



Longitudinal co-trajectories of depression and alcohol problems in adults during the COVID-19 pandemic

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Abstract

We examined person-centered heterogeneity in the longitudinal co-development of depression and alcohol problems during the COVID-19 outbreak. We also investigated the risk factors (personality and coping) for being in “higher” relative to “lower” risk subgroups of combined depressive symptoms and alcohol problems. Canadian participants ($N=364$, $M_{\text{age}}=32.16$, 54.67% male) completed questionnaires four times every three months, starting approximately 2 months after Canada announced its COVID-19 State-of-Emergency. Parallel-process latent class growth analysis found evidence for three latent subgroups: a “moderate increasing depression and alcohol problems” subgroup (Class 1); a “moderate stable depression, moderate decreasing alcohol problems” subgroup (Class 2); and a “low-risk normative” subgroup (with mild depression that was stable and mild alcohol problems that decreased; Class 3). Multinomial logistic regressions found that higher levels of hopelessness, impulsivity, and boredom proneness distinguished Class 1 from Class 3. Further, lower levels of general self-efficacy distinguished Class 1 from Classes 2 and 3. Linear mixed models found that Class 1 increasingly used maladaptive avoidant coping strategies (denial, drugs/alcohol, behavioural disengagement) as the pandemic progressed, whereas Class 2 increasingly used adaptive approach-oriented strategies (planning, seeking emotional support from others). We analyzed longitudinal data to detect classes of individuals with depressive and alcohol-related difficulties during COVID-19 and to characterize the vulnerability factors for increased difficulties. Highlighting the heterogeneity in the co-trajectory of depression and alcohol problems during COVID-19 and the personality and coping factors associated with combined increases in these mental health difficulties can inform treatment practices and bolster peoples’ preparedness and resilience for future pandemics.

Keywords COVID-19 · Depression · Alcohol problems · Longitudinal · Personality · Coping

The global coronavirus (COVID-19) pandemic has significantly impacted the mental health of people worldwide. Among 1,803 Canadian adults surveyed in late-April 2020, roughly two months after Canada announced its COVID-19 State-of-Emergency, the percentage of respondents with high self-reported depression had increased from 4 percent to 10 percent since the beginning of the pandemic.

One-third of respondents with depression additionally reported increased alcohol consumption since the beginning of COVID-19 (Dozois, 2021), and a separate study (McPhee et al., 2020) found that depression severity and risky drinking were greater in May 2020, after social distancing guidelines were introduced, relative to pre-social distancing. Although some studies found increases in alcohol consumption and alcohol problems during the pandemic (e.g., Pollard et al., 2020; Tucker et al., 2022), the results of studies have been mixed, with most finding that *certain* subgroups, such as those with depressive symptoms, report an increase in their alcohol consumption and alcohol problems (Acuff et al., 2022; Baptist-Mohseni et al., 2022; Capasso et al., 2021; Shield et al., 2022). This underscores the need to study depression and alcohol-related difficulties *jointly* for a well-rounded understanding of how the pandemic has impacted peoples’ mental health and well-being.

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The complexity in the comorbidity of depression and alcohol problems

Importantly, while much of the literature supports an associative link between depression and alcohol consumption (Khantzian, 1997, 2012), many individuals who struggle with depression symptoms do not problematically consume alcohol (e.g., Pedrelli et al., 2016). Indeed, some studies find a low-to-moderate comorbidity prevalence (4–37 percent) between major depressive disorder and alcohol use disorder within the general population (see Castillo-Carniglia et al., 2019 for a review). Notably, longitudinal research suggests that even *over time*, the co-trajectories of depression and alcohol problems (the adverse consequences of alcohol consumption) is not uniform or consistent across all adults. In particular, Frohlich and colleagues (2018) and Orui and colleagues (2020) conducted longitudinal studies examining the heterogeneity in the co-trajectories of depressive symptoms and alcohol problems among Canadians after and during university, respectively. In both studies, participants completed measures of depression and alcohol problems over the course of 12–18 months and the authors used parallel-process latent class growth analysis (LCGA) to classify groups of individuals based on the simultaneous growth of these difficulties. This novel analysis has advantages relative to other statistical analyses because it studies the *comorbidity* of psychological processes over time, which allows researchers to identify distinct unobserved (latent) classes of individuals with the same developmental trajectories (see Muthen & Muthen, 2000). This person-centered analytical approach models heterogeneous longitudinal data by classifying individuals into smaller classes or subgroups with homogenous co-patterns of psychological processes over time.

Frohlich and colleagues (2018) and Orui and colleagues (2020) identified multiple latent classes of depression and alcohol problems, most consistently: a “high-risk” (comorbid) class, who had high depression that was stable and high alcohol problems that were stable; a “moderate-risk” (depression-only) class, who had high depression that was stable and low alcohol problems (Orui et al., 2020 found that the latter decreased among participants in this class, whereas Frohlich et al., 2018 found that the latter remained stable among participants in this class); and a “low-risk” (normative) class, who had low depression that was stable and low alcohol problems that decreased. Together, these studies highlight the *heterogeneity* and hence, complexity in the comorbidity of depression and alcohol problems over time.

Unfortunately, at the time of writing the present work, the number of COVID-19 cases and deaths continue to rise, indicating that COVID-19 remains a global health

emergency (World Health Organization, 2022). Equally troubling, epidemiological research suggests that the probability of future pandemics that are comparable to COVID-19 in severity is increasing (e.g., Marani et al., 2021). Accordingly, continued scholarship on the impact of the pandemic on peoples’ mental health is important because it can advance prevention and intervention practices and strengthen peoples’ preparedness and resilience during these and other extraordinary times. As depression and problem drinking are clear concerns in the context of COVID-19, the current study meaningfully extended on prior research, using parallel-process LCGA to study the longitudinal co-trajectories of depression and alcohol problems in adults during the pandemic.

Risk factors for comorbid depression and alcohol problems

While identifying specific subgroups of depression and alcohol problems during the pandemic is imperative, it is equally critical to establish the factors that can increase one’s risk for being in the “high-risk” subgroup of elevated depressive symptoms and alcohol problems, in comparison to the other subgroups. Doing so can elucidate which people are at greater risk for combined depression and alcohol problems and which factors are important targets for prevention strategies and intervention approaches during such stressful circumstances. In their study, Orui et al. (2020) found that higher levels of hopelessness, impulsivity, and anxiety sensitivity differentiated the “high-risk” subgroup from the “low-risk” subgroup; higher levels of hopelessness differentiated the “moderate-risk” subgroup from the “low-risk” subgroup; and higher levels of impulsivity and lower levels of hopelessness differentiated the “high-risk” subgroup from the “moderate-risk” subgroup. In addition to these personality traits explored by Orui et al. (2020), the current study examined other factors and mechanisms that are relevant to the pandemic and that may distinguish the “high-risk” subgroup from other subgroups of combined depression and alcohol problems: *boredom proneness*, *general self-efficacy*, *external stressors*, and *coping strategies*.

Boredom proneness refers to the tendency to frequently and intensely feel bored, which is “the aversive experience of having an unfulfilled desire to engage in satisfying activity” (Fahlman et al., 2013, p. 69). Research during the COVID-19 outbreak, as well as the SARS outbreak, suggest that boredom is one of the most commonly experienced feelings associated with social distancing/quarantine (Barari et al., 2020; Droit-Volet et al., 2020; Reynolds et al., 2008), as well as that boredom proneness impairs peoples’ adherence to such critical public health

measures (see Westgate et al., 2022 for a review). Three cross-sectional studies published during COVID-19 suggest a positive relationship between boredom proneness and depression symptoms (McCurdy et al., 2022; Weiss et al., 2022; Yan et al., 2021), while most studies published during COVID-19 suggest that state boredom is positively linked to depressive symptoms and alcohol consumption (e.g., Canadian Centre on Substance Use & Addiction, 2020; Chao et al., 2020; Droit-Volet et al., 2020; Schmits & Glowacz, 2022). Given its characterization, it is plausible that boredom proneness would be a risk factor for “high-risk” depression and alcohol problems. Some work (Danckert et al., 2018) proposes that boredom proneness represents a chronic failure to respond adaptively to the state of boredom, such that people seek unhealthy experiences (e.g., self-harm) to alleviate boredom—even if *other* negative states are elicited (e.g., Bench & Lench, 2019; Havermans et al., 2015). Other work suggests that boredom proneness represents the general tendency toward maladaptive motivations—irrespective of the state of boredom—such as the desire to act destructively, the desire to act without thinking things through, the desire to avoid one’s emotions, uncertainty (not knowing what to do), and amotivation (not caring to do anything; Bambrah et al., 2020).

The pandemic has additionally engendered significant difficulties with self-efficacy, which are the beliefs in one’s abilities to meet given situational demands (Wood & Bandura, 1989). Indeed, several published studies found that self-efficacy was low or declined (compared to before the pandemic) across various samples (e.g., teachers, nurses, mental health workers, first-time parents; Cataudella et al., 2021; Pressley & Ha, 2021; Simonetti et al., 2021; Sun et al., 2020; Xue et al., 2021; Yildirim & Güler, 2020). General self-efficacy, a trait-like dimension of self-efficacy conceptualized as the propensity to view oneself as capable of meeting the demands or challenges of tasks within a wide range of contexts (Chen et al., 2001), has been consistently and positively associated with greater psychological resilience among adults during the pandemic, including less severe depressive symptoms and less hazardous alcohol use (e.g., Dogan-Sander et al., 2021; Kohls et al., 2021; Spoorthy et al., 2020; Volken et al., 2021). As evidence across multiple countries shows that people with low general self-efficacy tend to experience self-doubt when they encounter environmental challenges, think in self-debilitating ways, cope less functionally with stressors, and avoid demands that are perceived as threatening (e.g., Luszczynska et al., 2005), it follows that this trait would be a risk factor for “high-risk” depression and alcohol problems.

Additionally, positive links between external stressors and depression and alcohol problems have emerged during the pandemic. For example, research has found that income loss/financial worry, having children under 18 years of age at home, and living alone were related to increased depression,

alcohol consumption, and problematic drinking (e.g., binge drinking) at the beginning of the pandemic (e.g., Fancourt et al., 2021; Wardell et al., 2020) and throughout the pandemic (e.g., Acuff et al., 2022; Centre for Addiction and Mental Health, 2020). While these relationships suggest that these external stressors may distinguish “high-risk” from “low-risk” subgroups of depression and alcohol problems, it is alternatively possible that these external stressors are of less importance to comorbidity in comparison to internal factors. Indeed, a longitudinal study of drinking habits among adults over the course of COVID-19 found that a combined increase in alcohol use and related problems was unrelated to the aforementioned external stressors, but was related to pre-pandemic hazardous alcohol use and solitary drinking, as well as possessing a higher alcohol demand prior to COVID-19 (Baptist-Mohseni et al., 2022).

Finally, there is a clear relevance to understanding peoples’ coping strategies during stressful circumstances, as in the case of the pandemic (see Kar et al., 2021), which may mechanistically distinguish high- and low-risk subgroups of depression and alcohol problems. To reduce psychological distress, a range of coping styles can be adopted, such as *avoidant coping* characterized by cognitive or physical efforts to disengage from stressors, *problem-focused coping* characterized by doing something to modify the source of the stress or problem-solving, and *emotion-focused coping* characterized by managing and easing the emotional distress that is linked to the situation (Carver et al., 1989). Several studies examining the relationships of depression and alcohol consumption during the pandemic with these coping patterns suggest that these difficulties are positively associated with avoidant coping (e.g., using substances to cope, self-distraction, behavioural disengagement) and negatively associated with approach-oriented problem-focused coping (e.g., positive reframing, active coping) and emotion-focused coping (e.g., religion, low self-blame, low venting; e.g., Chodkiewicz et al., 2020; Gurvich et al., 2021; Shamblaw et al., 2021). Although these studies examined only the one-to-one cross-sectional relations of avoidant, problem-focused, and emotion-focused coping with depression and alcohol misuse, these patterns suggest that increased avoidant coping over time might distinguish a “high-risk” subgroup of depression and alcohol problems from a “low-risk” subgroup.

Current study

Most prior studies have reported on the overall relations between depressive problems and alcohol problems and posit that individuals with elevated depression have elevated alcohol-related difficulties (Grant et al., 2015). During the pandemic, there has been a limited number of longitudinal findings related to depression and alcohol problems and many

studies use cross-sectional designs that fail to examine *both* depression and alcohol problems as related (but distinct) aspects of psychological adjustment that can both change dynamically over time. Whereas previous studies (Baptist-Mohseni et al., 2022; Leventhal et al., 2022; Tucker et al., 2022) have examined the trajectories of alcohol consumption and alcohol problems over the course of the pandemic, the current four-wave longitudinal study extended upon this work, using data from Baptist-Mohseni et al. (2022) and conducting parallel-process LCGA to characterize the *co-trajectories* of depression and alcohol problems among Canadians during the first nine months of the COVID-19 outbreak. Based on prior research (i.e., Frohlich et al., 2018; Orui et al., 2020), we expected to find *multiple* subgroups of depression and alcohol problems, at minimum a high-risk (comorbid) subgroup; a moderate-risk (depression-only) subgroup; and a low-risk (normative) subgroup. Further, informed by previous literature, we expected personality risk factors measured at baseline (namely, anxiety sensitivity, hopelessness, impulsivity, boredom proneness, and low general self-efficacy) to increase the likelihood of combined (i.e., comorbid) depression and alcohol problems over time during the pandemic. Drawing on Baptist-Mohseni et al. (2022), we also examined the relative predictive importance of external stressors (namely, experiencing a loss of income during the pandemic, being a parent who lives with a child under 18 years of age, and living alone). Given the noted sex differences in depression and alcohol difficulties (e.g., Karpyak et al., 2016; Polak et al., 2015; Tucker et al., 2022), we also explored differences between males and females in the co-trajectories of depression and alcohol problems. Finally, based on prior research, we expected that the high-risk depression and alcohol problems subgroup would be distinguished from the other subgroups based on their increasing use of avoidant coping strategies during the pandemic.

Method

Participant and procedures

We received ethics approval from our institutional research ethics board. Data for this study was drawn from larger longitudinal studies that examined addictive behaviours during COVID-19 (see Baptist-Mohseni et al., 2022 and Wardell et al., 2020 for more details). We recruited participants via Prolific (Palan & Schitter, 2018). Adults who live in Canada, report consuming alcohol (i.e., report having more than 1 standard drink in the three months prior to wave 1), and have a high approval rating from prior studies on Prolific completed the current study. Data was collected for wave 1 between April 30th and May 4th of 2020, roughly 2 months after Canada's COVID-19 State-of-Emergency and strict public health measures (e.g., extensive closures,

stay-at-home guidelines) first went into effect. Data was collected for wave 2, wave 3, and wave 4 in July 2020, October 2020, and January 2021, respectively, during which time many public health measures remained in place throughout Canada. Participants were compensated with \$13 CAD at each wave.

For information on data exclusion, please see Wardell et al. (2020) and Baptist-Mohseni et al. (2022). The final sample comprised of 364 participants ($M_{age} = 32.16$, $SD_{age} = 9.54$; 54.67% male). Of this sample, 294 participants (80.77%) completed wave two, 263 participants (72.25%) completed wave three, and 246 participants (67.58%) completed wave four. Most participants identified as White (65.66%) and as a non-student (75.82%), with the majority possessing a College or University degree (53.85%). Half of the participants resided in Ontario (51.65%). The median income reported by participants was \$80,000–\$99,000. Nearly a quarter of the sample (24.79%) endorsed hazardous alcohol consumption at wave one (based on a score of 8 or more on the 10-item Alcohol Use Disorders Identification Test; $M_{AUDIT} = 5.95$, $SD_{AUDIT} = 5.08$). Table S1 in the Supplemental Materials presents the full demographic characteristics of the sample.

Measures

See the [Supplemental Materials](#) for a full description of the measures administered in the current study, including example items and how each item was rated. Participants' depression severity was assessed at all four waves using the *Patient Health Questionnaire* (PHQ-9; Kroenke et al., 2001, $\alpha = 0.87$ –0.90), with the total score corresponding to different levels of severity (i.e., 5–9 = Mild; 10–14 = Moderate; 15–19 = Moderately Severe; 20–27 = Severe). Participants' alcohol problems was assessed at all four waves using the *Short Inventory of Problems-Revised* (SIP-2R; Kiluk et al., 2013; $\alpha = 0.91$ –0.95). With respect to personality factors, participants' anxiety sensitivity, hopelessness, impulsivity, and sensation seeking were assessed at wave 1 using the *Substance Use Risk Profile Scale* (SURPS; Woicik et al., 2009; $\alpha = 0.71$ –0.89). Participants' boredom proneness was assessed at wave 1 using the *Short Boredom Proneness Scale* (SBPS; Struk et al., 2017; $\alpha = 0.90$). Participants' general self-efficacy was assessed at wave 1 using the *New General Self-Efficacy Scale* (NGSES; Chen et al., 2001; $\alpha = 0.90$). Drawing from Wardell and colleagues (2020) and Baptist-Mohseni and colleagues (2022), we assessed three pandemic-relevant external stressors; specifically, participants indicated if they experienced a loss of income during the pandemic; are a parent who lives with a child under 18; and live alone (each item was rated as “No” = 0 or “Yes” = 1). Finally, participants' avoidant coping (characterized by facets of self-distraction, denial, substance use, and behavioural

disengagement), problem-focused coping (characterized by facets of active coping, using instrumental supports, positive reframing, and planning), and emotion-focused coping (characterized by facets of using emotional supports, venting, humour, acceptance, religion, and self-blame) were assessed at all four waves using the *Brief Coping Orientation to Problems Experienced Inventory* (Brief-COPE; Carver, 1997; $\alpha = 0.71\text{--}0.93$; see the [Supplemental Materials](#) for a description of the three coping facets that were excluded from data analyses due to poor internal reliability).

Data analyses

First, the data was inspected for outliers and to verify other statistical testing assumptions (e.g., normality). Outlying values were Winsorized (i.e., replaced with the highest value within ± 3.29 standard deviations of a given variable's the mean score). We additionally conducted independent *t*-tests to examine potential baseline differences between participants with complete data across all four waves ($n = 246$) and participants with incomplete data across all four waves ($n = 118$) on the continuous variables at wave one.

Second, parallel-process LCGA was used to identify unique latent (unobserved) classes or subgroups of participants based on their initial levels of depression (PHQ-9) and alcohol problems (SIP-2R) and based on the changes in these difficulties over the first nine months of the pandemic. More specifically, models with one latent class to six latent classes were run and tested consecutively to determine how many classes the data best supports. Three fit indices (Jung and Wickrama, 2007) were used to compare the six models. The sample size-adjusted Bayesian Information Criterion (SA-BIC) is a relative fit index, with lower values suggesting better model fit. As reported by Raftery (1995), a difference of 10 between models in the SA-BIC suggests superior fit. Entropy characterizes a model's quality for classifying participants into smaller subgroups, with values ≥ 0.80 suggesting that the model's overall classification quality is good (Ram & Grimm, 2009). To determine whether or not a model with "*k*" latent subgroups fits the data statistically significantly better than a model with "*k* - 1" latent subgroups, the parametric bootstrapped likelihood ratio test (BLRT) was used (Nylund et al., 2007). Drawing on Wickrama and colleagues (2016), we additionally examined the class size when selecting the class model, ensuring that the size of the smallest class was not less than 5 percent of the overall sample. Finally, to ensure that the latent classes were unique and theoretically meaningful, we examined all six models visually by graphing the co-trajectories of depression and alcohol problems of each class within each model (Williams & Kibowski, 2016).

Third, multinomial logistic regressions were estimated to explore whether personality and pandemic-relevant external stressors predict participants' membership in the depression-alcohol problems subgroups. The first model examined anxiety sensitivity, hopelessness, impulsivity, and sensation seeking (following Orui et al., 2020), and the second model examined boredom proneness and general self-efficacy. External stressors (i.e., experiencing a loss of income during the pandemic, being a parent who lives with a child under 18, and living alone) were included in these models in order to understand the relative predictive importance of internal versus external factors on class membership. Biological sex was also included as a predictor.

Finally, linear mixed models (LMMs) were estimated to explore differences between the depression-alcohol problems classes on avoidant coping, problem-focused coping, and emotion-focused coping over time during the pandemic. The class (subgrouping) variable was represented with dummy coded variables, which were used to create the "*Class by Time*" interaction terms. Age, biological sex, and race (non-White versus White) were included in all LMMs as covariates. Prior to these LMMs, we investigated the intercept and the slope of each coping strategy outcome variable and we observed variability across participants in both parameter estimates for each outcome variable, which suggests that these parameter estimates are not fixed across participants. Thus, we specified random intercepts and random slopes within each LMM. Further, in each LMM, we initially modelled the linear effect of time and the quadratic effect of time, as well as their respective interactions with class. As the quadratic effect of time and the "*quadratic time by class*" interaction were not supported across all models (all p 's > 0.05), these two terms were removed from all of the LMMs in order to streamline the interpretation of the linear effect of time and for parsimony. Full information maximum likelihood was used to estimate the parallel-process LCGAs and the LMMs.

Results

Descriptive statistics

Table S2 in the Supplemental Materials presents descriptive data for the continuous variables included in the parallel-process LCGA, multinomial logistic regressions, and LMMs. With respect to the pandemic-relevant external stressors, 41.21% of participants reported experiencing a loss of income during the pandemic, 20.39% of participants identified as parents who live with at least one child under 18 during the pandemic, and 12.91% of participants reported lived alone during the pandemic. The independent *t*-tests indicated no significant baseline differences between

participants who had data at all four waves and participants who had missing data at one or more waves on all continuous wave one measures (all p 's > 0.05).

Subgroups for combined depression and alcohol problems

Table 1 presents the results for the models consisting of one latent class to six latent classes, estimated using parallel-process LCGA. Across all models, the SA-BIC index decreased, with a difference greater than 10 between each model (i.e., the model with “ k ” classes) and the previous model (i.e., the model with “ $k - 1$ ” classes). Across all models, the entropy values were above 0.80, which suggests that the overall classification quality was good, and the parametric BLRT was significant (all p 's < 0.001). However, the sizes for models with four, five, and six classes were small (< 4 percent of the sample). These models also possessed low classification probabilities (specifically < 0.49), which suggests that the models were not accurately classifying participants into the

smaller classes/subgroups. Furthermore, models with four, five, and six classes had classes with similar co-trajectories of depression severity and alcohol problems across the study's four waves, which suggests that the classes were not meaningfully distinct or unique. Accordingly, upon considering all fit statistics, class sizes, and classification probabilities, as well as graphing the co-trajectories for each class in all of the six models, we determined that the three-class solution, which possessed high classification probabilities (> 0.87), was the most interpretable model.

The first class (6.32% of participants in the sample, $n = 23$) had moderate initial levels of both depression and alcohol problems that both significantly *increased* over the course of the COVID-19 pandemic (i.e., from wave one to wave four). The second class (7.69% of participants in the sample, $n = 28$) had mild-to-moderate initial levels of depression that remained stable over time and moderate initial levels of alcohol problems that significantly *decreased* over time. The third class (85.99% of participants in the sample, $n = 313$) had mild initial levels of both depression and alcohol problems, with the latter significantly *decreasing* over time. In subsequent sections, we refer to these three classes as followed: “moderate increasing depression and alcohol problems” subgroup (Class 1), “moderate stable depression, moderate decreasing alcohol problems” subgroup (Class 2), and “low-risk normative” subgroup (Class 3). Table 2 and Fig. 1 present the parameter estimates and graphs, respectively, of each latent class.

Risk factors

Next, multinomial logistic regressions were estimated to determine the predictive roles of personality, pandemic-relevant external stressors, and biological sex on participants'

Table 1 Fit Information for the Parallel Process Latent Class Growth Models

Class #:	SA-BIC	Entropy	Parametric BLRT p -value	Smallest Class Size (%)
1	11974.391	N/A	N/A	100
2	11671.368	0.973	< .001	8.0
3	11551.286	0.954	< .001	6.3
4	11462.690	0.954	< .001	3.7
5	11395.905	0.914	< .001	3.2
6	11363.622	0.906	< .001	3.0

The fit information of the retained model is bolded

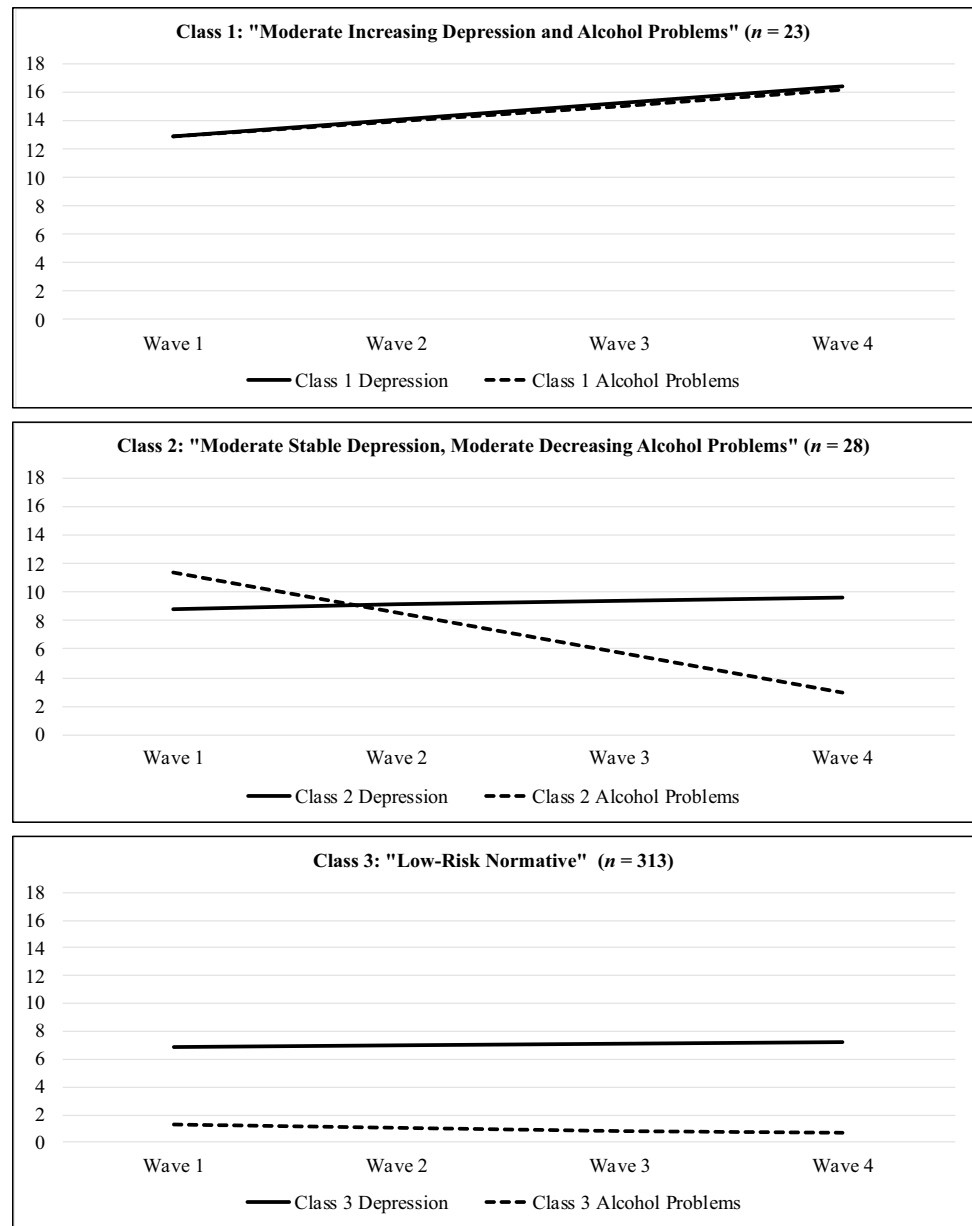
Table 2 Parameter Estimates for Parallel Process Latent Class Growth Model – Three Latent Classes

Class:		Depression Severity (PHQ-9)	Alcohol Problems (SIP-2R)
1 – “Moderate Increasing Depression and Alcohol Problems” ($n = 23$)	<i>Intercept</i>	12.86 ($p < .001$)	12.87 ($p < .001$)
	<i>95% CI</i>	[10.75, 14.98]	[11.82, 13.92]
	<i>Slope</i>	1.19 ($p = .001$)	1.10 ($p < .001$)
2 – “Moderate Stable Depression, Moderate Decreasing Alcohol Problems” ($n = 28$)	<i>95% CI</i>	[0.47, 1.91]	[0.64, 1.56]
	<i>Intercept</i>	8.84 ($p < .001$)	11.38 ($p < .001$)
	<i>95% CI</i>	[6.80, 10.88]	[10.20, 12.57]
3 – “Low-risk Normative” ($n = 313$)	<i>Slope</i>	0.28 ($p = .386$)	-2.80 ($p < .001$)
	<i>95% CI</i>	[-0.35, 0.91]	[-3.25, -2.34]
	<i>Intercept</i>	6.93 ($p < .001$)	1.26 ($p < .001$)
	<i>95% CI</i>	[6.39, 7.47]	[0.97, 1.56]
	<i>Slope</i>	0.11 ($p = .202$)	-0.19 ($p = .002$)
	<i>95% CI</i>	[-0.06, 0.29]	[-0.31, -0.07]

Class 1 = “Moderate Increasing Depression and Alcohol Problems”; Class 2 = “Moderate Stable Depression, Moderate Decreasing Alcohol Problems”; Class 3 = “Low-risk Normative” (mild stable depression, mild decreasing alcohol problems)

The bolded values signify the significant results

Fig. 1 Longitudinal Co-Trajectories of Depression and Alcohol Problems of the Three Latent Classes



membership in the depression-alcohol problems subgroups. These results are presented in Tables 3 and 4, which report each predictor’s odds ratio with 95% confidence intervals (CIs). A variable was considered to be a predictor of being in the “high-risk” (versus “low-risk”) depression-alcohol problems subgroup if the 95% CI for the odds ratio did not include 1.0.

In the first model (Table 3), we found that compared to the “low-risk normative” subgroup, the “moderate increasing depression and alcohol problems” subgroup endorsed significantly higher hopelessness and impulsivity, and the “moderate stable depression, moderate decreasing alcohol problems” subgroup endorsed significantly higher

impulsivity. The reference subgroup was then changed in order to determine if these personality traits differentiated the “moderate increasing depression and alcohol problems” subgroup from the “moderate stable depression, moderate decreasing alcohol problems” subgroup; we found that these traits did not relate to subgroup membership.

In the second model (Table 4), we found that compared to the “low-risk normative” subgroup, the “moderate increasing depression and alcohol problems” subgroup endorsed significantly higher boredom proneness and significantly lower general self-efficacy, and the “moderate stable depression, moderate decreasing alcohol problems”

Table 3 Multinomial Logistic Regression Models – Anxiety Sensitivity, Hopelessness, Impulsivity, Sensation Seeking, and External Stressors

Baseline Predictors:	<i>B</i>	<i>SE</i>	<i>p</i>	<i>OR</i>	<i>95% CI (OR)</i>
Model 1:					
“Moderate Increasing Depression and Alcohol Problems” (versus “Low-risk Normative”)					
Intercept	-9.85	2.11	<.001		
Sex (0 = male; 1 = female)	-0.41	0.51	.413	0.66	0.25, 1.78
Anxiety Sensitivity	0.02	0.10	.868	1.02	0.84, 1.23
Hopelessness	0.17	0.06	.004	1.19	1.06, 1.34
Impulsivity	0.44	0.11	<.001	1.55	1.26, 1.91
Sensation Seeking	-0.06	0.07	.391	0.94	0.82, 1.08
Income loss	0.92	0.48	.054	2.52	0.98, 6.45
Young child (< 18 years old)	-0.11	0.60	.851	0.89	0.28, 2.90
Living alone	-0.52	0.73	.481	0.60	0.14, 2.51
“Moderate Stable Depression, Moderate Decreasing Alcohol Problems” (versus “Low-risk Normative”)					
Intercept	-7.92	1.78	<.001		
Sex (0 = male; 1 = female)	-0.37	0.44	.399	0.69	0.29, 1.64
Anxiety Sensitivity	0.03	0.09	.699	1.03	0.88, 1.22
Hopelessness	0.10	0.06	.067	1.11	0.99, 1.24
Impulsivity	0.22	0.09	.014	1.25	1.05, 1.49
Sensation Seeking	0.07	0.06	.267	1.07	0.95, 1.21
Income loss	0.70	0.41	.093	2.01	0.89, 4.52
Young child (< 18 years old)	-0.13	0.53	.801	0.88	0.31, 2.47
Living alone	-0.47	0.67	.489	0.63	0.17, 2.35
“Moderate Increasing Depression and Alcohol Problems” (versus “Moderate Stable Depression, Moderate Decreasing Alcohol Problems”)					
Intercept	-1.93	2.51	.443		
Sex (0 = male; 1 = female)	-0.04	0.62	.948	0.96	0.28, 3.25
Anxiety Sensitivity	-0.02	0.12	.889	0.98	0.78, 1.24
Hopelessness	0.07	0.07	.317	1.08	0.93, 1.24
Impulsivity	0.22	0.13	.086	1.24	0.97, 1.59
Sensation Seeking	-0.13	0.09	.131	0.88	0.74, 1.04
Income loss	0.23	0.59	.700	1.26	0.39, 4.00
Young child (< 18 years old)	0.02	0.73	.977	1.02	0.24, 4.29
Living alone	-0.05	0.93	.957	0.95	0.15, 5.89

Statistically significant (*p* < .05) parameters are bolded

subgroup endorsed significantly higher boredom proneness. Compared to the “moderate stable depression, moderate decreasing alcohol problems” subgroup, the “moderate increasing depression and alcohol problems” subgroup endorsed significantly lower general self-efficacy.

Table 4 Multinomial Logistic Regression Models – Boredom Propensity, Self-Efficacy, and External Stressors

Baseline Predictors:	<i>B</i>	<i>SE</i>	<i>p</i>	<i>OR</i>	<i>95% CI (OR)</i>
Model 2:					
“Moderate Increasing Depression and Alcohol Problems” (versus “Low-risk Normative”)					
Intercept	-0.95	1.83	.606		
Sex (0 = male; 1 = female)	-0.33	0.48	.489	0.72	0.28, 1.84
Boredom Propensity	0.06	0.03	.032	1.06	1.01, 1.12
General Self-Efficacy	-0.13	0.05	.005	0.88	0.81, 0.96
Income loss	0.68	0.47	.154	1.97	0.78, 4.98
Young child (< 18 years old)	0.44	0.59	.461	1.55	0.49, 4.91
Living alone	-0.46	0.73	.523	0.63	0.15, 2.61
“Moderate Stable Depression, Moderate Decreasing Alcohol Problems” (versus “Low-risk Normative”)					
Intercept	-4.53	1.76	.010		
Sex (0 = male; 1 = female)	-0.48	0.43	.260	0.62	0.27, 1.42
Boredom Propensity	0.07	0.02	.002	1.07	1.03, 1.12
General Self-Efficacy	-0.00	0.05	.948	1.00	0.91, 1.09
Income loss	0.74	0.41	.071	2.10	0.94, 4.69
Young child (< 18 years old)	0.13	0.52	.804	1.14	0.41, 3.12
Living alone	-0.39	0.67	.556	0.68	0.18, 2.49
“Moderate Increasing Depression and Alcohol Problems” (versus “Moderate Stable Depression, Moderate Decreasing Alcohol Problems”)					
Intercept	3.58	2.37	.131		
Sex (0 = male; 1 = female)	0.15	0.61	.809	1.16	0.35, 3.80
Boredom Propensity	-0.01	0.03	.702	0.99	0.93, 1.05
General Self-Efficacy	-0.12	0.06	.035	0.88	0.79, 0.99
Income loss	-0.06	0.59	.915	0.94	0.29, 3.00
Young child (< 18 years old)	0.31	0.73	.676	1.36	0.32, 5.73
Living alone	-0.07	0.93	.938	0.93	0.15, 5.77

Statistically significant (*p* < .05) parameters are bolded

In both models (Tables 3 and 4), biological sex, experiencing a loss of income during the pandemic, being a parent who lives with a child under 18, and living alone were unrelated to depression-alcohol problems subgroup membership.

Coping strategies

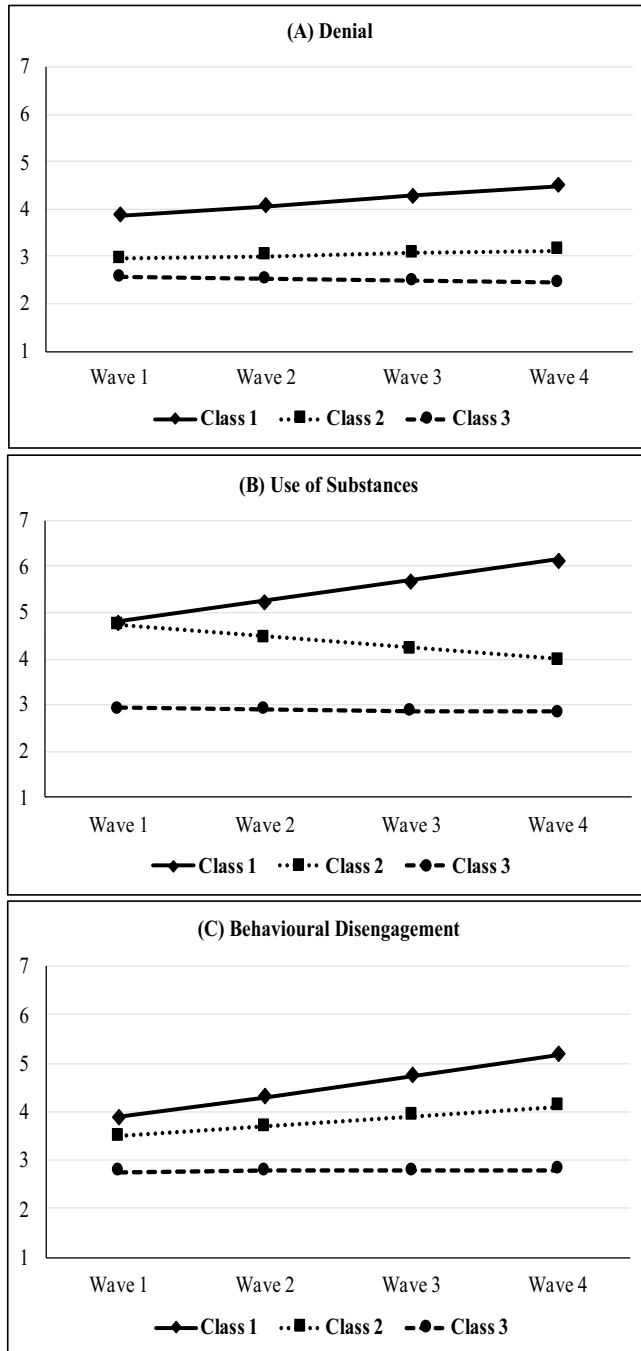
Finally, linear mixed models (LMMs) were conducted to explore differences between the depression-alcohol problems subgroups in avoidant, problem-focused, and emotion-focused coping over time during the pandemic. Time was coded as 0, 1, 2, and 3. Age, sex, and race were statistically controlled for. Table 5 and Fig. 2 present the parameter

Table 5 Linear Mixed Models – Coping Strategies

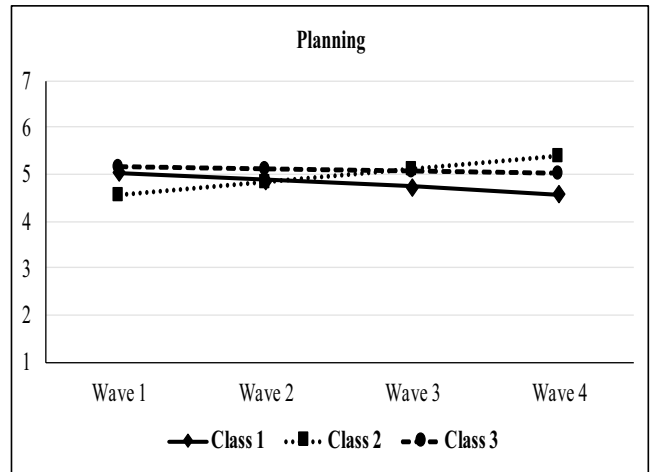
Predictors:	B	SE	t	p	95% CI
Model 1: Denial					
Intercept	3.86	0.24	16.38	<.001	3.3977, 4.3243
Time (coded as 0, 1, 2, and 3)	0.21	0.10	2.15	.032	.0184, .4005
Age	-0.00	0.00	-1.22	.222	-.0054, .0013
Sex (male=0; female=1)	0.06	0.09	0.68	.496	-.1153, .2370
Race (non-White=0; White=1)	-0.03	0.09	-0.27	.786	-.2116, .1600
D1 (Class 1=0; Class 3=1)	-1.30	0.23	-5.66	<.001	-1.7485, -.8472
D2 (Class 1=0; Class 2=1)	-0.92	0.30	-3.07	.002	-1.5035, -.3290
Time*D1	-0.25	0.10	-2.45	.015	-.4433, -.0484
Time*D2	-0.15	0.13	-1.16	.245	-.3949, .1004
Model 2: Substance Use					
Intercept	4.80	0.33	14.72	<.001	4.1622, 5.4456
Time (coded as 0, 1, 2, and 3)	0.45	0.12	3.86	<.001	.2185, .6732
Age	-0.00	0.00	-1.05	.294	-.0074, .0023
Sex (male=0; female=1)	0.14	0.13	1.03	.302	-.1242, .3988
Race (non-White=0; White=1)	0.12	0.14	0.88	.382	-.1525, .3979
D1 (Class 1=0; Class 3=1)	-1.86	0.31	-5.92	<.001	-2.4803, -1.2440
D2 (Class 1=0; Class 2=1)	-0.06	0.41	-0.16	.877	-.8687, .7414
Time*D1	-0.48	0.12	-4.01	<.001	-.7132, -.2438
Time*D2	-0.70	0.15	-4.68	<.001	-.9934, -.4050
Model 3: Behavioural Disengagement					
Intercept	3.88	0.23	16.68	<.001	3.4249, 4.3398
Time (coded as 0, 1, 2, and 3)	0.43	0.12	3.65	<.001	.1971, .6616
Age	-0.00	0.00	-1.61	.109	-.0064, .0006
Sex (male=0; female=1)	0.12	0.10	1.22	.224	-.0729, .3096
Race (non-White=0; White=1)	-0.10	0.10	-0.95	.341	-.2979, .1032
D1 (Class 1=0; Class 3=1)	-1.12	0.22	-5.00	<.001	-1.5545, -.6764
D2 (Class 1=0; Class 2=1)	-0.38	0.29	-1.32	.189	-.9555, .1888
Time*D1	-0.42	0.12	-3.44	<.001	-.6588, -.1790
Time*D2	-0.23	0.15	-1.49	.136	-.5284, .0734
Model 4: Planning					
Intercept	4.54	0.31	14.51	<.001	3.9204, 5.1410
Time (coded as 0, 1, 2, and 3)	0.28	0.12	2.32	.021	.0569, .5332
Age	-0.00	0.00	-1.74	.082	-.0102, .0006
Sex (male=0; female=1)	0.09	0.14	0.65	.519	-.1926, .3932
Race (non-White=0; White=1)	-0.09	0.15	-0.60	.550	-.3865, .2038
D1 (Class 2=0; Class 3=1)	0.60	0.30	2.02	.045	.0038, 1.2131
D2 (Class 2=0; Class 1=1)	0.50	0.43	1.17	.242	-.3322, 1.4577
Time*D1	-0.33	0.13	-2.59	.010	-.5858, -.0884
Time*D2	-0.44	0.19	-2.28	.023	-.8285, -.0700
Model 5: Emotional Supports					
Intercept	3.48	0.32	10.74	<.001	2.8452, 4.1193
Time (coded as 0, 1, 2, and 3)	0.25	0.11	2.24	.026	.0305, .4685
Age	-0.00	0.00	-0.87	.387	-.0078, .0030
Sex (male=0; female=1)	0.63	0.15	4.16	<.001	.3304, .9231
Race (non-White=0; White=1)	0.19	0.16	1.18	.241	-.1266, .4984
D1 (Class 2=0; Class 3=1)	0.75	0.31	2.43	.016	.1442, 1.3602
D2 (Class 2=0; Class 1=1)	0.48	0.44	1.09	.275	-.3838, 1.3480
Time*D1	-0.29	0.12	-2.50	.013	-.5214, -.0623
Time*D2	-0.11	0.17	-0.64	.523	-.4542, .2328

Class 1 = “Moderate Increasing Depression and Alcohol Problems”; Class 2 = “Moderate Stable Depression, Moderate Decreasing Alcohol Problems”; Class 3 = “Low-risk Normative” (mild stable depression, mild decreasing alcohol problems)

Avoidant Coping:



Problem-Focused Coping:



Emotion-Focused Coping:

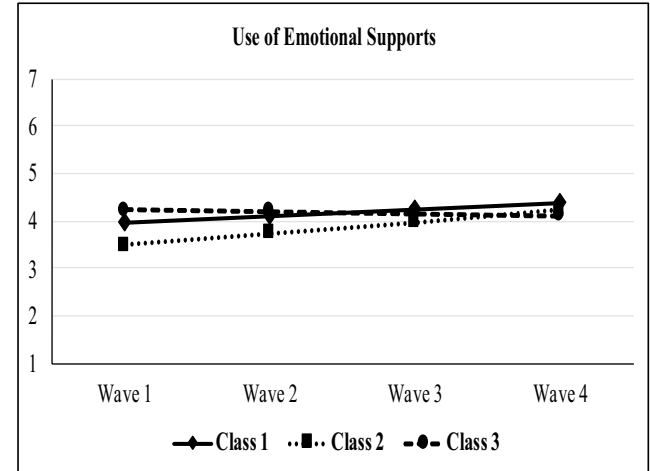


Fig. 2 Coping Strategies of the Three Latent Classes

estimates and graphs, respectively, of the Class by Time effects. Using Rights and Sterba’s (2019) Integrative Framework of R-Squared Measures, we report the proportion of total dependent variable variance explained by the LMM (i.e., by predictors through fixed slopes and random slope variation-covariation and by group outcome means through variations in the random intercept).

With respect to avoidant coping, there was a supported *Class* (Class 1 versus Class 3) by *Time* interaction in predicting the use of denial (Table 5, Model 1). Simple slopes analysis indicated that the “moderate increasing depression and alcohol problems” subgroup’s use of denial significantly increased over time ($B_{Class1} = 0.21, SE_{Class1} = 0.10, p_{Class1} = 0.032, 95\% CI_{Class1} [0.0184, 0.4005]$), whereas

denial did not significantly change over time among the “moderate stable depression, moderate decreasing alcohol problems” subgroup ($B_{\text{Class2}} = 0.06$, $SE_{\text{Class2}} = 0.08$, $p_{\text{Class2}} = 0.436$, 95% CI_{Class2} [-0.0953, 0.2198]) and the “low-risk normative” subgroup ($B_{\text{Class3}} = -0.04$, $SE_{\text{Class3}} = 0.03$, $p_{\text{Class3}} = 0.149$, 95% CI_{Class3} [-0.0857, 0.0129]). Similarly, there was a supported *Class* (Class 1 versus Class 2; Class 1 versus Class 3) by *Time* interaction in predicting the use of substances to cope (Table 5, Model 2). Simple slopes analysis found that the use of drugs and alcohol to cope significantly increased over time among the “moderate increasing depression and alcohol problems” subgroup ($B_{\text{Class1}} = 0.45$, $SE_{\text{Class1}} = 0.12$, $p_{\text{Class1}} < 0.001$, 95% CI_{Class1} [0.2185, 0.6732]), but significantly decreased over time among the “moderate stable depression, moderate decreasing alcohol problems” subgroup ($B_{\text{Class2}} = -0.25$, $SE_{\text{Class2}} = 0.09$, $p_{\text{Class2}} = 0.008$, 95% CI_{Class2} [-0.4395, -0.0672]). Among the “low-risk normative” subgroup, the use of substances to cope did not significantly change ($B_{\text{Class3}} = -0.03$, $SE_{\text{Class3}} = 0.03$, $p_{\text{Class3}} = 0.273$, 95% CI_{Class3} [-0.0910, 0.0257]). Finally, there was a supported *Class* (Class 1 versus Class 3) by *Time* interaction in predicting behavioural disengagement (Table 5, Model 3). Simple slopes analysis indicated that behavioural disengagement (i.e., giving up on attempts to cope) significantly increased over time among the “moderate increasing depression and alcohol problems” subgroup ($B_{\text{Class1}} = 0.43$, $SE_{\text{Class1}} = 0.12$, $p_{\text{Class1}} < 0.001$, 95% CI_{Class1} [0.1971, 0.6616]), but did not significantly change over time among the “moderate stable depression, moderate decreasing alcohol problems” subgroup ($B_{\text{Class2}} = 0.20$, $SE_{\text{Class2}} = 0.10$, $p_{\text{Class2}} = 0.059$, 95% CI_{Class2} [-0.0103, 0.3934]) and the “low-risk normative” subgroup ($B_{\text{Class3}} = 0.01$, $SE_{\text{Class3}} = 0.03$, $p_{\text{Class3}} = 0.729$, 95% CI_{Class3} [-0.0495, 0.0705]). The LMMs explained 53.9%, 66.1%, and 51.7% of the total outcome variance in the use of denial, drugs and alcohol, and behavioural disengagement, respectively.

With regards to problem-focused coping, there was a supported *Class* (Class 1 versus Class 2; Class 2 versus Class 3) by *Time* interaction in predicting planning (Table 5, Model 4). Simple slopes analysis found that the “moderate stable depression, moderate decreasing alcohol problems” subgroup’s use of planning significantly increased over time ($B_{\text{Class2}} = 0.28$, $SE_{\text{Class2}} = 0.12$, $p_{\text{Class2}} = 0.021$, 95% CI_{Class2} [0.0569, 0.5332]), but that the use of planning did not significantly change over time among the “moderate increasing depression and alcohol problems” subgroup ($B_{\text{Class1}} = -0.16$, $SE_{\text{Class1}} = 0.15$, $p_{\text{Class1}} = 0.298$, 95% CI_{Class1} [-0.4846, 0.1307]) and the “low-risk normative” subgroup ($B_{\text{Class3}} = -0.05$, $SE_{\text{Class3}} = 0.04$, $p_{\text{Class3}} = 0.210$, 95% CI_{Class3} [-0.1259, 0.0261]). The LMM explained 52.0% of the total outcome variance in the use of planning.

Finally, with respect to emotion-focused coping, the LMM supported a *Class* (Class 2 versus Class 3) by *Time* interaction in predicting the use of emotional supports (Table 5, Model 5). Simple slopes analysis indicated that the “moderate stable depression, moderate decreasing alcohol problems” subgroup’s use of emotional supports significantly increased over time ($B_{\text{Class2}} = 0.25$, $SE_{\text{Class2}} = 0.11$, $p_{\text{Class2}} = 0.026$, 95% CI_{Class2} [0.0305, 0.4685]), whereas this coping strategy did not significantly change over time among the “moderate increasing depression and alcohol problems” subgroup ($B_{\text{Class1}} = 0.14$, $SE_{\text{Class1}} = 0.13$, $p_{\text{Class1}} = 0.304$, 95% CI_{Class1} [-0.1258, 0.4033]) and the “low-risk normative” subgroup ($B_{\text{Class3}} = -0.04$, $SE_{\text{Class3}} = 0.03$, $p_{\text{Class3}} = 0.227$, 95% CI_{Class3} [-0.1111, 0.0264]). The LMM explained 66.0% of the total outcome variance in the use of emotional supports. The *Class* by *Time* interactions for predicting the other problem-focused and emotion-focused strategies were non-significant (see Table S3 in the Supplemental Materials).

Discussion

The current study analyzed longitudinal data, using parallel-process LCGA to determine whether distinct and meaningful classes of individuals would emerge based on their co-trajectories of depression and alcohol problems during the COVID-19 pandemic. Equally importantly, the current study sought out to determine the risk factors for combined increases in depression and alcohol problems among some people. To our knowledge, no prior study has examined person-centered changes in *both* depression and alcohol problems over a long period of time during the COVID-19 pandemic. Consistent with the above-reviewed studies, we found *multiple* classes of individuals based on their co-trajectories of depression and alcohol problems during the first nine months of the COVID-19 outbreak: a “low-risk normative” subgroup, a “moderate stable depression, moderate decreasing alcohol problems” subgroup, and a “moderate increasing depression and alcohol problems” subgroup. These patterns underscore the heterogeneity in the co-trajectory of depression and alcohol problems during the pandemic, highlighting that depression and alcohol problems remained stable and improved over time for some people, but both significantly worsened over time for other people. The multinomial logistic regressions and LMMs elucidate the risk factors for increasing depression and alcohol problems, which has significant clinical implications.

The “moderate increasing depression and alcohol problems” subgroup endorsed higher hopelessness and impulsivity than the “low-risk normative” subgroup. These results suggest depression characteristics (i.e., hopelessness) do not solely contribute to increasing depression and alcohol problems over time, and that impulsivity also plays a unique

role, which is consistent with prior work linking impulsivity to both mood- (e.g., Adams et al., 2019; Keough et al., 2016) and alcohol-related difficulties (e.g., Adams et al., 2019; Gonzalez et al., 2011). Further, the “moderate stable depression, moderate decreasing alcohol problems” subgroup also endorsed higher impulsivity than the “low-risk” subgroup, however this result may be accounted for the fact that people in this subgroup had *moderate initial* levels of alcohol problems (before these problems decreased over time). Overall, these results emphasize the need to further explore the way in which impulsivity exacerbates or maintains peoples’ mood- and alcohol-related difficulties.

The “moderate increasing depression and alcohol problems” subgroup also endorsed higher levels of boredom proneness and lower levels of general self-efficacy than the “low-risk normative” subgroup. Tying together boredom proneness and general self-efficacy is the tendency to focus on oneself, which, according to theories of objective self-awareness (Brockmeyer et al., 2015; Duval & Wicklund, 1972; Higgins, 1987; Steenbarger & Aderman, 1979), is associated with negative mood states (e.g., disappointment, dejection) and attempts to escape/avoid self-awareness. A person high in boredom proneness is persistently self-focused—aware of their lack of cognitive engagement but unable to articulate actionable desires and meaningfully engage with their external environment (Eastwood & Bambrah, 2021). Indeed, correlational studies suggest a positive relationship between trait indices of self-directed attention and boredom proneness (Eastwood et al., 2007; Gana et al., 2000; Harris, 2000; Seib & Vodanovich, 1998; von Gemmingen et al., 2003; as cited in Eastwood & Bambrah, 2021). Similarly, a person with low general self-efficacy is often focused on the discrepancy between what they would like to achieve and what they believe they are capable of achieving across contexts; and hence, they often avoid external circumstances/tasks perceived as difficult and they lack the ability to regulate their emotional states in such circumstances/tasks (Bandura, 1999, 2010; Carr, 2004). The current study’s results suggest that, compared to the “normative” subgroup, people with combined increases in depression and alcohol problems might be more self-focused—stuck on themselves, unable to transition into engaging, meaningful, and self-regulatory behaviour(s). Future studies could explicitly test whether or not the relationships of boredom proneness and general self-efficacy with combined increases in depression and alcohol problems involve self-directed attention.

Speaking to the findings from the LMMs, the “moderate increasing depression and alcohol problems” subgroup increasingly used avoidant coping over time, specifically the tendency to deny the reality of their problems, to use alcohol or other drugs to make themselves feel better and get through challenges, and to give up on attempts to cope with or address stressors. In tandem with the multinomial

logistic regressions, which found that low general self-efficacy uniquely distinguished the “moderate increasing depression and alcohol problems” subgroup from *both* of the other subgroups, these findings suggests that people with combined increases in depression and alcohol problems not only tend to possess little belief in their ability to manage difficult and stressful circumstances, but also coped with such circumstances during the pandemic in a manner where they avoided thinking about, feeling, and doing difficult things. Importantly, given the pandemic’s enduring nature, these findings are the first to suggest that worsening depression and alcohol problems among a subset of people was associated *longitudinally* with unhealthy cognitive (denial) and behavioural (substance use, behavioural disengagement) coping efforts to avoid stressors. In contrast, the “moderate stable depression, moderate decreasing alcohol problems” subgroup decreasingly used avoidant coping, specifically substances to cope. They also increasingly used adaptive problem-focused coping skills and emotion-focused coping skills over time, specifically the tendency to plan and strategize the steps to deal with stressors and to seek comfort and emotional support from others. Avoidant coping strategies are associated with a host of psychological difficulties, including depression and problematic alcohol use, whereas problem- and emotion-focused strategies that are characterized by *approaching* and addressing challenging circumstances and concomitant negative emotions are associated with indices of psychological well-being (e.g., Dias et al., 2012; Eisenberg et al., 2012; Poulus et al., 2020).

Notably, increases in depression and alcohol problems were more strongly related to internal factors (hopelessness, impulsivity, boredom proneness, general self-efficacy, and avoidant coping) than external stressors (experiencing a loss of income during the pandemic, being a parent who lives with a child under 18, living alone), which is consistent with the above-reviewed longitudinal findings by Baptist-Mohseni et al. (2022). Elucidating the personality and coping risk factors associated with combined increases in depression and alcohol problems not only underscores who may benefit most from prevention and intervention approaches during the pandemic (and other such stressful circumstances), but also areas of focus for these approaches. Indeed, psychoeducation and treatment approaches that target those prone to hopelessness, boredom, and low self-efficacy and that focus on reducing the use avoidant coping and enhancing the use of approach-oriented coping may help facilitate recovery and preparedness for those at risk for worsening depression alcohol problems. Specifically, approaches that emphasize the impeding nature of denial, substance use, and behavioural disengagement on one’s ability to cope with stressors in the long-term (e.g., that denying the reality of a stressor, using alcohol/drugs, and reducing one’s efforts

to address a stressor might permit the stressor to become more serious, making it more difficult to address; Carver et al., 1989), as well as emphasize how these avoidant coping strategies maintain hopelessness, the aversive feeling of being cognitively unengaged with one's surroundings, and the low belief in one's ability to achieve goals despite difficult circumstances, may be helpful in this regard. Cognitive behavioural therapy, which seeks to mitigate peoples' avoidance, escape, and other safety behaviours, as well as increase behavioural activation and engagement in meaningful and valued activities, may be an important avenue for intervention, as it has shown to be effective for people struggling with depression symptoms and problematic drinking (e.g., Pedrelli et al., 2013; Riper et al., 2014).

The use of a longitudinal design with four waves of assessment and with a large sample from across Canada, as well as the use of attention checks and validated measures to ensure accurate data on functioning, are key strengths of the current study. However, there are some shortcomings, including the lack of racial, provincial/territorial, and financial representation in our sample as the majority of participants identified as Caucasian, predominantly resided in Ontario, British Columbia, and Alberta, and had a high socioeconomic status. Additionally, we are precluded from extrapolating the current study's findings and the implications for treating pandemic-related depression and alcohol problems to individuals with severe psychopathology as we did not examine (and control for) pre-existing mental health diagnoses in our community sample. Moreover, there was 32.42% overall attrition across waves and this could have reduced statistical power in the analyses that involved our smaller classes, such as the multinomial logistic regressions. Relatedly, our sample size could be viewed as modest for latent class growth analyses. Larger sample sizes would permit the use of other analytical approaches (see Kamata et al., 2018 for a review of the *one-* and *three-step* approaches) that are less likely to yield biased estimates of the relationships of the latent classes with auxiliary variables (i.e., covariates, outcomes) because they retain the classification measurement errors across models. Future studies on pandemic-related depression and alcohol problems that reproduce our results in more diverse and larger groups of participants, as well as that have lower attrition rates, will be important.

In spite of these shortcomings, we obtained high classification probabilities in the current study, which, in tandem with prior findings (Frohlich et al., 2018; Orui et al., 2020), reinforces the distinct and meaningful classes of depression and alcohol problems that emerged in the data, as well as the risk factors for people with combined increases in depression and alcohol problems. As noted above, future studies should endeavour to understand exactly how features of personality (e.g., self-directed attention in boredom proneness and general self-efficacy) contribute to increased depression and

alcohol problems, so that interventions can be tailored more incisively. Furthermore, longitudinal studies that assess peoples' depression and alcohol problems beyond the waves that were examined in the current study will be critical in order to continue to understand the persisting and lingering effects of the pandemic on individuals' psychological functioning.

Supplementary Information The online version contains supplementary material available at <https://doi.org/10.1007/s12144-022-04109-4>.

Authors' Contributions All authors contributed to the study's conception and design. Material preparation and data collection were performed by Jeffrey D. Wardell and Matthew T. Keough. Data analyses were performed by Veerpal Bambah and Matthew T. Keough. The first draft of the manuscript was written and revised by Veerpal Bambah and all authors commented on previous versions of the manuscript. All authors read and approved the final manuscript.

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Data availability Subject to the ethical requirements of the York University Human Participants Research Committee and the Canadian Tri-Council policy statement on ethical conduct for research involving humans, all data, analysis code, and research materials underlying this study will be made available to researchers upon submission of an approved ethics protocol from their academic or research institution (or equivalent) to the corresponding author.

Declarations

Ethics approval Ethics approval for this study was obtained from the York University Human Participants Review Committee (Ethics certificate number: e2020-118).

Consent to participate Informed consent was obtained from all individual participants.

Conflicts of interest The authors have no conflicts of interests to disclose.

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