



Leveraging Smartphones for the Detection of and Scalable Interventions for Depressive Symptoms in Adolescents

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Depression in Adolescence



% of teens ages 12-17 who have had at least one major depressive episode in the past year, 2007-2017



Pew Research Center, 2019

see also Griffith et al., 2021; Bailen et al., 2019; Borghuis et al., 2017

Rumination

- Rumination (i.e., repetitive and negative self-focused thinking) is a transdiagnostic risk factor involved in the development of depression and anxiety (Aldao et al., 2010; Nolen-Hoeksema et al., 2008; Nolen-Hoeksema & Watkins, 2011; Watkins & Roberts, 2020).
- Rumination prospectively predicts the onset of depressive and anxiety symptoms in adolescents (e.g., McLaughlin & Nolen-Hoeksema, 2011; Muris et al., 2004; Gibb et al., 2012; Rood et al., 2009).
- Mindfulness training, with its emphasis on cultivating metacognitive awareness and present-moment attention, may be effective at reducing rumination in teens.



Youth Smartphone Ownership

FIGURE D. Smartphone ownership, by age, 2015 vs. 2019



Rideout & Robb, 2019

Table 1: Online health seekers

Among 14-22 year-olds, percent who have gone online for information about:

Fitness and exercise		
Diet and nutrition	52%	
Stress	44%	
Anxiety	42%	
Depression	39%	
Birth control	30%	
Pregnancy	28%	
Sleep disorders	27%	
Sexually transmitted diseases	26%	
Drug or alcohol abuse	24%	

Rideout & Fox, 2018

Mindfulness Apps

- Most frequently used apps for depression (Headspace; 5 million monthly active users) and anxiety (Calm: 9 million monthly active users) are mindfulness apps
- > 260 mindfulness apps available on the Apple App store and Google Play
- Headspace > 42 million users and 250k downloads per month
- 11% teens use mindfulness apps
- 38% of teens w/ elevated stress or depressive sxs have reported using mental health apps (18% used a mindfulness app)



Rideout & Robb, 2019; Wasil et al., 2020

App-Based Mindfulness Training for Ruminating Adolescents

- <u>Study 1</u>: Single-arm trial of mindfulness app (Hilt & Swords, 2021; Webb et al., 2021)
- <u>Study 2</u>: RCT of mindfulness app vs. mood monitoring app (Webb et al., 2022)
- Mindfulness app for 3 weeks w/ a higher likelihood of receiving mindfulness exercise (1-12 mins in duration) if report higher sadness/anxiety
- Mean # of mindfulness exercises completed over 21 days
 - <u>Study I</u>: 28.7 (avg I.4/day)
 - <u>Study 2</u>: 33.7 (avg 1.6/day)



"Immediate" vs. "Cumulative" Effects of Meditation on Rumination



Webb et al., Mindfulness, 2021

App-Based Mindfulness Training for Ruminating Adolescents

- Outcome is change in rumination (EMA)
 - Problem-focused and emotion-focused rumination
- <u>Study I (n = 80)</u>:
 - Significant reduction in prob-focused rumination (b = -.10, p < .001).
 - Significant reduction in emo-focused rumination (b = -.13, p < .001).



Hilt & Swords, 2021 Webb et al. 2021

Overall Outcomes

- <u>Study 2 (n = 152)</u>:
 - Group x Day interaction for prob-focused rumination (t = -3.22, p = .001).
 - Group x Day interaction for emo-focused rumination (t = -2.14, p = .03).



Webb et al., JCCP, 2022



Study I Predictive Model

Results of Elastic Net Regularization Models

	Immediate Effects of		Cumulative (3-week)		
	Mindfulness Exercises		Intervention Effects		
Predictors	KUMProb	KUIIIEmo	KUMProb	Kum _{Emo}	
Age	0.31				
Gender	1.29	1.50			
ERQ Suppression		-3.77	-0.42	-0.09	
ERQ Reappraisal		-0.92		0.19	
CRSQ Rumination		2.49		0.15	
CRSQ Distraction		2.56	-0.12	-0.38	
CRSQ Problem Solving			0.12	0.17	
FFMQ Observe	1.66				
FFMQ Describe				-0.20	
FFMQ Awareness		0.68			
FFMQ Nonjudgement			-0.28		
FFMQ Nonreactivity		-0.80	0.22		
CDI Total					
MASC Total				0.05	
History of NSSI					
History of Suicidal Ideation					
Prior mindfulness exposure					
R-Squared (RMSE)	.14 (10.9)	.22 (14.4)	.25 (1.3)	.17 (1.3)	

Webb et al., Mindfulness, 2021



:021 Webb et al., *JMIR,* 2022

Using Smartphones to Predict Negative Emotions



Ren et al., Psych Med, 2022

Study Design



Ren et al., Psych Med, 2022

Features Extracted



Ren et al., Psych Med, 2022

Results



		AUC			Accuracy		
Method	Anger	Sadness	Nervousness	Anger	Sadness	Nervousness	
GLMER	0.64	0.59	0.60	0.63	0.56	0.58	
PEM-ENet	0.69	0.61	0.67	0.67	0.59	0.63	
PEM-SVM	0.70	0.60	0.67	0.64	0.57	0.63	
PEM-RF	0.73	0.66	0.70	0.69	0.63	0.66	
PDEM	0.74	0.66	0.71	0.71	0.65	0.67	

Limitations & Future Directions

- I. Limited set of smartphone-derived variables
 - i. Other features may improve predictive performance (e.g., meta-data on calls/texts, teen social media use, recorded vocal tone/characteristics, peripheral measures of physiological arousal from wearable)
- 2. Sample size was small (although focused on within-person predictions)
 - i. 80 subjects w/ I year of data collection (w/ Justin Baker, MD, PhD)
- 3. Denser EMA sampling strategy (i.e., more surveys per day) would provide more granularity in assessments of affect fluctuations and a larger within-person dataset for modeling
 - i. But careful not to overburden teen participants (e.g., careless, random or stereotyped responding)

Thank You

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