence between unadjusted and
267 adjusted prevalence estimates, to quantify the effect of these covariates on binge drinking trends. For
268 example, if unadjusted binge drinking forecasted prevalence estimates are greater than those adjusted for

269 big-5 social roles, this would suggest that those covariates are important determinants of future binge
270 drinking, and the percent change would represent the proportion of the estimates that were due to big-5
271 social role patterns. To reflect this interpretation, we subsequently refer to unadjusted and adjusted model
272 estimates as total-effect and direct-effect estimates, respectively.

273
274 Seventh, we repeated step 7 in models stratified by gender to estimate gender differences in forecasted
275 binge drinking.

276
277 All analyses were implemented in R (version 4.0.2), and multiple imputation was implemented with the
278 ‘MICE’ package (Buuren and Groothuis-Oudshoorn, 2010). Syntax to implement these steps can be
279 found in the Supplementary materials.

280
281 Results

282
283 Trends in social determinants of binge drinking across cohorts

284
285 Figure 2 and Supplementary Tables 3-5 present trends across cohorts in covariates used to forecast binge
286 drinking for age 18, 23-24, and 29-30 groups. The cohort trends were linear for all covariates, except non-
287 Hispanic Black prevalence, which did not change over the study period. Cohort trends were generally
288 similar for all ages, except perceived risk and disapproval of binge drinking, which increased at ages 18
289 and 23-24 and decreased at ages 29-30.

290
291 Forecasted binge drinking prevalence trends

292
293 Total- and direct-effect binge drinking prevalence trends across cohort are presented graphically in Figure
294 3 and estimates are provided in Supplementary Table 6. For parsimony, we focus on contrasts in binge
295 drinking estimates between models with no covariates versus those with all covariates. Differences
296 between these two models were most appreciable, and the latter model had the best fit to the data.
297 Sequentially-adjusted model estimates are presented in Supplementary Figures 1-3.

298
299 Among age 18 respondents, total-effect (i.e., unadjusted) binge drinking prevalence decreased from 48%
300 (95% CI: 42-55%) in the 1976-1980 cohort to 11% (95% CI: 4-27%) in the 2036-2040 cohort group. In
301 the direct-effect (i.e., fully adjusted) models, age 18 decreases in binge drinking prevalence were much
302 smaller, decreasing to 36% (95% CI: 14-65%) in the 2036-2040 cohort group. Among age 23-24

303 respondents, total-effect observed binge drinking prevalence decreased from 41% (95% CI: 34-49%) in
304 the 1981-1985 cohort group to 34% (95% CI: 18-55%) in the 2036-2040 cohort group. In the direct-effect
305 models, observed and forecasted binge drinking prevalence estimates ranged from 41% (95% CI: 33-
306 51%) to 45% (95% CI: 36-54%) across cohort groups, with no clear pattern of change over time. Among
307 age 29-30 respondents, total observed binge drinking prevalence was 29% (95% CI: 23-36%) in the 1986-
308 1990 cohort group and 35% (95% CI: 16-59%) in the 2036-2040 cohort. The direct-effect binge drinking
309 prevalence was 34% (95% CI: 15-59%) in the 2036-2040 cohort group.

310

311 The relative difference between total- and direct-effect estimates are presented in Supplementary Table 8
312 (shown visually in Figure 3), quantifying the magnitude of the effect that each set of covariates had on
313 binge drinking rates for each cohort group. Compared with direct-effect estimates in the first cohort
314 group, the total-effect estimates were 227% lower for age 18 (i.e., 11% vs. 36%), 26% lower for ages 23-
315 24, and 3% higher for ages 29-30 in the 2036-2040 cohort group.

316

317 Gender-stratified estimates

318 Gender-stratified binge drinking prevalence estimates for the 1976-1980 through 2036-2040 cohort
319 groups are presented in Figure 4 and Supplementary Table 7. At age 18, total-effect binge drinking
320 decreased from 60% (95% CI: 56-79%) to 14% (95% CI: 5-38%) among men and from 37% (95% CI:
321 30-45%) to 9% (95% CI: 2-23%) among women. After adjustment, direct-effect estimates were 44%
322 (95% CI: 16-76%) and 29% (95% CI: 10-60%) among men and women in 2036-2040 cohort group. At
323 ages 23-24, total-effect binge drinking decreased from 54% (95% CI: 44-63%) to 40% (95% CI: 20-65%)
324 among men and decreased from 29% (95% CI: 21-37%) to 28% (95% CI: 12-51%) among women. After
325 adjustment, direct-effect estimates were 54% (95% CI: 46-62%) among men and 33% (95% CI: 46-62%)
326 among women in 2036-2040 cohort group. At ages 29-30, total-effect binge drinking trends did not
327 change from 41% (95% CI: 34-49%) among men and increased from 20% (95% CI: 12-32%) to 28%
328 (95% CI: 11-56%) among women. Compared with total effects, direct-effect binge drinking estimates
329 were 43% (95% CI: 19-71%) among men and 24% (95% CI: 8-52%) among women in 2036-2040 cohort
330 groups. The relative difference between total- and direct-effect estimates stratified by gender are
331 presented in Supplementary Table 8.

332

333

334 Discussion

335

336 Forecasting provides useful information to estimate future burden from health outcomes and behaviors
337 and understand important determinants of future health patterns, in order to determine resources and
338 priorities accordingly. To our knowledge, this is the first paper to apply forecasting methods to estimate
339 future binge drinking trends in young adults. We highlight four key findings. First, in line with observed
340 trends in binge drinking, total-effect rates of binge drinking through 2040 were estimated to continue to
341 decline at age 18, holding steady at ages 23-24, and increase slightly at ages 29-30. Second, these trends
342 were partially due to changing drinking norms and related substance use, though these indirect effects
343 explained less of the total trend as age increases. Third, gender-stratified forecasts suggested further
344 convergence in binge drinking prevalence between men and women, though trends in base rates differ by
345 age. Fourth, gender-specific convergences were partially due to changing trends in binge drinking norms
346 and cigarette and marijuana use.

347
348 Binge drinking declined substantially among 18-year-olds from 1976-2015 (Miech et al., 2019; Patrick et
349 al., 2017; Schulenberg et al., 2019), and our models extend those trends to forecast further decline, falling
350 to nearly 10% by 2040. This echoes previous work in this (Patrick et al., 2019) and similar samples
351 (Gruza et al., 2009), showing decreases among young adults. Among ages 23-24 similar trends were also
352 forecasted, though less sharply across cohorts. Binge drinking has typically peaked between ages 20-23
353 (Patrick et al., 2019), therefore, decreasing trends in this age group are a hopeful sign that binge drinking
354 will attenuate during the transition to adulthood. On the other hand, we found continued increases in
355 binge drinking among ages 29-30, concordant with recent evidence of an upward shift in the peak ages of
356 binge drinking (Patrick et al., 2019). While relatively small (i.e., from 30-35% over 13 cohorts), this trend
357 suggests that strategies to reduce binge drinking should be prioritized throughout early adulthood.

358
359 Across all ages, adjustment for several sets of sociodemographic determinants of binge drinking
360 suggested that the strongest drivers of past and future binge drinking patterns are related to alcohol norms,
361 peer use, and use of cigarettes and marijuana. In other words, had these variables not changed in the way
362 they did, change in binge drinking trends would have been far less substantial. This builds on prior work
363 showing the importance of binge drinking disapproval (Keyes et al., 2012), by examining multiple
364 measures of norms about substance use, and forecasting how these measures may influence future rates of
365 binge drinking. By age, we found evidence that trends in drinking norms have been reversing across early
366 adulthood. Specifically, disapproval and perceived risk of binge drinking have been increasing among 18-
367 year-olds and decreasing among ages 29-30 (age 23-24 time trends are somewhat static). Likewise, trends
368 in any drinking among the respondent's friends has followed similar patterns. Concordant with other
369 surveys, we found that use of cigarettes and marijuana also decreased across all ages. Use of these

370 substances often co-occurs with alcohol (Bobo and Husten, 2000; Midanik et al., 2007; Weitzman and
371 Chen, 2005), and while their decreasing popularity can be considered public health successes in their own
372 right, they also appear to be meaningfully related to decreasing binge drinking trends. Additionally, the
373 effects of adjusting for covariates diminished with age, suggesting that either any cohort effects at age 29-
374 30 are completely mediated by age 18 trends, or different determinants of binge drinking behavior are
375 more important at later ages (e.g., income). Taken together, these findings suggest that future prevention
376 activities should continue to focus on changing norms among young adults and consider additional
377 determinants of binge drinking trends that may be more salient among adults approaching middle
378 adulthood.

379
380 Gender-stratified forecasts show a continuation in the narrowing of binge drinking prevalence between
381 young men and women (Keyes et al., 2019), however, patterns in the gender-specific base rates changed
382 with age. Among those ages 18 and 23-24, the narrowing was due to greater decreases in binge drinking
383 among men than women, while among ages 29-30, the narrowing was driven by greater increases in binge
384 drinking among women than men. In line with prior research (Keyes et al., 2019), these estimates
385 highlight the need to integrate historical and developmental perspectives to accurately describe age
386 differences in the present and future burden of binge drinking. Attenuating alcohol use among women as
387 they approach middle adulthood should be a priority.

388
389 At all ages, adjustment for alcohol use norms and co-occurring substances diminished the observed
390 gender convergence, which implies that historical variation in these covariates has been a partial driver of
391 gender convergence. That is, had covariates not changed the way they did, gender convergence would be
392 less evident at every age. However, there were distinct patterns in rates among men and women. At ages
393 18 and 23-34, trends in total effects (i.e., unadjusted estimates) were lower than direct effects (i.e.,
394 covariate adjusted estimates) for both genders; however, the gap between total effects and direct effects
395 was larger for men than women. However, at ages 29-30, trends in total effects were lower than direct
396 effects for males but higher for females, suggesting that for females the changes in binge drinking
397 determinants have increased binge drinking levels. This finding for females is contrary to what was found
398 at other ages, however it is consistent with prior research that has found changing acceptability of heavy
399 alcohol use among adult women (Keyes et al., 2012; Skog, 1985). These norms have changed in concert
400 with (or as a result of) targeted marketing toward women in this age group through marketing (Kindy and
401 Keating, 2016; Petticrew et al., 2017) and targeted social media campaigns (Lyons et al., 2017). These
402 trends appear to be especially strong among women with higher socio-economic status (Kuntsche et al.,

403 2011; Lui et al., 2018), a group which has grown substantially during the study period, driven by
404 increasing college attendance and employment.

405
406 This study highlighted the role of modifiable risk factors in influencing binge drinking prevalence. Norms
407 may be modified through targeted interventions to increase knowledge of the danger and decrease the
408 social acceptability of heavy alcohol use, adapting prior research on college campuses (Borsari and Carey,
409 2003; Scott-Sheldon et al., 2009) as well as decades of successful smoking cessation interventions
410 (Bruvold, 1993; Viswesvaran and Schmidt, 1992). The impact of decreased cigarette and marijuana use
411 on binge drinking suggests that policies to diminish the use of one harmful substance may have spillover
412 effects for other co-occurring substances. These types of interventions are consistently needed, in order to
413 counteract the actions of alcohol producers to influence norms for alcohol use in emerging priority groups
414 (e.g., young adult women).

415
416 Limitations

417
418 These findings should be interpreted in light of the following limitations. All survey responses were based
419 on self-report, the sample design excluded high school drop-outs, and attrition was higher among
420 substance users than non-users. These issues are addressed by using attrition weights, however, there may
421 be residual selection bias. There were additional limitations concerning the forecasting approach. First,
422 forecasting introduces inherent uncertainty into regression models, which in the MTF were amplified by
423 the imputation-based forecasting procedure. However, we accounted for this uncertainty given the MTF
424 data structure at three points in the methods: 1) future covariate values are randomly chosen (from a
425 known distribution); 2) individual forecasted binge drinking status is multiply imputed with 20 imputed
426 datasets, which are then pooled and corrected to avoid spuriously small standard errors, and 3) confidence
427 intervals were horizon-adjusted, to acknowledge the uncertainty in forecasting long-term future values.
428 Furthermore, the utility of forecasting methods is not to provide one correct estimate, but rather predict
429 general trends. We have transparently described how we derived and validated model estimates to
430 understand the levels of morbidity that might be expected, given patterns of several sets of binge drinking
431 determinants. Second, we sought to identify the potential effect of determinants of future binge drinking
432 by lagging covariates, however, lag time may differ for specific determinants (i.e., short for norms, longer
433 for having children). In general, prior research suggests that norm changes typically precede behavior
434 changes (Borsari and Carey, 2003) and interventions that reduce multiple comorbid substance use
435 outcomes would be highly effective from a public health standpoint. Third, from 1976-2004, racial
436 identification was limited to one response per person. Beginning in 2005, respondents were able to select

437 multiple races; however, to maintain consistency across all years of observation, we limited race to a
438 single response and included multiple responses in the “Other” category. Future research should include a
439 more detailed study of binge drinking trends among individuals who identify as having multiple races.
440 Finally, unmeasured covariates may be important determinants of forecasted estimates. However, the
441 initial validation steps suggested that the forecasting model performed well overall. While beyond the
442 scope of the current analysis, future research could optimize forecasting models by age and gender,
443 incorporating more variables and effect modifiers.

444
445 Fourth, in building the forecasting models, we made the unverifiable assumption that the observed
446 variables will follow the same future trends. Trends in most covariates were relatively consistent from
447 1976 through 2015, increasing our confidence that a similar continuation was the most valid assumption
448 regarding future trends. However, unanticipated events may substantially impact forecasted estimates. For
449 example, our forecasting did not account for the COVID-19 pandemic, which has influenced widespread
450 social, economic, and health trends that will likely impact short- and long-term rates of binge drinking
451 (Clay and Parker, 2020). While empirical evidence is currently limited (Dumas et al., 2020; Pollard et al.,
452 2020), public health researchers have issued growing concern around an increase in alcohol intake and
453 alcohol-related harms (Clay and Parker, 2020; Ramalho, 2020). More research is needed to further
454 understand the long-term effects of the pandemic on binge drinking, and future forecasting models should
455 incorporate additional predictors as they become available.

456
457 Conclusion

458
459 This paper utilized data from a large US nationally representative study of 40 cohorts of high school
460 seniors followed into adulthood, in order to understand historical and developmental trends in alcohol use
461 and related factors and forecast future binge drinking through 2040. Overall, we identified important
462 gender- and age-specific differences in forecasted future levels of binge drinking, and important
463 determinants of those trends. No one study can estimate a true observed effect, much less a true future
464 effect; however, forecasting methods are valuable tools, and robust future patterns that emerge across
465 multiple studies will be useful to inform a proactive model of public health planning to reduce the harm
466 caused by binge drinking from adolescence to adulthood.

467 **References**

- 468 Bachman, J.G., Wadsworth, K.N., O'Malley, P.M., Schulenberg, J., Johnston, L.D.,
469 1997. Marriage, divorce, and parenthood during the transition to young
470 adulthood: Impacts on drug use and abuse. *Health risks and developmental*
471 *transitions during adolescence* 246–279.
- 472 Bobo, J.K., Husten, C., 2000. Sociocultural influences on smoking and drinking. *Alcohol*
473 *Research & Health* 24, 225.
- 474 Borsari, B., Carey, K.B., 2003. Descriptive and injunctive norms in college drinking: a
475 meta-analytic integration. *Journal of studies on alcohol* 64, 331–341.
- 476 Bray, F., Møller, B., 2006. Predicting the future burden of cancer. *Nature Reviews*
477 *Cancer* 6, 63–74.
- 478 Bruvold, W.H., 1993. A meta-analysis of adolescent smoking prevention programs.
479 *American Journal of Public Health* 83, 872–880.
- 480 Buuren, S. van, Groothuis-Oudshoorn, K., 2010. mice: Multivariate imputation by
481 chained equations in R. *Journal of statistical software* 1–68.
- 482 Cheng, H.G., Anthony, J.C., 2017. A new era for drinking? Epidemiological evidence on
483 adolescent male–female differences in drinking incidence in the United States
484 and Europe. *Social Psychiatry and Psychiatric Epidemiology* 52, 117–126.
485 <https://doi.org/10.1007/s00127-016-1318-0>
- 486 Cheng, H.G., Cantave, M.D., Anthony, J.C., 2016. Taking the First Full Drink:
487 Epidemiological Evidence on Male-Female Differences in the United States.
488 *Alcoholism, clinical and experimental research* 40, 816–25.
489 <https://doi.org/10.1111/acer.13028>
- 490 Chikritzhs, T.N., Stockwell, T.R., Jonas, H.A., Heale, P.F., Dietze, P.M., 2001. Mortality
491 and life-years lost due to alcohol: a comparison of acute and chronic causes.
492 *Medical Journal of Australia* 174, 281–284.
- 493 Choi, J., Cho, Y., Shim, E., Woo, H., 2016. Web-based infectious disease surveillance
494 systems and public health perspectives: a systematic review. *BMC public health*
495 16, 1238.
- 496 Chretien, J.-P., George, D., Shaman, J., Chitale, R.A., McKenzie, F.E., 2014. Influenza
497 forecasting in human populations: a scoping review. *PloS one* 9, e94130.

498 Clay, J.M., Parker, M.O., 2020. Alcohol use and misuse during the COVID-19
499 pandemic: a potential public health crisis? *The Lancet Public Health* 5, e259.

500 de Vocht, F., Tilling, K., Pliakas, T., Angus, C., Egan, M., Brennan, A., Campbell, R.,
501 Hickman, M., 2017. The intervention effect of local alcohol licensing policies on
502 hospital admission and crime: a natural experiment using a novel Bayesian
503 synthetic-time-series method. *J Epidemiol Community Health* 71, 912–918.

504 Dumas, T.M., Ellis, W., Litt, D.M., 2020. What does adolescent substance use look like
505 during the COVID-19 pandemic? Examining changes in frequency, social
506 contexts, and pandemic-related predictors. *Journal of Adolescent Health* 67,
507 354–361.

508 Grucza, R.A., Norberg, K.E., Bierut, L.J., 2009. Binge drinking among youths and young
509 adults in the United States: 1979–2006. *Journal of the American Academy of*
510 *Child & Adolescent Psychiatry* 48, 692–702.

511 Holmila, M., Raitasalo, K., 2005. Gender differences in drinking: why do they still exist?
512 *Addiction* 100, 1763–1769.

513 Hyndman, R.J., Athanasopoulos, G., 2018. *Forecasting: principles and practice.*
514 *OTexts.*

515 Jager, J., Keyes, K.M., Schulenberg, J.E., 2015. Historical variation in young adult binge
516 drinking trajectories and its link to historical variation in social roles and minimum
517 legal drinking age. *Developmental psychology* 51, 962.

518 Keyes, K.M., Jager, J., Mal-Sarkar, T., Patrick, M.E., Rutherford, C., Hasin, D., 2019. Is
519 there a recent epidemic of women’s drinking? A critical review of national studies.
520 *Alcoholism: clinical and experimental research* 43, 1344–1359.

521 Keyes, K.M., Schulenberg, J.E., O’Malley, P.M., Johnston, L.D., Bachman, J.G., Li, G.,
522 Hasin, D., 2012. Birth cohort effects on adolescent alcohol use: The influence of
523 social norms from 1976-2007. *Arch Gen Psychiatry* 69, 1304–1313.
524 <https://doi.org/10.1001/archgenpsychiatry.2012.787>

525 Kindy, K., Keating, D., 2016. For women, heavy drinking has been normalized. That’s
526 dangerous. *The Washington Post* [Internet]. Washington, DC: The Washington
527 Post.

- 528 Kuntsche, S., Knibbe, R.A., Kuntsche, E., Gmel, G., 2011. Housewife or working
529 mum—each to her own? The relevance of societal factors in the association
530 between social roles and alcohol use among mothers in 16 industrialized
531 countries. *Addiction* 106, 1925–1932.
- 532 Ladrón de Guevara, F., Washington, S.P., Oh, J., 2004. Forecasting Crashes at the
533 Planning Level: Simultaneous Negative Binomial Crash Model Applied in Tucson,
534 Arizona. *Transportation Research Record* 1897, 191–199.
535 <https://doi.org/10.3141/1897-25>
- 536 Lemstra, M., Bennett, N.R., Neudorf, C., Kunst, A., Nannapaneni, U., Warren, L.M.,
537 Kershaw, T., Scott, C.R., 2008. A Meta-analysis of Marijuana and Alcohol Use by
538 Socio-economic Status in Adolescents Aged 10–15 Years. *Can J Public Health*
539 99, 172–177. <https://doi.org/10.1007/BF03405467>
- 540 Little, T.D., Rhemtulla, M., 2013. Planned missing data designs for developmental
541 researchers. *Child Development Perspectives* 7, 199–204.
542 <https://doi.org/10.1111/cdep.12043>
- 543 Lui, C.K., Kerr, W.C., Mulia, N., Ye, Y., 2018. Educational differences in alcohol
544 consumption and heavy drinking: An age-period-cohort perspective. *Drug and*
545 *alcohol dependence* 186, 36–43.
- 546 Lyons, A., McCreanor, T., Goodwin, I., Barnes, H.M., 2017. Youth drinking cultures in a
547 digital world: Alcohol, social media and cultures of intoxication. Taylor & Francis.
- 548 Madley-Dowd, P., Hughes, R., Tilling, K., Heron, J., 2019. The proportion of missing
549 data should not be used to guide decisions on multiple imputation. *Journal of*
550 *Clinical Epidemiology* 110, 63–73. <https://doi.org/10.1016/j.jclinepi.2019.02.016>
- 551 Midanik, L.T., Tam, T.W., Weisner, C., 2007. Concurrent and simultaneous drug and
552 alcohol use: results of the 2000 National Alcohol Survey. *Drug and alcohol*
553 *dependence* 90, 72–80.
- 554 Miech, R., Johnston, L., O'Malley, P., Bachman, J., Schulenberg, J., Patrick, M., 2020.
555 Monitoring the Future national survey results on drug use, 1975-2019: Volume I,
556 Secondary school students.

557 Miech, R., Johnston, L.D., O'Malley, P.M., Bachman, J.G., Schulenberg, J.E., Patrick,
558 M.E., 2019. Monitoring the Future national survey results on drug use, 1975–
559 2018: Volume I, Secondary school students. Ann Arbor: Institute for Social
560 Research, The University of Michigan.

561 Naimi, T.S., Brewer, R.D., Mokdad, A., Denny, C., Serdula, M.K., Marks, J.S., 2003.
562 Binge drinking among US adults. *Jama* 289, 70–75.

563 Noble, D.W.A., Nakagawa, S., 2018. Planned missing data design: stronger inferences,
564 increased research efficiency and improved animal welfare in ecology and
565 evolution. *bioRxiv* 247064. <https://doi.org/10.1101/247064>

566 O'Connor, P.J., 2005. Forecasting of spinal cord injury annual case numbers in
567 Australia. *Archives of physical medicine and rehabilitation* 86, 48–51.

568 Patrick, M.E., Terry-McElrath, Y.M., Lanza, S.T., Jager, J., Schulenberg, J.E., O'Malley,
569 P.M., 2019. Shifting age of peak binge drinking prevalence: historical changes in
570 normative trajectories among young adults aged 18 to 30. *Alcoholism: Clinical
571 and Experimental Research* 43, 287–298.

572 Patrick, M.E., Terry-McElrath, Y.M., Miech, R.A., Schulenberg, J.E., O'Malley, P.M.,
573 Johnston, L.D., 2017. Age-specific prevalence of binge and high-intensity
574 drinking among US young adults: changes from 2005 to 2015. *Alcoholism:
575 Clinical and Experimental Research* 41, 1319–1328.

576 Patrick, M.E., Wightman, P., Schoeni, R.F., Schulenberg, J.E., 2012. Socioeconomic
577 status and substance use among young adults: a comparison across constructs
578 and drugs. *Journal of studies on alcohol and drugs* 73, 772–782.

579 Petticrew, M., Shemilt, I., Lorenc, T., Marteau, T., Melendez-Torres, G., O'Mara-Eves,
580 A., Stautz, K., Thomas, J., 2017. Alcohol advertising and public health: systems
581 perspectives versus narrow perspectives. *J Epidemiol Community Health* 71,
582 308–312.

583 Pigott, T.D., 2009. Handling missing data, in: *The Handbook of Research Synthesis and
584 Meta-Analysis*. Russell Sage Foundation, New York, NY, US, pp. 399–416.

585 Pollard, M.S., Tucker, J.S., Green, H.D., 2020. Changes in Adult Alcohol Use and
586 Consequences During the COVID-19 Pandemic in the US. *JAMA Netw Open* 3,
587 e2022942. <https://doi.org/10.1001/jamanetworkopen.2020.22942>

588 Ramalho, R., 2020. Alcohol consumption and alcohol-related problems during the
589 COVID-19 pandemic: a narrative review. *Australas Psychiatry* 28, 524–526.
590 <https://doi.org/10.1177/1039856220943024>

591 Rhemtulla, M., Little, T., 2012. Tools of the trade: Planned missing data designs for
592 research in cognitive development. *Journal of Cognition and Development* 13,
593 10.1080/15248372.2012.717340. <https://doi.org/10.1080/15248372.2012.717340>

594 Robinson, W.R., Keyes, K.M., Utz, R.L., Martin, C.L., Yang, Y., 2013. Birth cohort
595 effects among US-born adults born in the 1980s: foreshadowing future trends in
596 US obesity prevalence. *Int J Obes* 37, 448–454.
597 <https://doi.org/10.1038/ijo.2012.66>

598 Rubin, D.B., 2004. Multiple imputation for nonresponse in surveys. John Wiley & Sons.

599 Schulenberg, J., Johnston, L., O'Malley, P., Bachman, J., Miech, R., Patrick, M., 2020.
600 Monitoring the Future national survey results on drug use, 1975-2019: Volume II,
601 college students and adults ages 19-60.

602 Schulenberg, J.E., Johnston, L.D., O'Malley, P.M., Bachman, J.G., Miech, R.A., Patrick,
603 M.E., 2019. Monitoring the Future national survey results on drug use, 1975-
604 2018: Volume II, college students and adults ages 19-60. Ann Arbor: Institute for
605 Social Research, The University of Michigan.

606 Scott-Sheldon, L.A., Demartini, K.S., Carey, K.B., Carey, M.P., 2009. Alcohol
607 interventions for college students improves antecedents of behavioral change:
608 Results from a meta-analysis of 34 randomized controlled trials. *Journal of Social
609 and Clinical Psychology* 28, 799–823.

610 Seedat, S., Scott, K.M., Angermeyer, M.C., Berglund, P., Bromet, E.J., Brugha, T.S.,
611 Demyttenaere, K., de Girolamo, G., Haro, J.M., Jin, R., Karam, E.G., Kovess-
612 Masfety, V., Levinson, D., Medina Mora, M.E., Ono, Y., Ormel, J., Pennell, B.-E.,
613 Posada-Villa, J., Sampson, N.A., Williams, D., Kessler, R.C., 2009. Cross-
614 national associations between gender and mental disorders in the World Health

615 Organization World Mental Health Surveys. *Arch. Gen. Psychiatry* 66, 785–795.
616 <https://doi.org/10.1001/archgenpsychiatry.2009.36>

617 Settersten Jr, R.A., 2007. The new landscape of adult life: Road maps, signposts, and
618 speed lines. *Research in Human Development* 4, 239–252.

619 Skog, O., 1985. The collectivity of drinking cultures: a theory of the distribution of
620 alcohol consumption. *British journal of addiction* 80, 83–99.

621 Slade, T., Chapman, C., Swift, W., Keyes, K., Tonks, Z., Teesson, M., 2016. Birth
622 cohort trends in the global epidemiology of alcohol use and alcohol-related harms
623 in men and women: systematic review and metaregression. *BMJ open* 6,
624 e011827.

625 Soyiri, I.N., Reidpath, D.D., 2012. Evolving forecasting classifications and applications
626 in health forecasting. *Int J Gen Med* 5, 381–389.
627 <https://doi.org/10.2147/IJGM.S31079>

628 Viswesvaran, C., Schmidt, F.L., 1992. A meta-analytic comparison of the effectiveness
629 of smoking cessation methods. *Journal of Applied Psychology* 77, 554.

630 Waters, M.C., Carr, P.J., Kefalas, M., Holdaway, J.A., 2019. *Coming of age in America:
631 The transition to adulthood in the twenty-first century.* University of California
632 Press.

633 Weitzman, E.R., Chen, Y.-Y., 2005. The co-occurrence of smoking and drinking among
634 young adults in college: national survey results from the United States. *Drug and
635 alcohol dependence* 80, 377–386.

636 White, A., Castle, I.P., Chen, C.M., Shirley, M., Roach, D., Hingson, R., 2015.
637 Converging patterns of alcohol use and related outcomes among females and
638 males in the United States, 2002 to 2012. *Alcoholism: clinical and experimental
639 research* 39, 1712–1726.

640 Wood, J., Matthews, G.J., Pellowski, J., Harel, O., 2019. Comparing different planned
641 missingness designs in longitudinal studies. *Sankhya B* 81, 226–250.
642 <https://doi.org/10.1007/s13571-018-0170-5>

643

644 **Figure Legends**

645 **Figure 1.** Correspondence between age, cohort group, and year in observed and forecasted samples

646 Note: ^a forecasted population sizes are based on the approximate average sample size of observed cohort
647 groups. Dotted red line delineates observed/forecasted values.

648 **Figure 2.** Ages 18, 23-24, and 29-30 observed trends across cohorts in covariates used to forecast binge
649 drinking

650 **Note:** Standardized means correspond to the average value within each cohort, where 0 equals the
651 average value across the total sample; no line indicates no change across cohorts; years vary by age, based
652 on the number of observed cohorts; NH=non-Hispanic; Friends drink = How many of your friends drink
653 alcoholic beverages? (None–All), Risk of weekend binge = How much do you think people risk harming
654 themselves (physically or in other ways) if they have five or more drinks once or twice each weekend (No
655 Risk–Great Risk), Disapprove of weekend binge = Do you disapprove of people (18 or older) having five
656 or more drinks once or twice each weekend (Don't Disapprove–Strongly Disapprove)

657 **Figure 3.** Ages 18, 23-24, and 29-30 binge drinking prevalence (with 95% prediction intervals) from
658 1976-2040. Adjusted for demographic, big 5 social roles, and drinking norms/substance use covariates

659 Note: dotted red line depicts the beginning of the forecasted estimates; Direct effect models adjusted for:
660 Demographics: sex, high school GPA, father's and mother's highest reported, race/ethnicity; Big-5 social
661 roles: attending college full-time, not living with parents, have children, married, work full-time;
662 Drinking norms & other substance use: disapproval of having 5 or more drinks on the weekend,
663 proportion of friends who drink alcohol, perceived risk of 5 or more weekend drinks, and past-year use of
664 marijuana and cigarettes

665 **Figure 4.** Ages 18, 23-24, and 29-30 binge drinking prevalence from 1976-2040, stratified by sex.

666 Adjusted for demographic, big 5 social roles, and drinking norms/substance use covariates

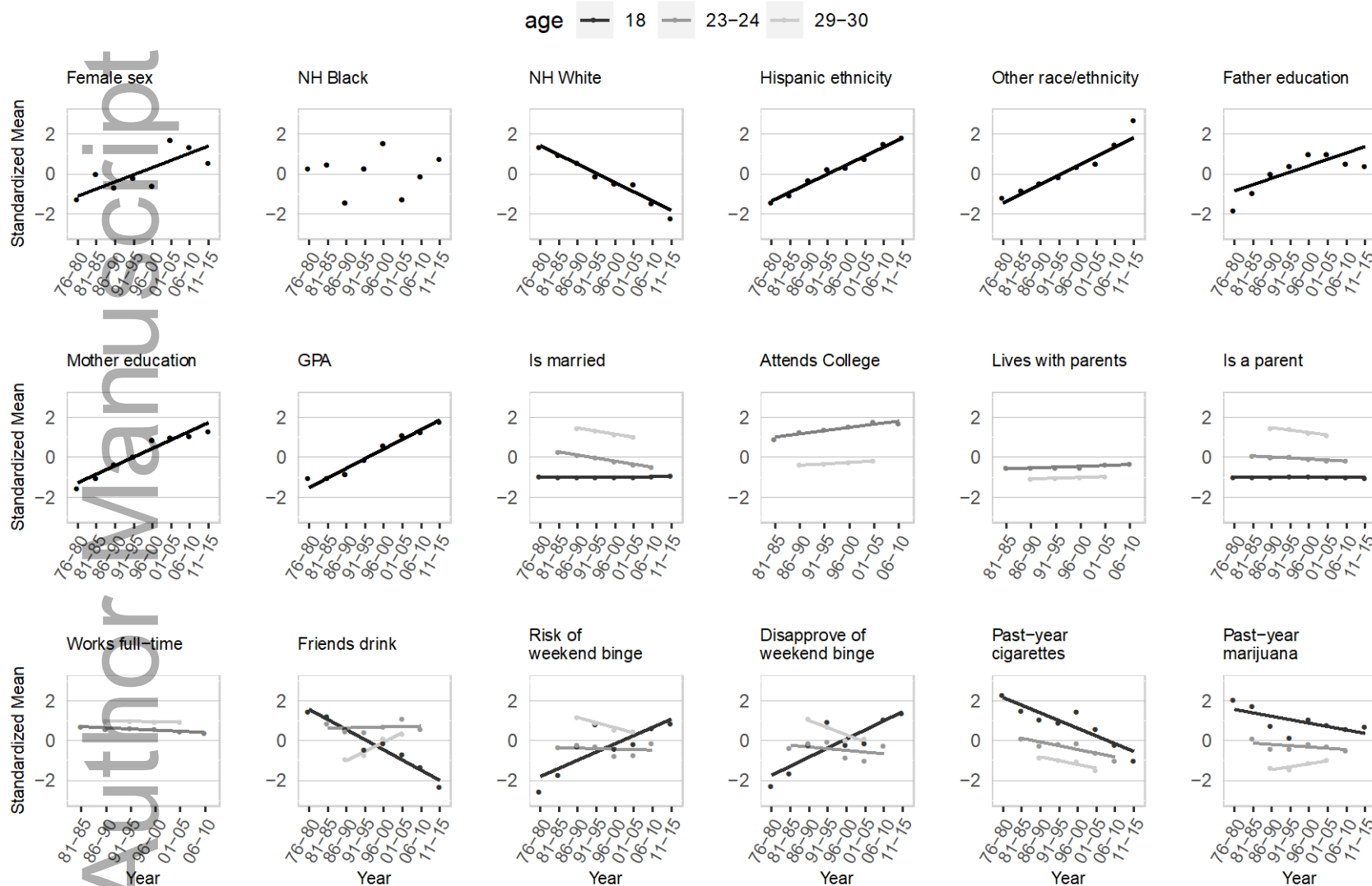
667 Note: dotted red line depicts the beginning of the forecasted estimates; Demographics: sex, high school
668 GPA, father's and mother's highest reported, race/ethnicity; Big-5 social roles: attending college full-
669 time, not living with parents, have children, married, work full-time; Drinking norms & other substance
670 use: disapproval of having 5 or more drinks on the weekend, proportion of friends who drink alcohol,
671 perceived risk of 5 or more weekend drinks, and past-year use of marijuana and cigarettes

Figure 1. Correspondence between age, cohort group, and year in observed and forecasted samples

	Year												
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)
	76-80	81-85	86-90	91-95	96-00	01-05	06-10	11-15	16-20	21-25	26-30	31-35	36-40
Age	Cohort group (base year)												
18	1	2	3	4	5	6	7	8	9	10	11	12	13
23-24		1	2	3	4	5	6	7	8	9	10	11	12
29-30			1	2	3	4	5	6	7	8	9	10	11
<i>n</i>	11888	12226	12331	12337	12266	12250	12261	12253	12200 ^a	12200	12200	12200	12200

^a forecasted population sizes are based on the approximate average sample size of observed cohort groups. Dotted red line delineates observed/forecasted values.

Figure 2. Ages 18, 23-24, and 29-30 observed trends across cohorts in covariates used to forecast binge drinking

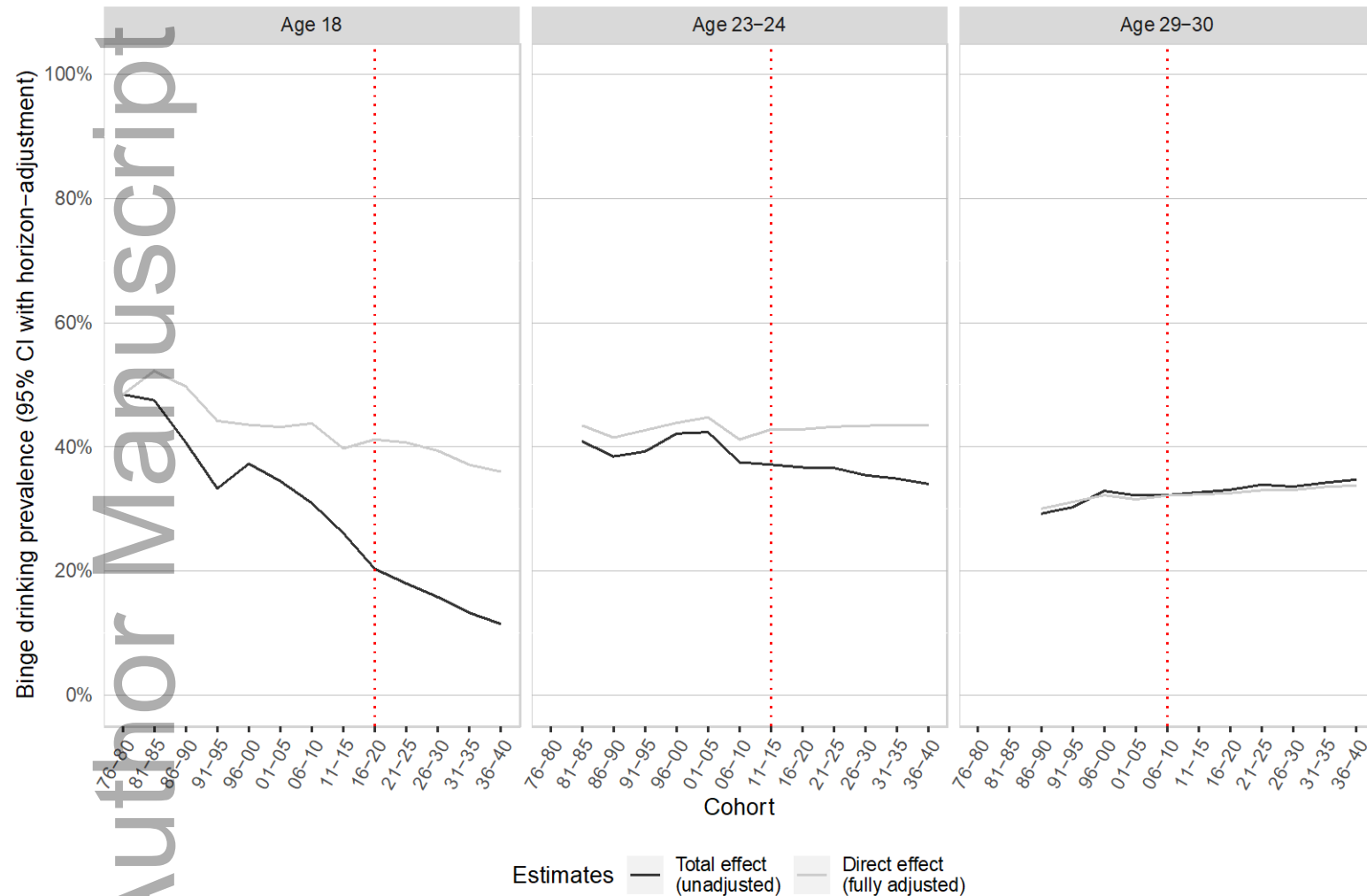


Note: Standardized means correspond to the average value within each cohort, where 0 equals the average value across the total sample; no line indicates no change across cohorts; years vary by age, based on the number of observed cohorts; NH=non-Hispanic; Friends drink = How many of your friends drink alcoholic beverages? (None-All), Risk of weekend binge = How much do you think people risk harming themselves (physically or in other ways) if they have five or more drinks once or twice each weekend (No Risk-

Great Risk), Disapprove of weekend binge = Do you disapprove of people (18 or older) having five or more drinks once or twice each weekend (Don't Disapprove–Strongly Disapprove)

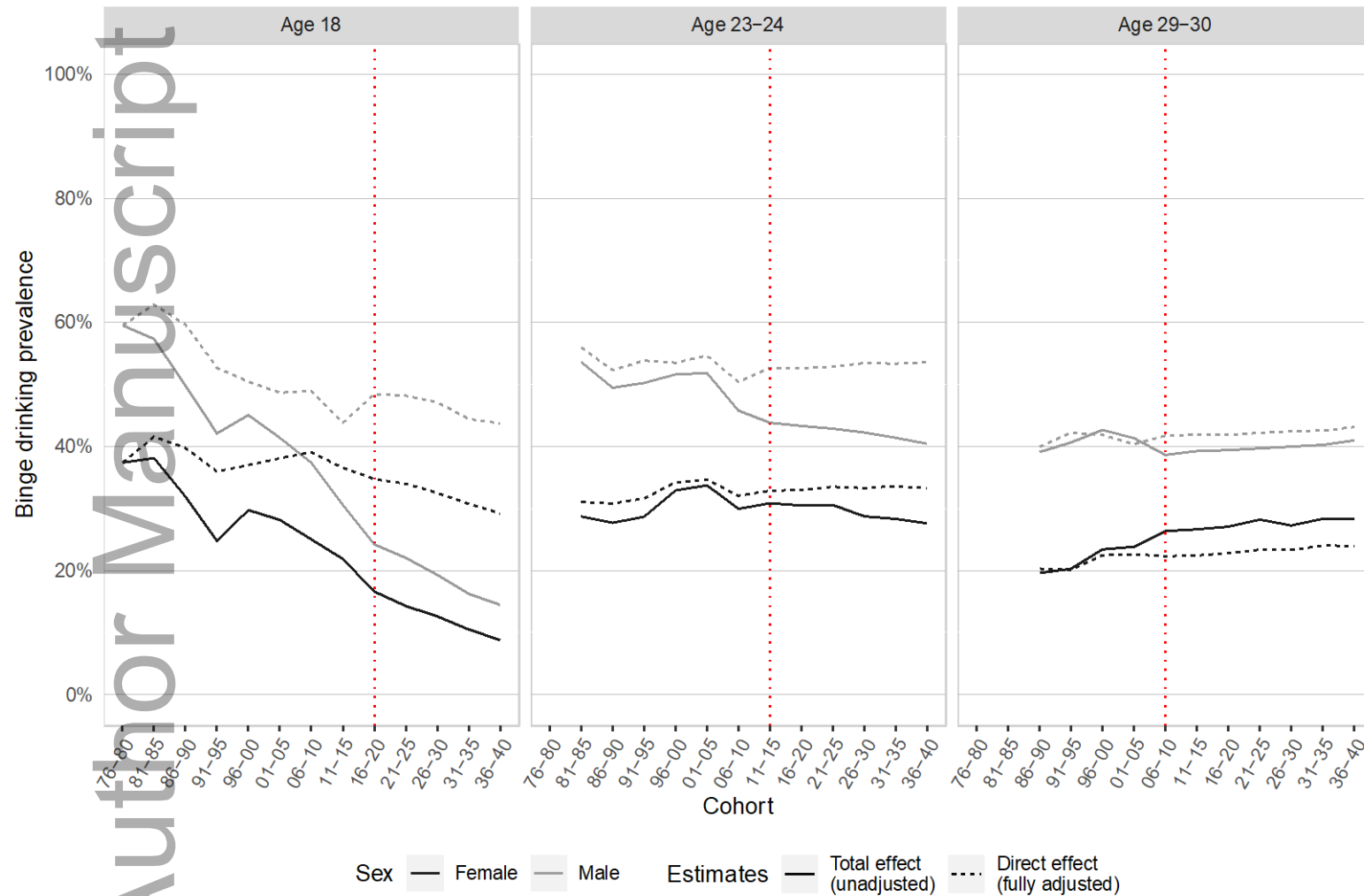
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Figure 3. Ages 18, 23-24, and 29-30 binge drinking prevalence (with 95% prediction intervals) from 1976-2040. Adjusted for demographic, big 5 social roles, and drinking norms/substance use covariates



Note: dotted red line depicts the beginning of the forecasted estimates; Direct effect models adjusted for: Demographics: sex, high school GPA, father's and mother's highest reported, race/ethnicity; Big-5 social roles: attending college full-time, not living with parents, have children, married, work full-time; Drinking norms & other substance use: disapproval of having 5 or more drinks on the weekend, proportion of friends who drink alcohol, perceived risk of 5 or more weekend drinks, and past-year use of marijuana and cigarettes

Figure 4. Ages 18, 23-24, and 29-30 binge drinking prevalence from 1976-2040, stratified by sex. Adjusted for demographic, big 5 social roles, and drinking norms/substance use covariates



Note: dotted red line depicts the beginning of the forecasted estimates; Demographics: sex, high school GPA, father's and mother's highest reported, race/ethnicity; Big-5 social roles: attending college full-time, not living with parents, have children, married, work full-time; Drinking norms & other substance use: disapproval of having 5 or more drinks on the weekend, proportion of friends who drink alcohol, perceived risk of 5 or more weekend drinks, and past-year use of marijuana and cigarettes