PROJECT INTRODUCTION

OVERVIEW

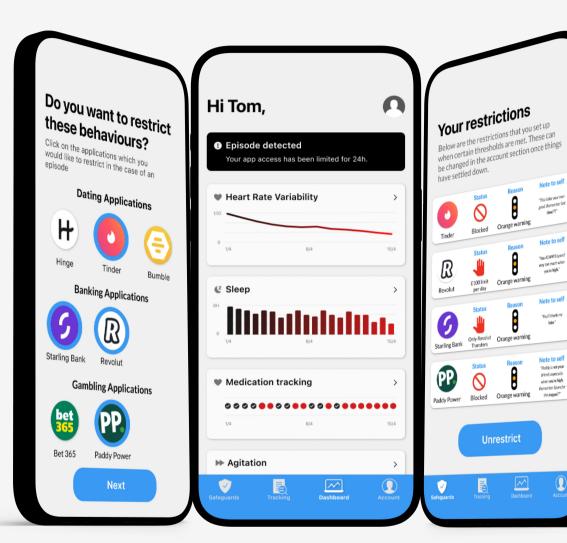
Moodpal is an application for individuals with Bipolar to help them deal with their condition in a manner that grants them autonomy over their episodes. It's an application that I designed for my Disability Interaction module during my masters degree.

THE PROBLEM

Individuals with Bipolar are often unaware when they are undergoing episodes, especially during periods of highs. As a result, they may engage in uncharacteristic risky behaviours, often with devastating consequences. While everyone has different manifestations, such behaviours may include gambling, excessive spending and hypersexuality to name but a few.

THE SOLUTION

Moodpal first identifies the individuals specific behaviours when they are in a high period. It monitors the individuals activities using machine learning techniques to detect when they are in the midst of an episode and alerts the individual to the likelihood that they are experiencing an episode, compensating for their lack of insight. It also restricts certain applications which are agreed on by the user in advance to prevent them from engaging in risky behaviour while in the midst of an episode.



Orange warning

Orange warning

MARKET RESEARCH

DEFINE PROBLEM

I began my project by reviewing applications on the market that cater to individuals with Bipolar Disorder. Along with this review and the latest in literature studies, I found that there was a major gap in the market for applications which used automatic sensing capabilities. Many studies were using this feature for different conditions, but all applications on the market were using manual tracking, which is cumbersome and time consuming.

Furthermore, most applications on the market were made with zero clinical input, which is a massive oversight considering the serious nature of the condition. As a result, my next step was to interview clinicians to see what they thought may be useful in the design of an application.







An example of manual tracking apps on the market

CrossCheck: Toward passive sensing and detection of mental health changes in people with schizophrenia

Rui Wang', Min S. H. Aung'', Saeed Abdullah', Rachel Brian, Andrew T. Campbell, Tanzeem Choudhury', Marta Hauser', John Kane', Michael Merrill', Emily A. Scherer, Vincent W. S. Tseng', and Dror Benz-Wede Comel University', Hofstra Northwell School of Medicine', (ruiwang. campbell')@cs.dartnouth.edu (musazez, junca)@northwell.edu (msaz42, smaz494, unzoem.choudhury, mam546, w1262)@cornell.edu (dror-benzeev, rachel-mbrian, emilya.scherre)@datmouth.edu

ABSTRACT

Early detection of mental bealth changes in individuals with serious mental illness is critical for effective intervention. CroxxCheck is the first sept towards the passive monitoring to the control of the 2-8.5 months. Our results indicate that there are statistically significant associations between automatically tracked behavioral features related to sleep, mobility, conversations, smartphone usage and self-reported indications of mental health indicates of the state of the st ble to train accurate personalized models that require fewer individual-specific data to quickly adapt to new users.

INTRODUCTION

Schizopheralis is a severe and complex poychiatric disorder that develops in approximately 1% of the world's population [49], Athough it is a chrosic condition, its symptom [49], Athough it is a chrosic condition, its symptom [40], Athough it is a chrosic condition, its symptom [40], and the property of to faster returns to remission [40].

LIBICOMP 16 SEPTEMBER 12-16 2016 HEIDELBERG GERMANN

Existing clinical practices are inefficient in detecting early precursors. Standard methods are based on face to face in-

StudentLife: Assessing Mental Health, Academic Performance and Behavioral Trends of College Students using Smartphones

Rui Wang[†], Fanglin Chen[†], Zhenyu Chen[†], Tianxing Li[†], Gabriella Harari[‡], Stefanie Tignor*, Xia Zhou*, Dror Ben-Zeev*, and Andrew T. Campbell†
Dartmouth College*, The University of Texas at Austin*, Northeastern University* {ruiwang, chentc, zhenyu, ltx, xia, campbell}@cs.dartmouth.edu gabriella.harari@utexas.edu, tignor.s@husky.neu.edu, dror.ben-zeev@dartmouth.edu

Much of the stress and strain of student life remains hidden. The StudentLife continuous sensing app assesses the day-to-day and week-by-week impact of workload on stress, sleep, day and week-by-week impact of workload on stress, sleep, activity, moot, sociability, mental well-being and academic performance of a single class of 48 students across a 10 week term at Dartmouth College using Android phones. Results from the StudentLife study show a number of significant corrections. relations between the automatic objective sensor data fron smartphones and mental health and educational outcomes of smarphones and metal relearn an educational outcomes of the student body. We also identify a Dartmouth term lifecycle in the data that shows students start the term with high pos-tive affect and conversation levels, low stress, and healthy sleep and daily activity patterns. As the term progresses and the workload increases, stress appreciably rises while posi-tive affect, sleep, conversation and activity drops off. The StudentLife dataset is publicly available on the web.

nartphones carried by students to find answers to some of these pressing questions

Consider students at Dartmouth College, an Ivy League col Consider students at Dartmouth College, an Ivy League col-lege in a small New England college town. Students typically take three classes over a 10-week term and live on campus. Dartmouth classes are generally demanding where student as-sessment is primarily based on class assignments, projects, midterms and final exams. Students live, work and socialize on a small self-contained campus representing a tightly knit community. The pace of the 10 week Dartmouth term is Knit community. The pace of the 10 week Dartmouth term is fast in comparison to a 15 week semester. The atmosphere among the students on campus seems to visibly change from a relaxed start of term, to an intense midterm and end of term. Typically classes at Dartmouth are small (e.g., 25-50). students), but introductory classes are larger (e.g., 100-170), making it difficult for a faculty to follow the engagement of performance of students on an individual level. Unless stu-

Some papers noting the promise of automatic or "passive" sensing

USER RESEARCH

INTERVIEWS WITH CLINICIANS

I conducted semi-structured interviews with two clinicians. The results from which are summarised on the right.

Some examples of questions I asked were as follows'

- Can you tell me a little bit Bipolar broadly?
- What, are the biggest challenges that people with bipolar face generally?
- Does every case of bipolar present the same or are their differences in symptoms?
- What are the most important things for people to keep well with Bipolar?

After performing both interviews, I used a bottom up thematic analysis to distill key intersections of both interviews.

THEMATIC ANALYSIS RESULTS

"It's a significant issue for people if they don't perceive that there is anything wrong with them."

Lack of insight on episodes

Sleep, exercise & diet crucial

Mechanism for alerting team or support

Objective markers to compensate for insight

"Metric points like HRV that are concurrently going up that shouldn't be influencing each other in clear sense."

"If there was a way to link someone's banking app card out that would be extremely useful."

Safeguard against reckless behaviours

Support network is essential

Medication tracking & adherence is critical

Notification when limits are met

"Maybe yellow is like we're approaching danger take discretion, orange is we're heading into the danger zone red is stop."

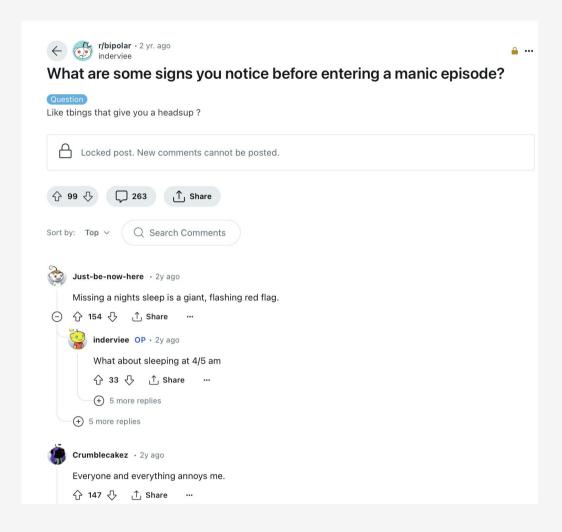
USER RESEARCH

DIGITAL ETHNOGRAPHY

To better understand user needs, I conducted a digital ethnography study on the Bipolar forum on Reddit, analysing the top 10 posts describing episodes of highs and the top 10 comments thereof.

I then performed a bottom up thematic analysis, the results of which are summarised below:

- Destructive, impulsive behaviours previously mentioned were common experiences for people during mania.
- The biggest triggers for episodes were generally not taking medication, stress and lack of sleep.
- Deep regret, shame and lack of control were common amongst those who had experienced major episodes.



An example post describing the precipitation of an episode

PROBLEM DEFINITION

USER JOURNEY MAP

From a combination of user research and market research I was able to create a user journey map for a fictional individual with Bipolar disorder, charting their journey from diagnosis to acceptance.

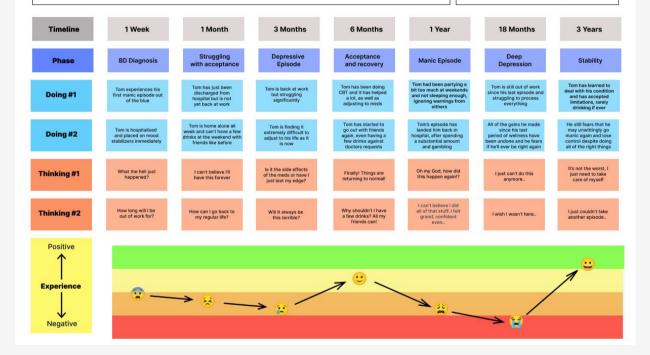
PRODUCT REQUIREMENTS

The user journey map helped crystalise the main requirements for the product. It should do as follows:

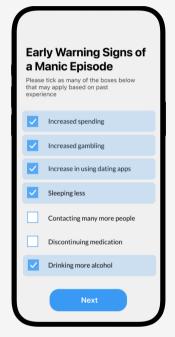
- Automatically detect manic episodes through objective phone and wearable sensing data.
- Include personal early warning signs to improve the system's detection capabilities.
- Alert the user in the case of a detected episode and suggest actions to take.
- Allow the user to alert others in the case of an episode for support.
- Take protective actions to prevent reckless behaviour during a manic episode.

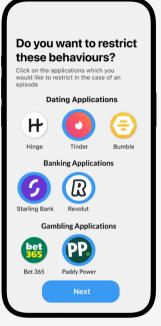
Scenario: Tom is a 24 year old young man who has been newly diagnosed with Bipolar Disorder, Type 1. The diagnosis has turned his life upside down, effecting his ability to work, socialise and just generally live as before. The user journey describes his process of acceptance and recovery, with reoccurring episodes and learnings along the way.

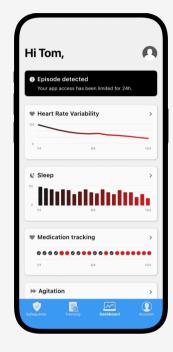
Goal: Tom wants to manage his condition to the best of his ability and live a normal life, but fears losing control again with another manic episode.



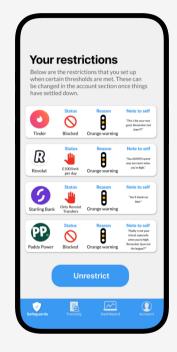
FINAL PRODUCT













Early Warnings Signs

This screen allows the individual warning signs so that the application knows what to look out for.

Restrictions

This screen allows the user to input their own user to restrict certain applications once thresholds of behaviour are met.

Dashboard 1 + 2

This dashboard shows the user trends in their behavioural data that are automatically sensed from their phone and wearable device along with an episode detector which predicts the likelihood of an episode occurring.

Safeguards

This section shows the user which applications have been restricted with a note placed by the user on the account to remind them why. It features a traffic light system as suggested during interviews.

Notification

This screen shows a notification alerting the user to an episode and letting them know that a trusted contact has been notified along with further restrictions.