MiWaves MRT Analyses Results

1. Primary Aims Analysis (Part 1) with Proximal Cannabis Use Outcome

Frequencies of Baseline Covariates

Firstly, there are two baseline records for three IDs. For these three IDs, the second baseline record is retained. The two baseline submissions are likely due to the survey timing out and participants re-submitting.

1.) CANN_IMPORTANCE_BL: "Right now, how important is it to you to cut back your cannabis use?" Response: 0-10 likert scale: 0=Not at all, 10=Very

Other options:

- CANN_LIKELY_BL: "Right now, how likely are you to cut back your use of cannabis or cannabis products?" Response: scale of 0=Not at all 10=Very
- CANN_CONF_BL: "How confident are you that you could cut back your use of cannabis or cannabis products if you wanted to?" Response: scale of 0=Not at all 10=Very
- 2.) CANNHOURS_BL: "During the past month, how many hours, on an average day, did you use cannabis?" Response: Drop down selection 0-24
- 3.) CANNWAKE_BL: "During the past month, how soon did you typically use any cannabis products after you woke up for the day?" Response: 1=Within 5 minutes, 2=6-30 minutes, 3=31 minutes to almost 1 hour, 4=1 to almost 2 hours, 5=2 to almost 4 hours, 6=4 or more hours

Other options:

- CANNDAYS_BL: "How many days in the past month have you used cannabis?" Response: Drop down selection 0-31
- CANNMONTH_BL: "In the past month, how many times per day did you use cannabis?" Response: Drop down selection 0-24

Note: If $CANNDAYS_BL>0$, then displays $CANNHOURS_BL$, $CANNWAKE_BL$, $CANNMONTH_BL$, $CANN_IMPORTANCE_BL$, $CANN_LIKELY_BL$, and $CANN_CONF_BL$.

Table 1: Frequency of baseline variable cannabis importance (N = 120EAs)

cann_importance_bl	count	percent
0	4	3.3
1	5	4.2
2	13	10.8
3	11	9.2
4	18	15.0
5	24	20.0

6	13	10.8
7	17	14.2
8	11	9.2
9	1	0.8
10	3	2.5

Table 2: Frequency of baseline dichotomized variable high cannabis importance $\left(N=120EAs\right)$

high_cann_importance_bl	count	percent
0	51	42.5
1	69	57.5

Table 3: Frequency of baseline variable cannabis likely (N=120EAs)

cann_likely_bl	count	percent
0	5	4.2
1	15	12.5
2	15	12.5
3	19	15.8
4	13	10.8
5	22	18.3
6	12	10.0
7	7	5.8
8	9	7.5
9	1	0.8
10	2	1.7

Table 4: Frequency of baseline variable cannabis confidence (N = 120EAs)

cann_conf_bl	count	percent
0	3	2.5
1	1	0.8
2	4	3.3
3	10	8.3
4	14	11.7
5	17	14.2
6	13	10.8
7	17	14.2
8	21	17.5
9	9	7.5
10	11	9.2

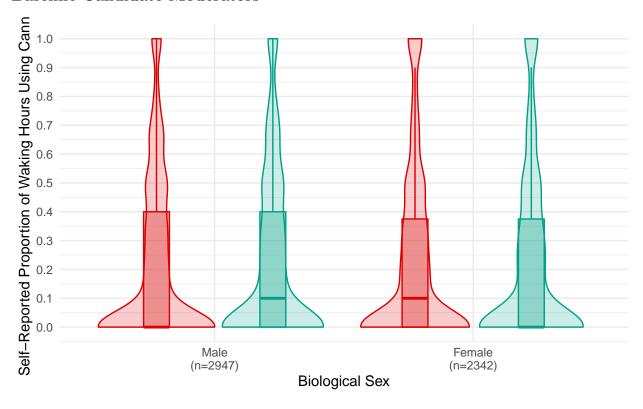
Table 5: Frequency of baseline variable cannabis hours (N=120EAs)

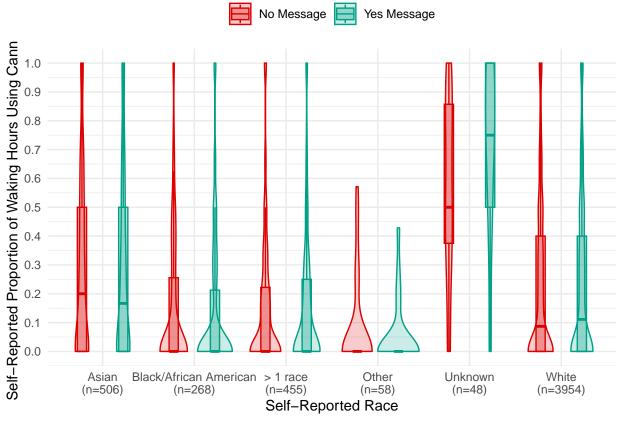
cannhours_bl	count	percent
0	2	1.7
1	16	13.3
2	34	28.3
3	28	23.3
4	13	10.8
5	8	6.7
6	8	6.7
7	1	0.8
8	2	1.7
10	2	1.7
11	1	0.8
13	1	0.8
14	1	0.8
17	1	0.8
24	2	1.7

Table 6: Frequency of baseline variable cannabis after waking (N=120EAs)

$cannwake_bl$	count	percent
0	10	8.3
1	4	3.3
3	18	15.0
4	6	5.0
5	13	10.8
6	69	57.5

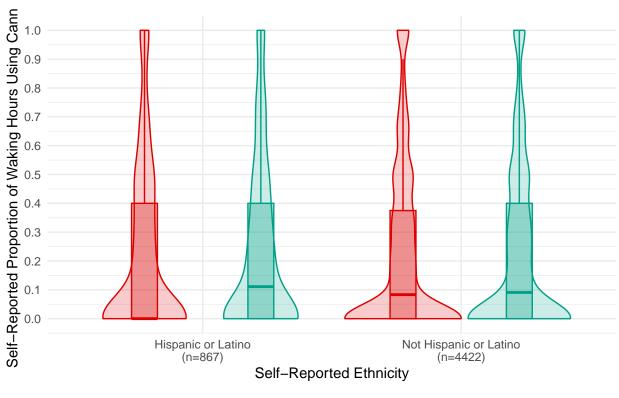
Distribution of Self-Reported Proportion of Waking Hours with Cannabis Use by Baseline Candidate Moderators



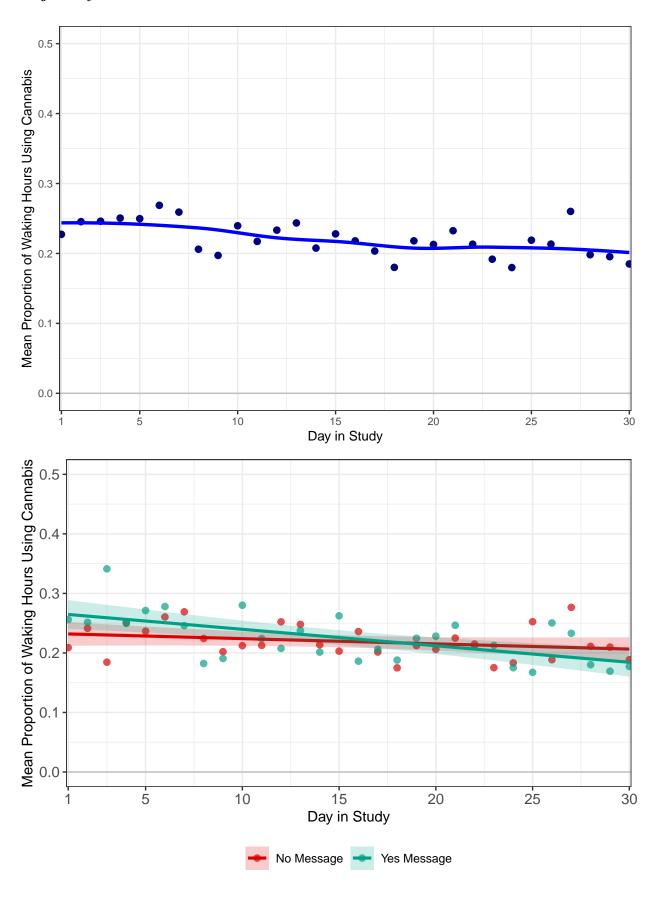


Yes Message

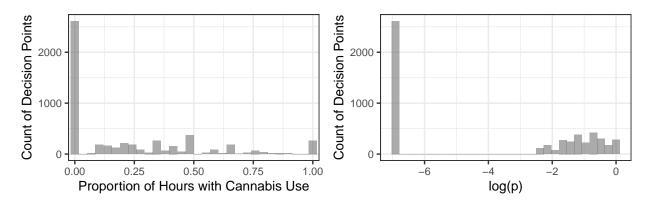
No Message

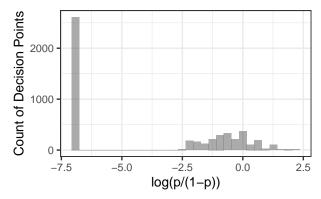


Trajectory of Cannabis Use Over Time



Overall Distribution of Proximal Outcome





Preliminary Causal Excursion Effect Estimates

Research Question 1: Examine whether, on average, there is a proximal effect of delivering an intervention message on proximal cannabis use

Proximal outcome $(Y_{i,t+1})$: Proportion of waking hours with self-reported cannabis use (0-1, treated as continuous)

Treatment indicator $(A_{i,t})$: Binary (1=Yes message, 0=No message)

Covariates:

- time of day binary (AM=0, PM=1),
- day of the week binary (weekday=1, weekend [Fri-Sun]=0),
- prior cannabis use proportion of waking hours averaged over past 4 decision points (i.e., approximately 48 hours),
- prior intervention engagement score that ranges from 0-3 averaged over past 6 decision points (i.e., approximately 72 hours),
- baseline motivation to change importance of cutting back cannabis use on a scale from 0 (Not at all) to 10 (Very) at time of baseline survey,
- baseline cannabis use self-reported average hours of cannabis use in prior day (range: 0-24), during the past month, and
- baseline time to cannabis use self-reported time to cannabis use, since awaking (1=Within 5 minutes, 2=6-30 minutes, 3=31 minutes to almost 1 hour, 4=1 to almost 2 hours, 5=2 to almost 4 hours, 6=4 or more hours), during the past month.

Research Question 2: Explore whether the effect of the intervention message on proximal cannabis use varies by each of the candidate moderators listed below.

Candidate Moderators:

- 1. timeofday: time of day binary (AM=0, PM=1),
- 2. interact_A_message: interaction type A message vs. no message binary (interaction type A message=1, no message=0),
- 2.2. interact_B_message: interaction type B message vs. no message binary (interaction type B message=1, no message=0),
- 2.3. interact_C_message: interaction type C message vs. no message binary (interaction type C message=1, no message=0),
- 3. prop_awakeuse_prior: prior cannabis use operationalized the same as the proximal outcome, at the prior decision point,
- 3.2. cov_prop_awakeuse_48hrs: prior cannabis use over the past 4 decision points,
- 4. wks_since_interv_start: time since under treatment (i.e., since intervention start) in weeks,
- 4.2 after_day15 [a]: time since under treatment (i.e., since intervention start) dichotomized (0=before and including day 15, 1=after day 15),
- 5. week_day_binary: day of the week binary (weekday=1, weekend [Fri-Sun]=0),

- 6. prior_interv_engag: prior intervention engagement operationalized the same as the proximal outcome, at the prior decision point,
- 6.2. cov_interv_engag_72hrs: over past 6 decision points,
- 7. prior_sent_message: prior delivery of a message binary (yes message=1, no message=0), at the prior decision point,
- 7.2. prior sent messages 48hrs: number of messages sent over past 4 decision points,
- 8. short message: short message vs. no message binary (short message=1, no message=0),
- 8.2. long_message: long message vs. no message binary (long message=1, no message=0),
- 9. male sex: baseline demographic of male biological sex (0=female,1=male),
- 10. white_race: baseline demographic of white race (0=not white, 1=white),
- 11. hispanic_ethn: baseline demographic of hispanic or latino ethnicity (0=not hispanic or latino, 1=hispanic or latino),
- 12. canndays_bl: baseline cannabis use severity that is the number of days used cannabis in past month (range: 0-31), which reflects cannabis use frequency,
 - 12.2. $dsmsc_tot_bl$: baseline cannabis use severity that is the count of number of symptoms endorsed (range: 0-11), which reflects diagnostic severity,
- 13. cann_importance_bl: baseline motivation to change that is the importance of cutting back cannabis use on a scale from 0 (Not at all) to 10 (Very) at time of baseline survey,
- 13.2. $high_cann_importance_bl$ [a]: baseline motivation to change binary (0=low motivation to change [score>=5]),
- 13.3. cann_likely_bl [a]: baseline likelihood to change that is how likely one is to cut back cannabis use on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey,
- 13.4. cann_conf_bl [a]: baseline confidence to change that is how confident one is to cut back cannabis use on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey,
- 13.5. cann_importance_bl [a]: baseline motivation to change on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey, when restricting to within week 1 (days 1-7),
- 13.6. cann_importance_bl [a]: baseline motivation to change on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey, when restricting to within week 2 (days 8-14),
- 13.7. cann_importance_bl [a]: baseline motivation to change on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey, when restricting to within week 3 (days 15-21),
- 13.8. cann_importance_bl [a]: baseline motivation to change on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey, when restricting to within week 4 onwards (days 22-30),
- 14. $phq2_tot_bl$: baseline mental health, which is the sum across two item scale PHQ-2 ("Over the last two weeks, how often have you been bothered by any of the following problems? Little interest or pleasure in doing things" and "Over the last two weeks, how often have you been bothered by any of the following problems? Feeling down, depressed, or hopeless" with response values of 0=Not at all, 1= Several days, 2=More than half the days, and 3=Nearly every day).

Notes: All continuous candidate moderators are mean-centered.

[a] This candidate moderator was included after the list of moderators was formalized.

Term	Estimate	95% LCL	95% UCL	StdErr	Wald	df1	df2	p-value
Main Effect Model (no covars)								
Intercept	0.009	-0.008	0.026	0.009	1.078	1	118	0.301
Main Effect Model (with covars)								
Intercept	0.009	-0.003	0.020	0.006	2.294	1	111	0.133
Moderation Effect Model 1								
Intercept	0.008	-0.005	0.022	0.007	1.531	1	110	0.219
timeofdayPM	0.001	-0.021	0.022	0.011	0.006	1	110	0.941
Moderation Effect Model 2	0.010	0.005	0.000	0.000	1 000	-1	444	0.150
Intercept	0.012	-0.005	0.029	0.009	1.833	1	111	0.178
Moderation Effect Model 2.2	0.011	0.000	0.007	0.000	1 (2) 4	1	111	0.004
Intercept	0.011	-0.006	0.027	0.008	1.634	1	111	0.204
Moderation Effect Model 2.3	0.001	0.015	0.010	0.000	0.020	1	111	0.000
Intercept	0.001	-0.015	0.018	0.008	0.030	1	111	0.862
Moderation Effect Model 3 Intercept	0.013	-0.008	0.033	0.010	1.501	1	110	0.223
prop_awakeuse_prior_c	0.013	-0.008	0.033	0.010	0.530	1	110	0.223 0.468
Moderation Effect Model 3.2	0.024	-0.041	0.000	0.000	0.000	1	110	0.400
Intercept	0.009	-0.003	0.020	0.006	2.332	1	110	0.130
cov_prop_awakeuse_48hrs_c	0.046	-0.017	0.109	0.032	2.071	1	110	0.153
Moderation Effect Model 4								
Intercept	0.015	-0.015	0.046	0.015	1.012	1	109	0.317
wks_since_interv_start	-0.003	-0.013	0.007	0.005	0.278	1	109	0.599
Moderation Effect Model 4.2								
Intercept	0.012	-0.005	0.029	0.009	1.890	1	109	0.172
after_day151	-0.007	-0.029	0.015	0.011	0.383	1	109	0.537
Moderation Effect Model 5								
Intercept	0.009	-0.010	0.028	0.010	0.971	1	110	0.327
week_day_binary1	-0.001	-0.028	0.025	0.013	0.008	1	110	0.929
Moderation Effect Model 6								
Intercept	0.008	-0.003	0.020	0.006	2.141	1	110	0.146
prior_interv_engag_c	-0.002	-0.018	0.014	0.008	0.066	1	110	0.798
Moderation Effect Model 6.2	0.000	0.000	0.000	0.000	2 20-		440	0.400
Intercept	0.009	-0.003	0.020	0.006	2.297	1	110	0.132
cov_interv_engag_72hrs_c	-0.012	-0.042	0.019	0.016	0.574	1	110	0.450
Moderation Effect Model 7 Intercept	0.008	-0.006	0.022	0.007	1.180	1	109	0.280
prior sent message	0.008	-0.000	0.022	0.007	0.006	1 1	109	0.280 0.937
Moderation Effect Model 7.2	0.001	-0.022	0.024	0.012	0.000	1	100	0.551
Intercept	0.006	-0.008	0.020	0.007	0.727	1	109	0.396
prior_sent_messages_48hrs_c	0.003	-0.011	0.020	0.007	0.189	1	109	0.665
Moderation Effect Model 8								
Intercept	0.009	-0.006	0.024	0.008	1.396	1	111	0.240
Moderation Effect Model 8.2								-
Intercept	0.008	-0.006	0.021	0.007	1.170	1	111	0.282
Moderation Effect Model 9								
Intercept	0.007	-0.006	0.021	0.007	1.082	1	109	0.301
male_sex1	0.004	-0.020	0.027	0.012	0.091	1	109	0.763
Moderation Effect Model 10								
Intercept	-0.002	-0.025	0.022	0.012	0.018	1	109	0.893
white_race1	0.014	-0.013	0.041	0.014	1.050	1	109	0.308

Term	Estimate	95% LCL	95% UCL	StdErr	Wald	df1	df2	p-value
Moderation Effect Model 11								
Intercept	0.007	-0.006	0.019	0.006	1.191	1	109	0.278
hispanic_ethn1	0.012	-0.019	0.043	0.016	0.552	1	109	0.459
Moderation Effect Model 12								
Intercept	0.009	-0.003	0.020	0.006	2.277	1	109	0.134
$canndays_bl_c$	0.000	-0.001	0.001	0.001	0.127	1	109	0.722
Moderation Effect Model 12.2								
Intercept	0.009	-0.003	0.020	0.006	2.321	1	109	0.131
$dsmsc_tot_bl_c$	0.002	-0.003	0.007	0.003	0.374	1	109	0.542
Moderation Effect Model 13								
Intercept	0.009	-0.003	0.020	0.006	2.360	1	110	0.127
cann_importance_bl_c	-0.005	-0.010	0.000	0.002	4.787	1	110	0.031
$Intercept + cann_importance_bl_c$	0.003	-0.007	0.014	0.006	0.309	2	110	0.735
Moderation Effect Model 13.2								
Intercept	0.019	0.001	0.037	0.009	4.152	1	109	0.044
high_cann_importance_bl1	-0.017	-0.040	0.006	0.012	2.228	1	109	0.138
Moderation Effect Model 13.3								
Intercept	0.009	-0.003	0.020	0.006	2.311	1	109	0.131
cann_likely_bl_c	-0.003	-0.008	0.001	0.002	1.898	1	109	0.171
Moderation Effect Model 13.4								
Intercept	0.009	-0.003	0.020	0.006	2.289	1	109	0.133
cann_conf_bl_c	0.001	-0.003	0.006	0.002	0.270	1	109	0.604
Moderation Effect Model 13.5								
Intercept	0.013	-0.013	0.040	0.013	0.979	1	109	0.325
cann_importance_bl_c	-0.008	-0.018	0.002	0.005	2.701	1	109	0.103
Intercept + cann_importance_bl_c	0.005	-0.021	0.030	0.015	0.115	2	109	0.891
Moderation Effect Model 13.6								
Intercept	0.005	-0.017	0.027	0.011	0.197	1	108	0.658
cann_importance_bl_c	-0.012	-0.022	-0.002	0.005	5.289	1	108	0.023
$Intercept + cann_importance_bl_c$	-0.007	-0.029	0.015	0.012	0.320	2	108	0.727
Moderation Effect Model 13.7								
Intercept	0.022	0.000	0.044	0.011	4.047	1	107	0.047
cann_importance_bl_c	0.001	-0.008	0.010	0.005	0.095	1	107	0.759
$Intercept + cann_importance_bl_c$	0.023	0.002	0.045	0.012	3.664	2	107	0.029
Moderation Effect Model 13.8								
Intercept	-0.002	-0.023	0.018	0.010	0.051	1	103	0.821
cann_importance_bl_c	-0.002	-0.010	0.006	0.004	0.200	1	103	0.656
Intercept + cann_importance_bl_c	-0.004	-0.024	0.016	0.011	0.131	2	103	0.877
Moderation Effect Model 14								
Intercept	0.009	-0.003	0.020	0.006	2.327	1	109	0.130
phq2_tot_bl_c	-0.001	-0.006	0.004	0.003	0.246	1	109	0.621

Notes: Standard errors are not yet adjusted to account for RL uncertainty. Moderation Effect Models include the covariates: time of day, day of week, prior cannabis use, prior intervention engagement, baseline motivation to change, baseline cannabis use, and baseline time to cannabis use.

2. Primary Aims Analysis (Part 2) with Proximal Intervention Engagement Outcome

Initial Diagnostics

First, let us examine the app_use_flag to see whether this solely captures browsing aside from EMA completion.

Table 8: Crosstabulation of app use flag variable and indicator for EMA completed (N=7038DPs)

app_use_flag_l	$completed_ema_l$	count	percent
FALSE	0	1404	19.9
FALSE	1	5060	71.9
TRUE	0	101	1.4
TRUE	1	473	6.7

Next, with the newly constructed engagement scores, let us examine the number of decision points where the engagement score changed values, from the old version to the new version. Note that the engagement score displayed below is the multi-category version ($engagement_multi$) and reflects proximal intervention engagement at t+1, i.e. following randomization at t.

Table 9: Crosstabulation of old engagement and new engagement scores (N=7038DPs)

1.1			
old engagement score	new engagement score	count	percent
0	0	578	8.2
0	1	104	1.5
0	2	522	7.4
0	3	206	2.9
1	0	24	0.3
1	1	6	0.1
1	2	44	0.6
1	3	25	0.4
2	0	65	0.9
2	1	515	7.3
2	2	3078	43.7
2	3	424	6.0
3	0	16	0.2
3	1	197	2.8
3	2	430	6.1
3	3	804	11.4

Table 10: Frequency of new engagement score (N = 7038DPs)

new engagement score	count	percent
0	683	9.7
1	822	11.7
2	4074	57.9
3	1459	20.7

Table 11: Frequency of old engagement score (N = 7038DPs)

count	percent
1410	20.0
99	1.4
4082	58.0
1447	20.6
	1410 99 4082

Preliminary Causal Excursion Effect Estimates

Research Question 3: Investigate whether, on average, there is a proximal effect of delivering an intervention message on proximal intervention engagement.

Proximal outcome $(Y_{i,t+1})$: Intervention engagement (discrete: 0-3, treated as continuous)

Treatment indicator $(A_{i,t})$: Binary (1=Yes message, 0=No message)

Covariates:

- time of day binary (AM=0, PM=1),
- day of the week binary (weekday=1, weekend [Fri-Sun]=0),
- prior intervention engagement score that ranges from 0-3 averaged over past 6 decision points (i.e., approximately 72 hours),
- prior human-touch engagement binary (1=yes, 0=no) for any email, text, or phone call made by study staff (after 72 hours, 120 hours, and 168 hours [1 week]) over past 4 decision points (i.e., approximately 48 hours).

Research Question 4: Explore whether the effect of the intervention message on proximal intervention engagement differs by each of the candidate moderators listed below.

Candidate Moderators:

- 1. timeofday: time of day binary (AM=0, PM=1),
- 2. prior_interv_engag: prior intervention engagement operationalized the same as the proximal outcome, at the prior decision point,
- 2.2. cov_interv_engag_72hrs: over past 6 decision points,
- 3. prop_awakeuse_prior: prior cannabis use operationalized the same as the proximal outcome, at the prior decision point,
- 3.2. cov_prop_awakeuse_48hrs: prior cannabis use over the past 4 decision points,
- 4. wks_since_interv_start: time since under treatment (i.e., since intervention start) in weeks,
- 5. week_day_binary: day of the week binary (weekday=1, weekend [Fri-Sun]=0),
- 6. prior_sent_message: prior delivery of a message binary (yes message=1, no message=0), at the prior decision point,
- 6.2. prior_sent_messages_48hrs: number of messages sent over past 4 decision points,
- 7. interact_A_message: interaction type A message vs. no message binary (interaction type A message=1, no message=0),

- 7.2. interact_B_message: interaction type B message vs. no message binary (interaction type B message=1, no message=0),
- 7.3. *interact_C_message*: interaction type C message vs. no message binary (interaction type C message=1, no message=0),
- 8. short message: short message vs. no message binary (short message=1, no message=0),
- 8.2. long_message: long message vs. no message binary (long message=1, no message=0),
- 9. cov_humtch_binary_48hrs: prior human-touch engagement binary (1=yes, 0=no) for any email, text, or phone call made by study staff (after 72 hours, 120 hours, and 168 hours [1 week]) over past 4 decision points (i.e., approximately 48 hours),
- 10. male_sex: baseline demographic of male biological sex (0=female,1=male),
- 11. white_race: baseline demographic of white race (0=not white, 1=white),
- 12. hispanic_ethn: baseline demographic of hispanic or latino ethnicity (0=not hispanic or latino, 1=hispanic or latino),
- 13. canndays_bl: baseline cannabis use severity that is the number of days used cannabis in past month (range: 0-31), which reflects cannabis use frequency,
- 13.2. dsmsc_tot_bl: baseline cannabis use severity that is the count of number of symptoms endorsed (range: 0-11), which reflects diagnostic severity,
- 14. cann_importance_bl: baseline motivation to change that is the importance of cutting back cannabis use on a scale from 0 (Not at all) to 10 (Very) at time of baseline survey, and
- 14.2. high_cann_importance_bl [a]: baseline motivation to change binary (0=low motivation to change [score<5], 1=high motivation to change [score>=5]),
- 14.3. cann_likely_bl [a]: baseline likelihood to change that is how likely one is to cut back cannabis use on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey,
- 14.4. cann_conf_bl [a]: baseline confidence to change that is how confident one is to cut back cannabis use on a scale from 0 (Not at all) to 10 (Very) at the time of baseline survey,
- 15. phq2_tot_bl: baseline mental health, which is the sum across two item scale PHQ-2 ("Over the last two weeks, how often have you been bothered by any of the following problems? Little interest or pleasure in doing things" and "Over the last two weeks, how often have you been bothered by any of the following problems? Feeling down, depressed, or hopeless" with response values of 0=Not at all, 1= Several days, 2=More than half the days, and 3=Nearly every day).

Notes: All continuous candidate moderators are mean-centered.

[a] This candidate moderator was included after the list of moderators was formalized.

Term	Estimate	$95\%~\mathrm{LCL}$	95% UCL	StdErr	Wald	df1	df2	p-value
Main Effect Model (no covars)								
Intercept	0.048	0.021	0.076	0.014	12.031	1	118	0.001
Main Effect Model (with covars) Intercept	0.045	0.019	0.072	0.013	11.548	1	114	0.001
Moderation Effect Model 1								
Intercept	0.069	0.039	0.099	0.015	20.220	1	113	0.000

$\stackrel{continued)}{=}$			01	a				
Term	Estimate	95% LCL	95% UCL	StdErr	Wald	df1	df2	p-value
time of day PM	-0.053	-0.092	-0.014	0.020	7.082	1	113	0.009
Moderation Effect Model 2								
Intercept	0.048	0.020	0.075	0.014	11.994	1	113	0.001
prior_interv_engag_c	0.021	-0.023	0.065	0.022	0.925	1	113	0.338
Moderation Effect Model 2.2								
Intercept	0.045	0.019	0.072	0.013	11.625	1	113	0.001
cov_interv_engag_72hrs_c	0.040	-0.032	0.111	0.036	1.191	1	113	0.277
Moderation Effect Model 3								
Intercept	0.037	0.008	0.067	0.015	6.362	1	112	0.013
prop_awakeuse_prior_c	-0.022	-0.105	0.060	0.042	0.282	1	112	0.596
Moderation Effect Model 3.2								
Intercept	0.045	0.019	0.072	0.013	11.491	1	112	0.001
cov_prop_awakeuse_48hrs_c	-0.034	-0.160	0.093	0.064	0.276	1	112	0.601
Moderation Effect Model 4								
Intercept	0.052	-0.007	0.111	0.030	3.101	1	112	0.081
$wks_since_interv_start$	-0.003	-0.023	0.017	0.010	0.074	1	112	0.787
Moderation Effect Model 5								
Intercept	0.057	0.021	0.092	0.018	10.193	1	113	0.002
week_day_binary1	-0.019	-0.070	0.033	0.026	0.511	1	113	0.476
Moderation Effect Model 6								
Intercept	0.008	-0.006	0.022	0.007	1.180	1	109	0.280
prior_sent_message	0.001	-0.022	0.024	0.012	0.006	1	109	0.937
Moderation Effect Model 6.2	0.054	0.005	0.000	0.044	4.4.00	_	440	0.000
Intercept	0.054	0.025	0.082	0.014	14.139	1	112	0.000
prior_sent_messages_48hrs_c	-0.014	-0.042	0.015	0.014	0.923	1	112	0.339
Moderation Effect Model 7	0.050	0.000	0.101	0.000	10.151	-	111	0.000
Intercept	0.079	0.036	0.121	0.022	13.171	1	114	0.000
Moderation Effect Model 7.2			0.0-4					
Intercept	0.038	0.003	0.074	0.018	4.520	1	114	0.036
Moderation Effect Model 7.3		0.010	0.040					
Intercept	0.015	-0.018	0.048	0.017	0.857	1	114	0.357
Moderation Effect Model 8								
Intercept	0.052	0.018	0.086	0.017	9.381	1	114	0.003
Moderation Effect Model 8.2								
Intercept	0.037	0.006	0.068	0.016	5.614	1	114	0.019
Moderation Effect Model 9								
Intercept	0.045	0.019	0.072	0.013	11.552	1	113	0.001
cov_humtch_binary_48hrs_c	-0.054	-0.404	0.296	0.177	0.093	1	113	0.761
Moderation Effect Model 10								
Intercept	0.040	0.005	0.075	0.018	5.184	1	112	0.025
$male_sex1$	0.012	-0.042	0.065	0.027	0.184	1	112	0.669
Moderation Effect Model 11								
Intercept	0.045	-0.010	0.101	0.028	2.638	1	112	0.107
white_race1	-0.002	-0.065	0.061	0.032	0.003	1	112	0.953
Moderation Effect Model 12					6 4			
Intercept	0.042	0.013	0.071	0.015	8.186	1	112	0.005
hispanic_ethn1	0.019	-0.053	0.091	0.036	0.271	1	112	0.604
Moderation Effect Model 13	_	_	_					_
Intercept	0.045	0.019	0.071	0.013	11.789	1	112	0.001

Term	Estimate	95% LCL	$95\%~\mathrm{UCL}$	StdErr	Wald	df1	df2	p-value
canndays_bl_c	0.002	-0.001	0.006	0.002	1.633	1	112	0.204
Moderation Effect Model 13.2								
Intercept	0.046	0.020	0.072	0.013	12.234	1	112	0.001
$dsmsc_tot_bl_c$	-0.008	-0.019	0.003	0.006	1.933	1	112	0.167
Moderation Effect Model 14								
Intercept	0.046	0.020	0.072	0.013	12.195	1	112	0.001
$cann_importance_bl_c$	0.008	-0.004	0.020	0.006	1.761	1	112	0.187
Moderation Effect Model 14.2								
Intercept	0.023	-0.011	0.058	0.017	1.821	1	112	0.180
high_cann_importance_bl1	0.039	-0.012	0.090	0.026	2.279	1	112	0.134
Moderation Effect Model 14.3								
Intercept	0.046	0.019	0.072	0.013	11.638	1	112	0.001
$cann_likely_bl_c$	0.000	-0.012	0.011	0.006	0.007	1	112	0.934
$Intercept + cann_likely_bl_c$	0.045	0.017	0.074	0.016	7.801	2	112	0.001
Moderation Effect Model 14.4								
Intercept	0.046	0.019	0.072	0.013	11.799	1	112	0.001
$cann_conf_bl_c$	-0.005	-0.015	0.005	0.005	1.027	1	112	0.313
$Intercept + cann_conf_bl_c$	0.040	0.015	0.066	0.015	7.533	2	112	0.001
Moderation Effect Model 15								
Intercept	0.046	0.020	0.073	0.013	12.448	1	112	0.001
$phq2_tot_bl_c$	0.008	-0.006	0.023	0.007	1.254	1	112	0.265

Notes: Standard errors are not yet adjusted to account for RL uncertainty. Model with covariates adjusts for the following: time of day, day of week, prior intervention engagement, and prior human-touch engagement.

Next, we examine the results for proximal intervention enagement when not restricting the sample to decision points with a completed ${\rm EMA}$.

Term	Estimate	95% LCL	95% UCL	StdErr	Wald	df1	df2	p-value
Main Effect Model (no covars)								
Intercept	0.021	-0.028	0.069	0.024	0.724	1	118	0.397
Main Effect Model (with covars) Intercept	0.028	-0.011	0.066	0.019	2.067	1	114	0.153
Moderation Effect Model 1								
Intercept	0.052	0.008	0.096	0.022	5.429	1	113	0.022
timeofdayPM	-0.048	-0.117	0.020	0.035	1.968	1	113	0.163
Moderation Effect Model 2	0.000	0.005	0.055	0.001	0.040	-1	110	0.004
Intercept prior_interv_engag_c	0.036 0.046	-0.005 -0.015	0.077 0.107	0.021 0.031	3.042 2.255	1 1	113 113	0.084 0.136
	0.040	-0.013	0.107	0.031	2.200	1	113	0.130
Moderation Effect Model 2.2 Intercept	0.028	-0.011	0.067	0.019	2.070	1	113	0.153
cov_interv_engag_72hrs_c	0.028	-0.011	0.007	0.019 0.035	0.882	1	113	0.155 0.350
Moderation Effect Model 3	0.002	0.000	0.101	0.000	0.002	_	110	0.000
Intercept	0.029	-0.011	0.068	0.020	2.059	1	112	0.154
prop_awakeuse_prior_c	0.028	-0.086	0.143	0.058	0.240	1	112	0.625
Moderation Effect Model 3.2								
Intercept	0.036	-0.001	0.074	0.019	3.746	1	112	0.055
cov_prop_awakeuse_48hrs_c	0.072	-0.105	0.250	0.090	0.650	1	112	0.422
Moderation Effect Model 4								
Intercept	0.077	0.001	0.154	0.039	4.024	1	112	0.047
$wks_since_interv_start$	-0.019	-0.046	0.008	0.014	1.957	1	112	0.165
Moderation Effect Model 5								
Intercept	0.033	-0.026	0.092	0.030	1.251	1	113	0.266
week_day_binary1	-0.009	-0.088	0.070	0.040	0.050	1	113	0.823
Moderation Effect Model 6								
Intercept	0.079	0.036	0.121	0.022	13.171	1	114	0.000
Moderation Effect Model 6.2								
Intercept	0.049	0.006	0.092	0.022	5.186	1	112	0.025
prior_sent_messages_48hrs_c	-0.038	-0.080	0.004	0.021	3.232	1	112	0.075
Moderation Effect Model 7	0.000	0.007	0.005	0.000	1 020	1	111	0.210
Intercept	0.029	-0.027	0.085	0.028	1.039	1	114	0.310
Moderation Effect Model 7.2 Intercept	0.022	-0.033	0.077	0.028	0.623	1	114	0.432
•	0.022	-0.033	0.077	0.028	0.023	1	114	0.432
Moderation Effect Model 7.3 Intercept	0.023	-0.026	0.073	0.025	0.850	1	114	0.359
•	0.023	-0.020	0.073	0.025	0.650	1	114	0.559
Moderation Effect Model 8 Intercept	0.029	-0.019	0.077	0.024	1.437	1	114	0.233
Moderation Effect Model 8.2	0.029	-0.019	0.011	0.024	1.491	1	114	0.200
Intercept	0.023	-0.020	0.065	0.021	1.137	1	114	0.289
Moderation Effect Model 9	0.023	-0.020	0.000	0.021	1.101	1	114	0.203
Intercept	0.028	-0.010	0.066	0.019	2.074	1	113	0.153
cov_humtch_binary_48hrs_c	0.028	-0.124	0.363	0.013	0.940	1	113	0.133
Moderation Effect Model 10	3.220	V	0.000	3		_		
Intercept	0.010	-0.040	0.060	0.025	0.161	1	112	0.689
male_sex1	0.039	-0.038	0.116	0.039	0.999	1	112	0.320
Moderation Effect Model 11								

Term	Estimate	95% LCL	95% UCL	StdErr	Wald	df1	df2	p-value
Intercept	0.036	-0.041	0.114	0.039	0.858	1	112	0.356
white_race1	-0.011	-0.101	0.078	0.045	0.064	1	112	0.801
Moderation Effect Model 12								
Intercept	0.018	-0.025	0.062	0.022	0.710	1	112	0.401
hispanic_ethn1	0.062	-0.023	0.146	0.043	2.070	1	112	0.153
Moderation Effect Model 13								
Intercept	0.028	-0.010	0.066	0.019	2.079	1	112	0.152
$canndays_bl_c$	0.003	-0.001	0.008	0.002	1.864	1	112	0.175
Moderation Effect Model 13.2								
Intercept	0.027	-0.011	0.065	0.019	2.029	1	112	0.157
$dsmsc_tot_bl_c$	-0.011	-0.026	0.004	0.008	2.096	1	112	0.150
$Intercept + dsmsc_tot_bl_c$	0.016	-0.023	0.055	0.022	0.550	2	112	0.579
Moderation Effect Model 14								
Intercept	0.028	-0.011	0.066	0.019	2.056	1	112	0.154
$cann_importance_bl_c$	-0.007	-0.025	0.010	0.009	0.681	1	112	0.411
Moderation Effect Model 14.2								
Intercept	0.034	-0.015	0.084	0.025	1.937	1	112	0.167
$high_cann_importance_bl1$	-0.011	-0.086	0.063	0.038	0.092	1	112	0.762
Moderation Effect Model 14.3								
Intercept	0.028	-0.009	0.065	0.019	2.231	1	112	0.138
$cann_likely_bl_c$	-0.022	-0.039	-0.005	0.009	6.640	1	112	0.011
$Intercept + cann_likely_bl_c$	0.006	-0.035	0.047	0.023	0.068	2	112	0.935
Moderation Effect Model 14.4								
Intercept	0.027	-0.011	0.066	0.019	2.035	1	112	0.156
$cann_conf_bl_c$	-0.006	-0.019	0.007	0.007	0.795	1	112	0.375
$Intercept + cann_conf_bl_c$	0.022	-0.016	0.059	0.021	1.028	2	112	0.361
Moderation Effect Model 15								
Intercept	0.028	-0.010	0.066	0.019	2.111	1	112	0.149
$phq2_tot_bl_c$	0.012	-0.010	0.034	0.011	1.166	1	112	0.282
$Intercept + phq2_tot_bl_c$	0.040	-0.003	0.083	0.024	2.687	2	112	0.072

Notes: Standard errors are not yet adjusted to account for RL uncertainty. Model with covariates adjusts for the following: time of day, day of week, prior intervention engagement, and prior human-touch engagement.

3. Draft Interpretation Sentences

Motivation Score Effect Moderation for Proximal Cannabis Use

The effect moderation model is specified as follows:

$$Y_{i,t+1}|Z = \beta_0 + \beta_1 I(A_{it} - 0.5) + \beta_2 M c_i + \beta_3 (I(A_{it} - 0.5) * M c_i) + \epsilon$$

where Z reflects the matrix of precision covariates and the action probabilities, denoted by A_{it} , are centered. We also grand mean center the candidate moderator of motivation score, denoted by Mc_i .

```
E(Y_{i,t+1}|Z, A_{it} = 0) = \beta_0 + \beta_2 M c_i + \epsilon

E(Y_{i,t+1}|Z, A_{it} = 1) = \beta_0 + \beta_1 + \beta_2 M c_i + \beta_3 M c_i + \epsilon

E(Y_{i,t+1}|Z, A_{it} = 1) = (\beta_0 + \beta_1) + (\beta_2 + \beta_3) M c_i + \epsilon
```

Next, let us examine the estimates and plug in the corresponding values to the simple slopes representation of the interaction effect.

```
##
## Call:
   wcls(data = df, id = "id", outcome = "prop_awakeuse", treatment = "actioni",
##
##
        rand_prob = "probi", moderator_formula = ~cann_importance_bl_c,
##
        control_formula = ~cov_prop_awakeuse_48hrs_c + cov_interv_engag_72hrs_c +
##
            timeofday + week_day_binary + cann_importance_bl_c +
##
            cannhours_bl_c + cannwake_bl_c)
##
## Coefficients:
##
                                 (Intercept)
                                                             cov_prop_awakeuse_48hrs_c
##
                                0.1815164788
                                                                            0.9582523084
##
                  cov_interv_engag_72hrs_c
                                                                             timeofdayPM
##
                               -0.0037127278
                                                                            0.0967171117
##
                           week_day_binary1
                                                                  cann_importance_bl_c
##
                               -0.0026982174
                                                                           -0.0004337124
##
                              cannhours_bl_c
                                                                           cannwake_bl_c
##
                                0.0005291601
                                                                           -0.0027282409
##
                           I(actioni - 0.5) cann_importance_bl_c:I(actioni - 0.5)
##
                                0.0086710413
                                                                           -0.0052313468
##
## Degrees of Freedom: 5289 Total (i.e. Null); Residual
##
## Scale is fixed.
##
## Correlation: Structure = independence
## Number of clusters:
                                   Maximum cluster size: 59
                            120
P(Y_{i,t+1}|Z, A_{it} = 0) = \hat{\beta}_0 + \hat{\beta}_2 M c_i
P(Y_{i,t+1}|Z, A_{it} = 0) = 0.1794963082 + -0.0004387147Mc_i
P(Y_{i,t+1}|Z, A_{it} = 1) = (\hat{\beta}_0 + \hat{\beta}_1) + (\hat{\beta}_2 + \hat{\beta}_3)Mc_i
P(Y_{i,t+1}|Z,A_{it}=1) = (0.1794963082 + 0.0086471773) + (-0.0004387147 + (-0.0052578510))Mc_i
P(Y_{i,t+1}|Z, A_{it} = 1) = (0.1881435) + (-0.005696566)Mc_i
```

Next, since motivation score is continuous to better understand the effect moderation, we plug in -1 SD, mean and +1 SD from the mean of motivation to change and compute the difference in slopes, that is, $\beta_2 + \beta_3 Mc$. The results are shown in plot below with three grey vertical lines at +/- 1 SD and the mean.

```
## $call
## wcls(data = df, id = "id", outcome = "prop_awakeuse", treatment = "actioni",
       rand prob = "probi", moderator formula = ~cann importance bl c,
       control_formula = ~cov_prop_awakeuse_48hrs_c + cov_interv_engag_72hrs_c +
##
##
           timeofday + week_day_binary + cann_importance_bl_c +
##
           cannhours_bl_c + cannwake_bl_c)
##
  $causal_excursion_effect
##
                                                 Estimate
                                                               95% LCL
##
                                              0.008671041 -0.002515706
## (Intercept)
                                            -0.005231347 -0.009969846
## cann_importance_bl_c
## (Intercept) - 2.28*cann_importance_bl_c
                                              0.020598512 0.007004376
## (Intercept) + 0.001*cann_importance_bl_c
                                             0.008665810 -0.001239052
## (Intercept) + 2.28*cann_importance_bl_c -0.003256429 -0.017199600
##
                                                   95% UCL
                                                                StdErr
                                                                            Wald
## (Intercept)
                                              0.0198577887 0.005644840 2.3596049
## cann_importance_bl_c
                                             -0.0004928478 0.002391050 4.7868505
## (Intercept) - 2.28*cann_importance_bl_c
                                              0.0341926487 0.007747463 7.0689136
## (Intercept) + 0.001*cann_importance_bl_c
                                              0.0185706721 0.005644901 2.3567075
## (Intercept) + 2.28*cann_importance_bl_c
                                              0.0106867408 0.007946382 0.1679363
##
                                            df1 df2
                                                         p-value
## (Intercept)
                                              1 110 0.127385179
## cann_importance_bl_c
                                               1 110 0.030793311
## (Intercept) - 2.28*cann_importance_bl_c
                                               2 110 0.001293809
## (Intercept) + 0.001*cann_importance_bl_c
                                               2 110 0.099498423
## (Intercept) + 2.28*cann_importance_bl_c
                                               2 110 0.845624048
```

