

StudentSADD: Rapid Mobile Depression and Suicidal Ideation Screening of College Students during the Coronavirus Pandemic

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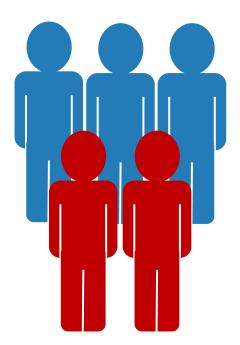
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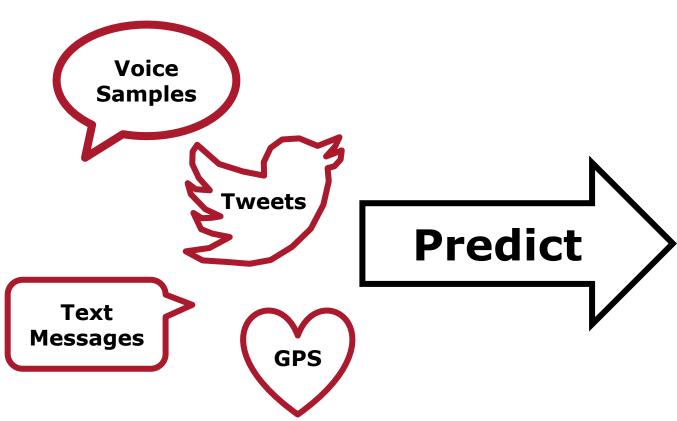
Mental Illnesses are Prevalent and Costly

2 in 5

U.S. **students** screened positive for depression in fall 2020 (National Healthy Minds Study)



Mental Illness Screening Research Goal



PHQ-9 For Depression Screening

Over the <i>last 2 weeks</i> , how often have you been bothered by any of the following problems?	Not at all	Several days	More than half the days	Nearly every day
1. Little interest or pleasure in doing things	0	1	2	3
2. Feeling down, depressed, or hopeless	0	1	2	3
3. Trouble falling or staying asleep, or sleeping too much	0	1	2	3
4. Feeling tired or having little energy	0	1	2	3
5. Poor appetite or overeating	0	1	2	3
 6. Feeling bad about yourself—or that you are a failure or have let yourself or your family down 	0	1	2	3
7. Trouble concentrating on things, such as reading the newspaper or watching television	0	1	2	3
8. Moving or speaking so slowly that other people could have noticed? Or the opposite—being so fidgety or restless that you have been moving around a lot more than usual	0	1	2	3
9. Thoughts that you would be better off dead or of hurting yourself in some way	0	1	2	3

K Kroenke, RL Spitzer, JB Williams, "The PHQ-9: validity of a brief depression severity measure", J of general internal medicine, 2001

Student Suicidal Ideation and Depression Detection (StudentSADD) Dataset

Goal: A dataset of active digital modalities from students for mental illness screening

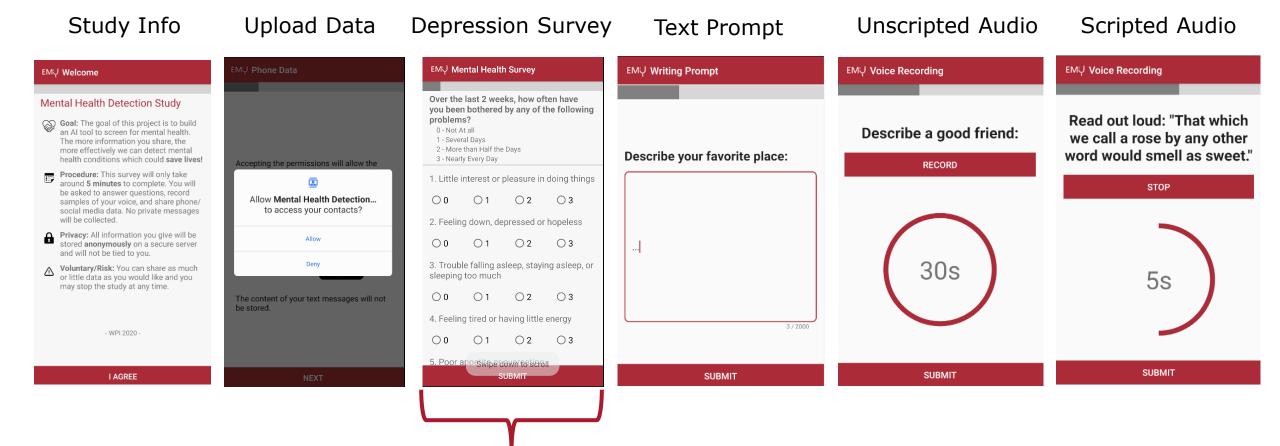
 Data: Typed replies, unscripted voice recordings, and scripted voice recordings

Population: 302 students from various universities

Collection: First fall semester impacted by COVID-19

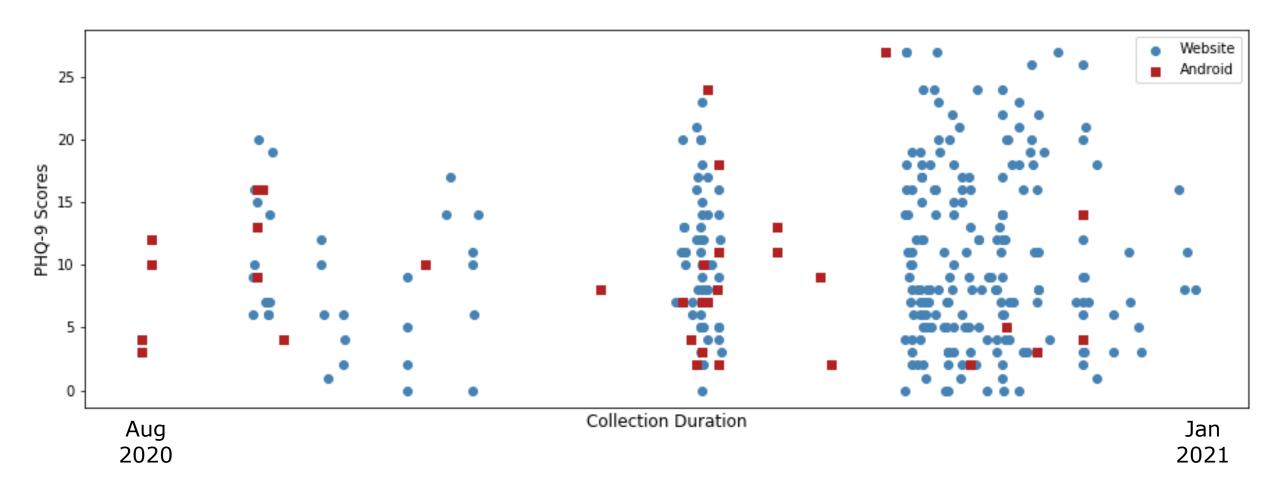
Analysis: Classification of audio and text modalities

StudentSADD Data Collection



Mental illness labels for the data

302 StudentSADD Participants



StudentSADD Data Shared

Almost half of the students screened positive for depression

	Shared	Depressed	Ideation
All	302 (100.0%)	142 (47.0%)	80 (26.5%)
Demographics	295 (97.7%)	139 (47 1%)	79 (26.8%)
Text Prompt	298 (98.7%)	141 (47.3%)	80 (26.9%)
Unscripted Audio	200 (66.2%)	90 (45.0%)	44 (22.0%)
Scripted Audio	194 (64.2%)	89 (45.9%)	44 (22.7%)
Unscripted Voice	110 (55.0%)	44 (40.0%)	22 (20.0%)
Scripted Voice	115 (59.3%)	45 (39.1%)	22 (19.1%)
Has Twitter	47 (27.5%)	21 (44.7%)	13 (27.7%)
Username	16 (34.0%)	5 (31.3%)	4(25.0%)
GPS	21 (63.6%)	7 (33.3%)	2(9.5%)
Calendar/Contacts	11 (33.3%)	5 (45.5%)	0(0.0%)
Call/Text Logs	10 (30.3%)	5 (50.0%)	0(0.0%)

Only 10 students shared call/text logs

Descriptive Analytics of StudentSADD Participants

		-
Website App Android App Computer Browser iPhone Browser Android Browser	269 (89.1%) 33 (10.9%) 86 (53.1%) 58 (35.8%) 18 (11.1%)	Most students completed the web version of the app
No Treatment Prior Treatment	217 (73.6%) 78 (26.4%)	A quarter of the students had prior depression treatment
Remote Hybrid Not Remote COVID-19 No/Unknown	97 (53.9%) 73 (40.6%) 10 (5.6%) 12 (6.3%) 168 (88.4%)	Remote students had the lowest rates of depression

StudentSADD Deep Learning Text Results

		D	epression	Suicidal Ideation		
Method	Data	F1	Balanced Accuracy	F1	Balanced Accuracy	
BERT	Text Reply	0.64 ± 0.01	0.52 ± 0.31	0.45 ± 0.02	0.56 ± 0.02	
BERT-LSTM	Text Reply	0.67 ± 0.01	0.57 ± 0.02	0.46 ± 0.05	0.56 ± 0.08	
BERT Attention	Text Reply	0.65 ± 0.01	0.53 ± 0.01	0.49 ± 0.00	0.62 ± 0.01	
BERT	Transcript	0.51 ± 0.01	0.55 ± 0.04	0.29 ± 0.07	0.58 ± 0.03	
BERT-LSTM	Transcript	0.50 ± 0.00	0.50 ± 0.00	0.32 ± 0.12	0.60 ± 0.06	
BERT Attention	Transcript	0.51 ± 0.01	0.51 ± 0.00	0.32 ± 0.01	0.59 ± 0.01	

StudentSADD Deep Learning Voice Results

		Γ	epression	Suicidal Ideation		
Method	Data	F1	Balanced Accuracy	F1	Balanced Accuracy	
VGGish	Unscripted Voice	0.56 ± 0.00	0.63 ± 0.01	0.59 ± 0.08	0.73 ± 0.06	
VGGish Attention	Unscripted Voice	0.63 ± 0.01	0.71 ± 0.01	0.46 ± 0.03	0.65 ± 0.01	
VGGish	Scripted Voice	0.56 ± 0.01	0.64 ± 0.01	0.44 ± 0.03	0.69 ± 0.04	
VGGish Attention	Scripted Voice	0.64 ± 0.02	0.74 ± 0.02	0.50 ± 0.14	0.68 ± 0.06	

StudentSADD Deep Learning Multimodal Results

		Depression		Suicidal Ideation		
Method	Data		F1 Bala	nced Accuracy	F1	Balanced Accuracy
BERT+VGGish	Unscripted	Transcript	0.54 ± 0.02	0.61 ± 0.05	0.39 ± 0.05	0.62 ± 0.02
AudiBERT	Unscripted	Transcript	0.52 ± 0.01	0.55 ± 0.03	0.37 ± 0.03	0.61 ± 0.02
BERT+VGGish	Scripted	Text Reply	0.64 ± 0.02	0.73 ± 0.02	0.20 ± 0.02	0.53 ± 0.02
AudiBERT	Scripted	Text Reply	0.69 ± 0.00	0.78 ± 0.00	0.37 ± 0.03	0.62 ± 0.02

Toto, Tlachac, Rundensteiner, "AudiBERT: A Deep Transfer Learning Multimodal Classification Framework for Depression Screening", CIKM, 2021

Takeaways

- 1. StudentSADD is a novel dataset collected during COVID-19
- 2. Students preferred the **website** app on a **computer** browser
- 3. Students preferred sharing the active screening modalities
- AudiBERT with scripted voice and text prompt was best at depression screening

StudentSADD Text & Feature Availability



StudentSADD

The Student Suicidal Ideation and Depression Detection (StudentSADD) dataset was collected between August 2020 and January 2021 by the 2020 REU team and 2020-2021 MQP team advised by ML Tlachac and Prof. Rundensteiner with assistance from Ermal Toto. The StudentSADD dataset contains text prompts, unscripted voice transcripts, unscripted voice openSMILE features, and scripted voice openSMILE features labeled with demographics and PHQ-9 depression screening scores from over 300 college student participants.

- Paper: https://dl.acm.org/doi/10.1145/3534604
- Data: StudentSADD data access
- Code: StudentSADD code for baseline models

To obtain access to the data please fill the StudentSADD data agreement form and email it to gr-studentsadd@wpi.edu using an email account affiliated with a higher education institution.

Thank You

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