



Making Digital Mental Health Work in the Real World

David C. Mohr, Ph.D.
Northwestern University
Feinberg School of Medicine

Disclosures

Honoraria/Consulting fees

Apple Inc.
Otsuka Pharmaceuticals
Pear Therapeutics
One Mind Foundation

Ownership Interest

Adaptive Health, Inc

Funding

National Institute of Mental Health

P50 MH119029

R01 MH100482

R01 MH109496

R01 MH111610

R44 MH114725



Center for Behavioral Intervention Technologies (CBITs)

- 6 Core faculty (Psychology, Human-Computer Interaction, Communications, Statistics/Data Science)
- Many Digital Mental Health projects in
 - University Settings
 - Eating Disorders
 - At-Risk Youth in Community Mental Health
 - Sensing in Mental Health
 - NIMH P50 ALACRITY Center: projects in Primary Care, OBGYN, and Geriatrics
- NIMH T32 Postdoctoral Training in Postdoctoral Fellowship

Outline

Problem Introduction

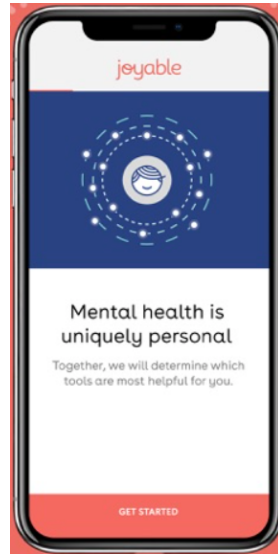
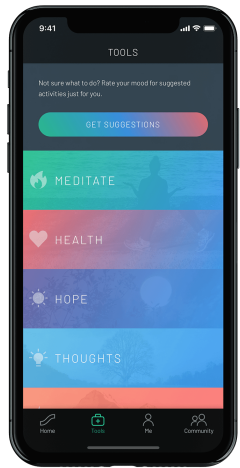
IntelliCare

Efficacy & Engagement Data

Implementation

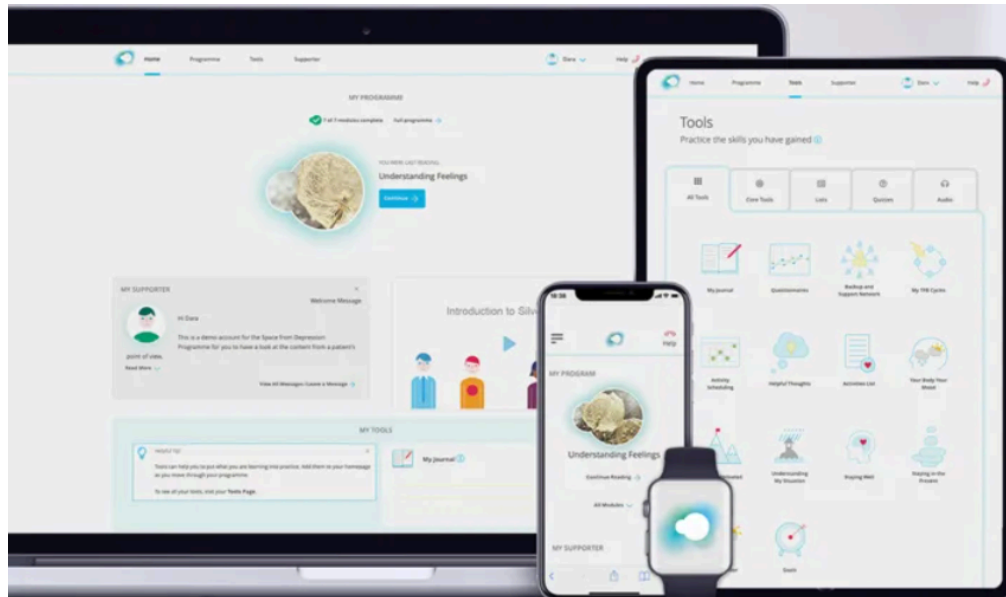
Implications for Digital Mental Health Research

Summary



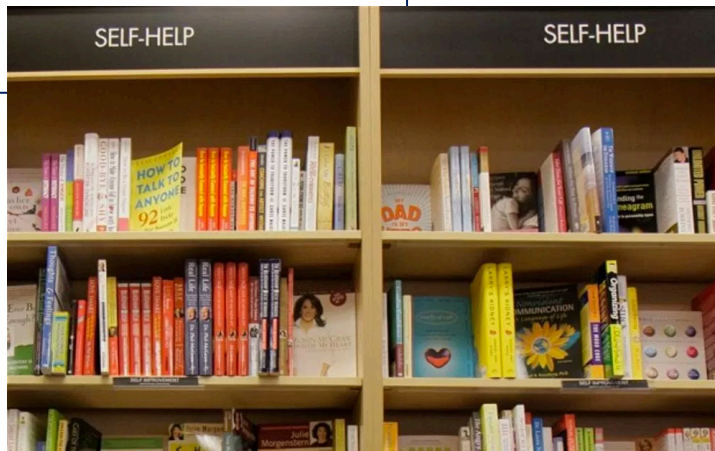
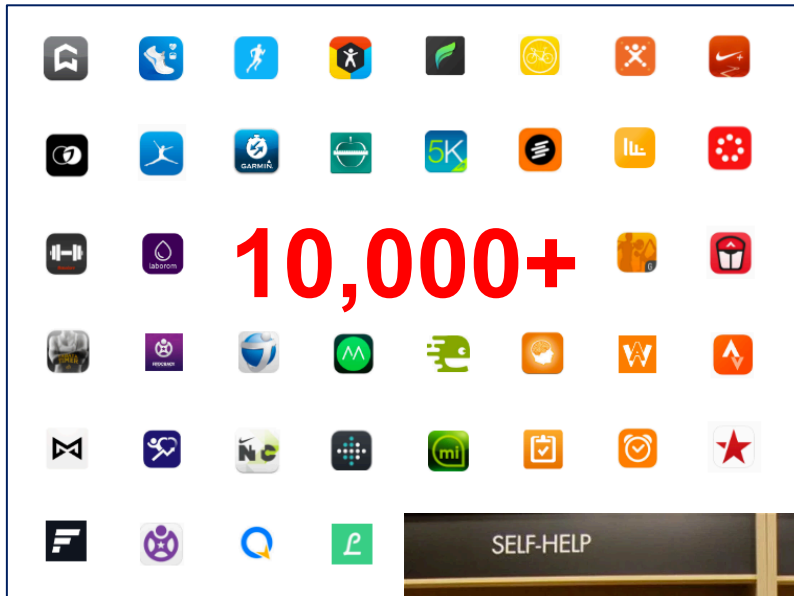
What is Digital Mental Health?

Uses commonly available technologies, such as apps or internet sites, in the treatment or management of mental illness

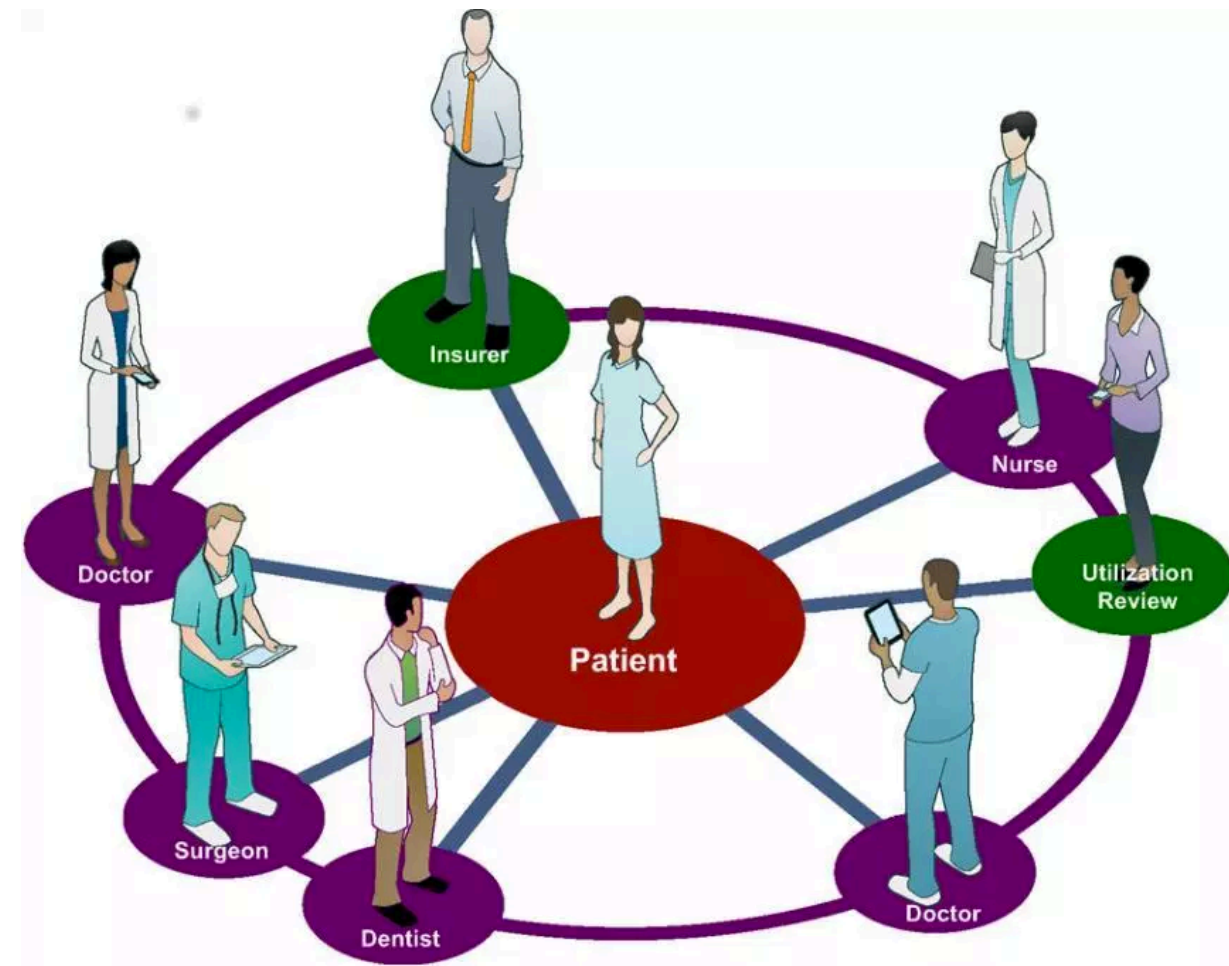


Digital Mental Health

Self Help



Technology-Enabled Services



How is Digital Mental Health Doing?

Lots of Validation

100+ of RCTs show

- Self-directed digital treatment treatments at best modestly effective
- Coach-supported digital interventions
 - Appear to be as effective as standard treatments ($d=.56 - 1.08$)
 - Coaches do not have to be mental health professionals

Karyotaki E, et al. *JAMA Psych*. 2017;74(4):351-359

Richards D, et al. *Clin Psychol Rev* 2012;32:329-342

Alvarez-Jimenez, et al. *Schizophr Res*. 2014;156:96-106

Real World Implementation



Mohr, et al., *Psychiatr Serv*, 2017;68:427-9

Contributors to Research-to-Practice Gap

Failure in Design

Failure to design for the patient

Failure to design for the provider

Failure to design for the organization

Design for the Patient

Design based on psychological and medical models



Design that embeds in the fabric of people's lives



Human Computer Interaction & Design

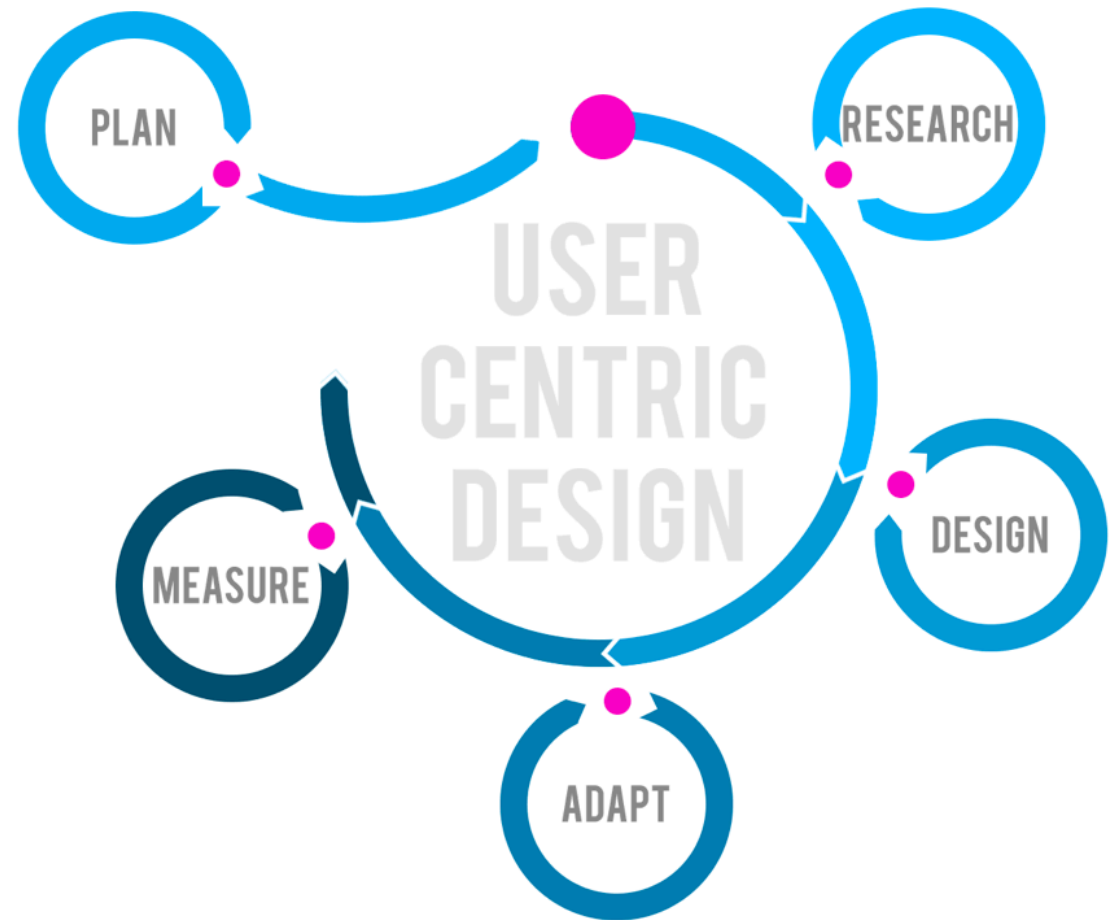
Giving a Voice to End Users

Stakeholders

- Patients
- Clinicians
- Administrative staff
- Leadership

Methods

- Interviews
- Focus groups
- Workflow observations
- Codesign workshops
- Usability testing



Design for Patients

How do people use tech for mental health?

Meng... Mohr, JMIR. 2018;20:e10139

Burgess, CHI, 2019

Eschler. CHI, 2020

- Most have tried mental health apps, but abandoned quickly
- They use technologies in bespoke ways
- Apps
 - Information
 - Mood tracking alone is not helpful
 - Behavior tracking – comparing 2 targets
 - Using non-mental health apps for mental health purposes: e.g. menstrual tracking, meditation
 - Daily tips: inspirational, encouragement, reminders
- **Apps generally have a single purpose**
- **Apps must easy, intuitive, and quick (usually ~30 sec.)**
- **Apps must give something of value back to the user (principle of reciprocity)**

Example: Transportation Needs



Mapping



Chicago Taxi App

Taxi



Parking



Ride Share



Car Rental



Public
Transportation



Airlines

Challenge for Mobile Mental Health

[illegible]

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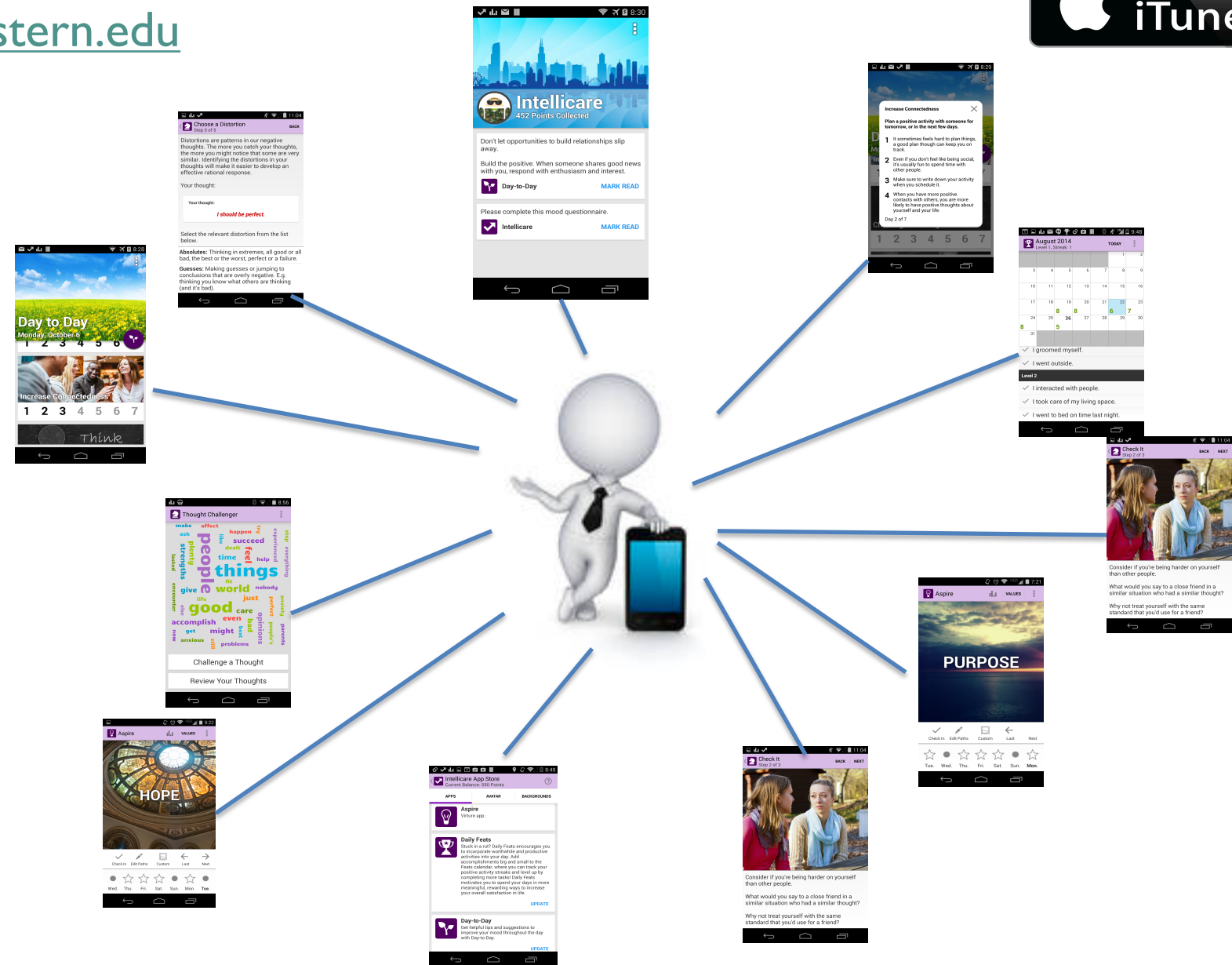
Summary

IntelliCare Platform

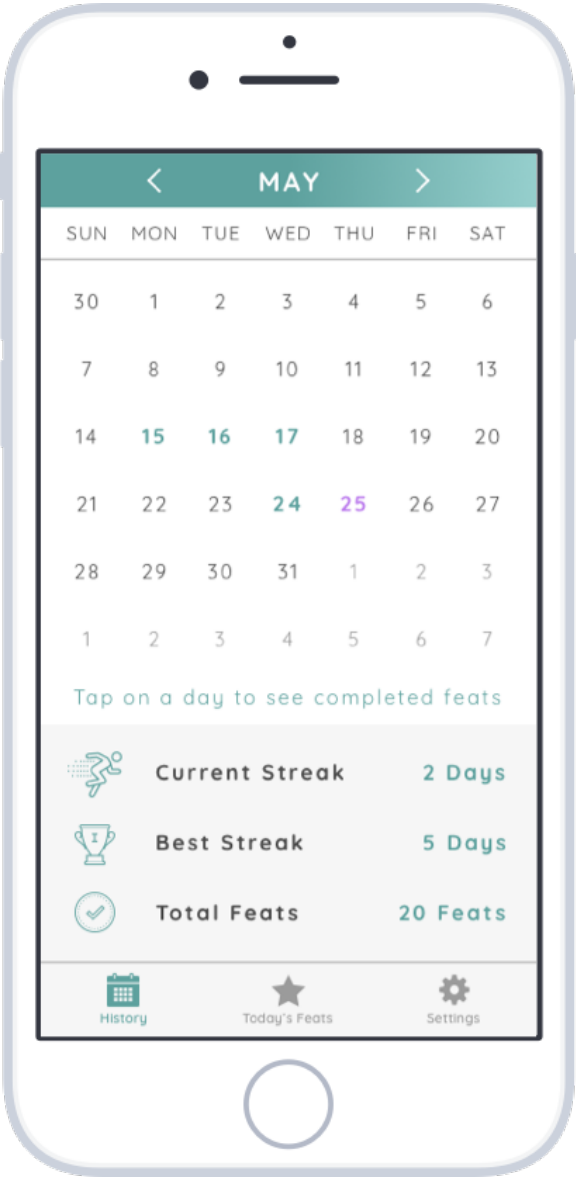
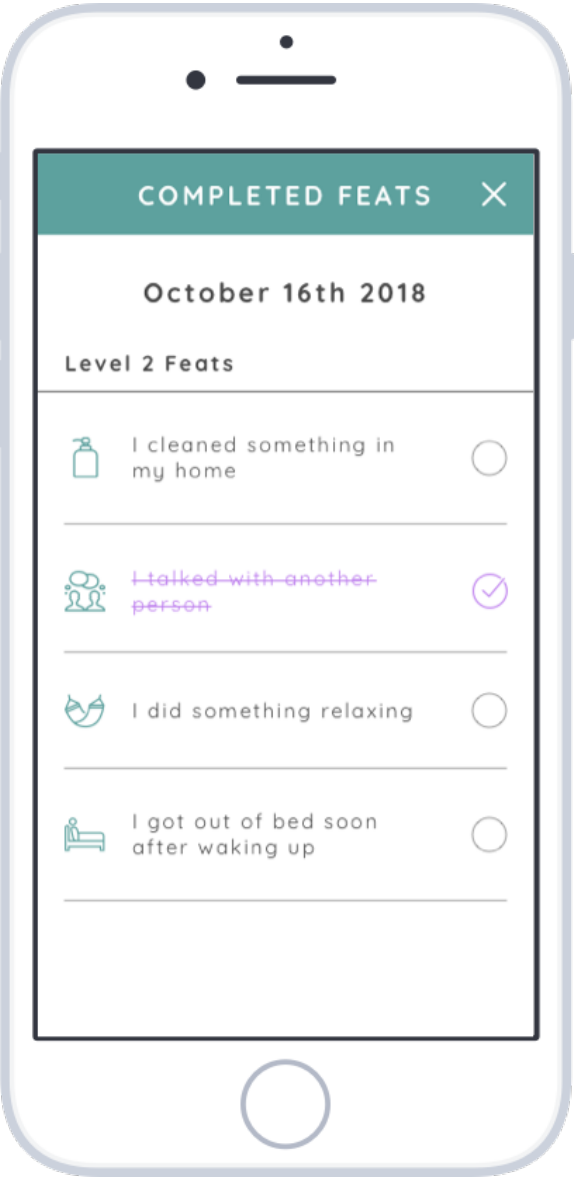
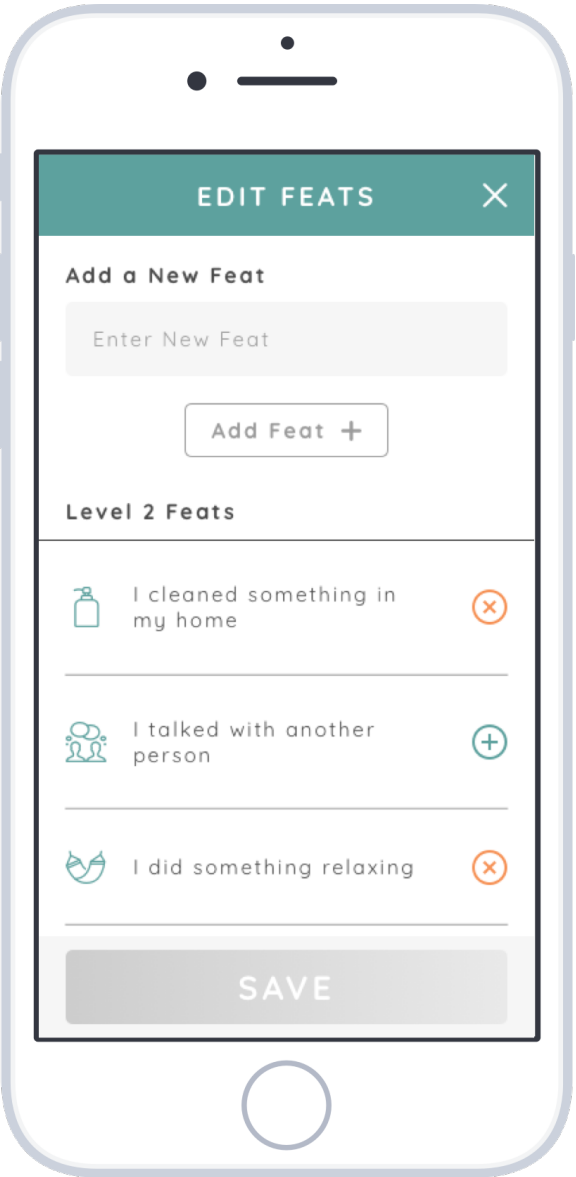
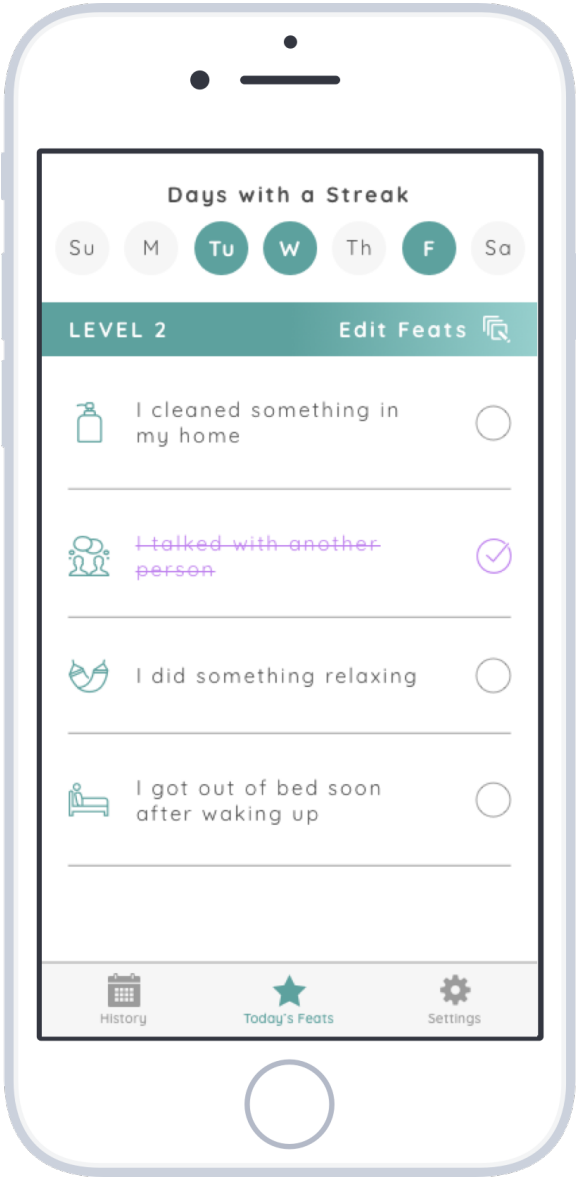
www.intellicare.northwestern.edu



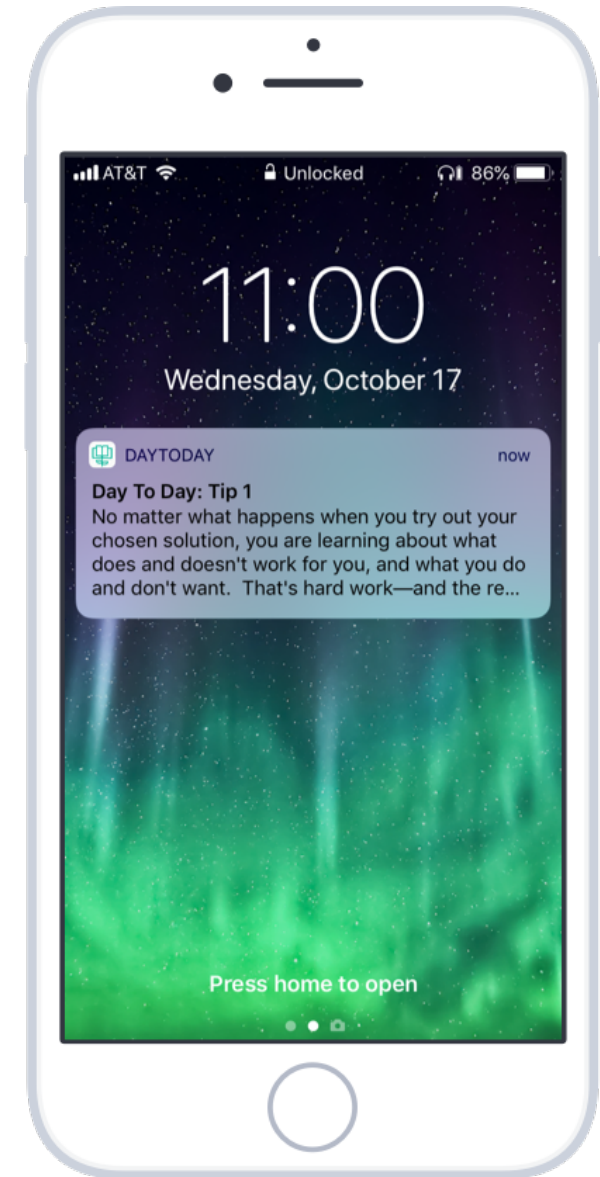
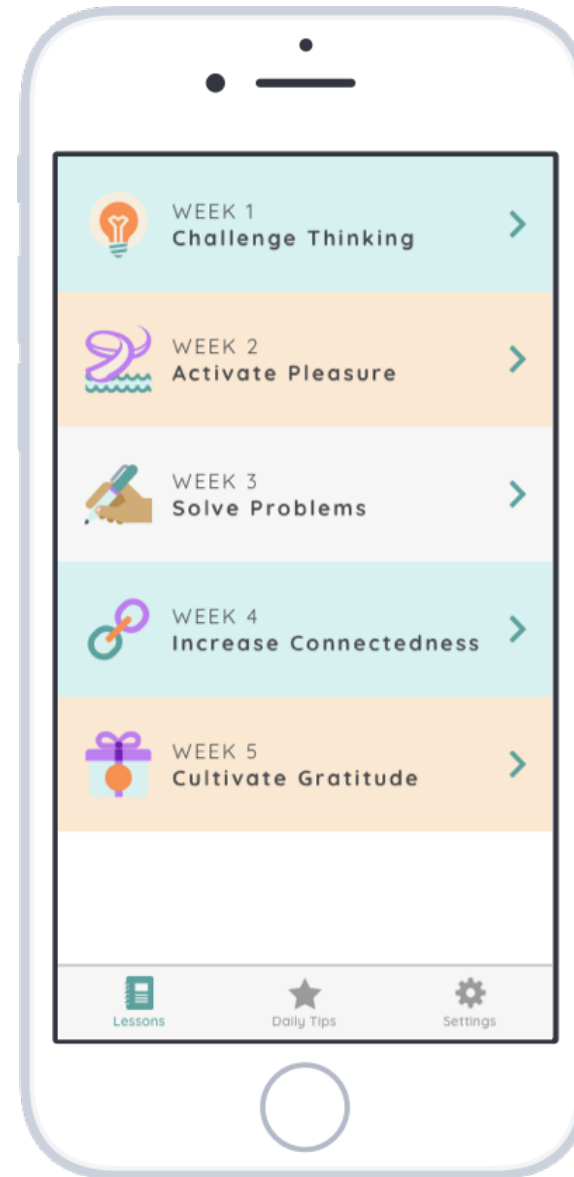
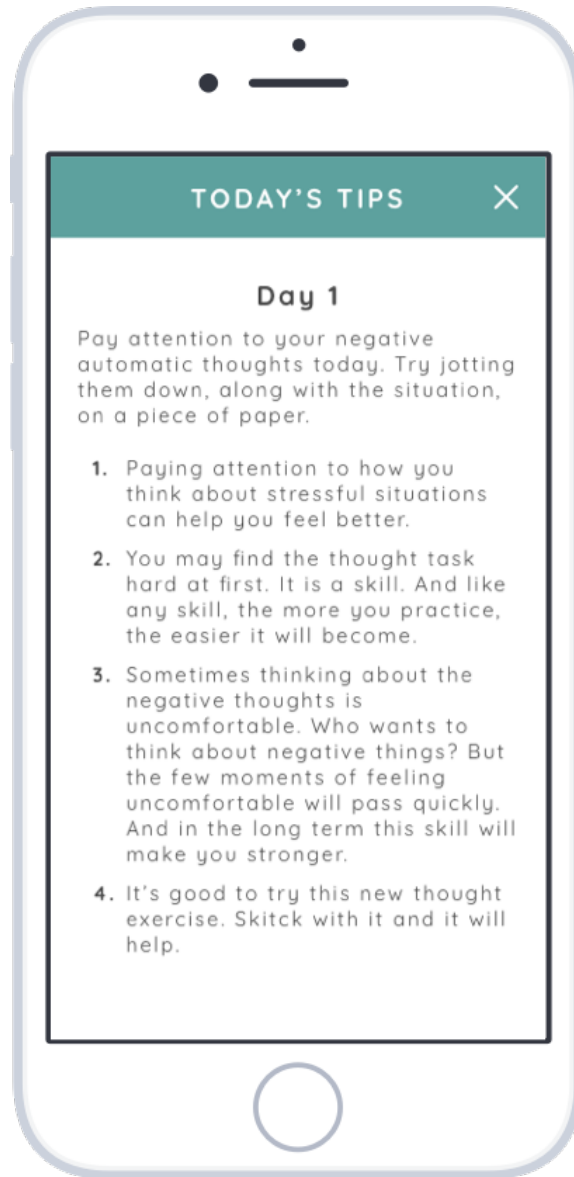
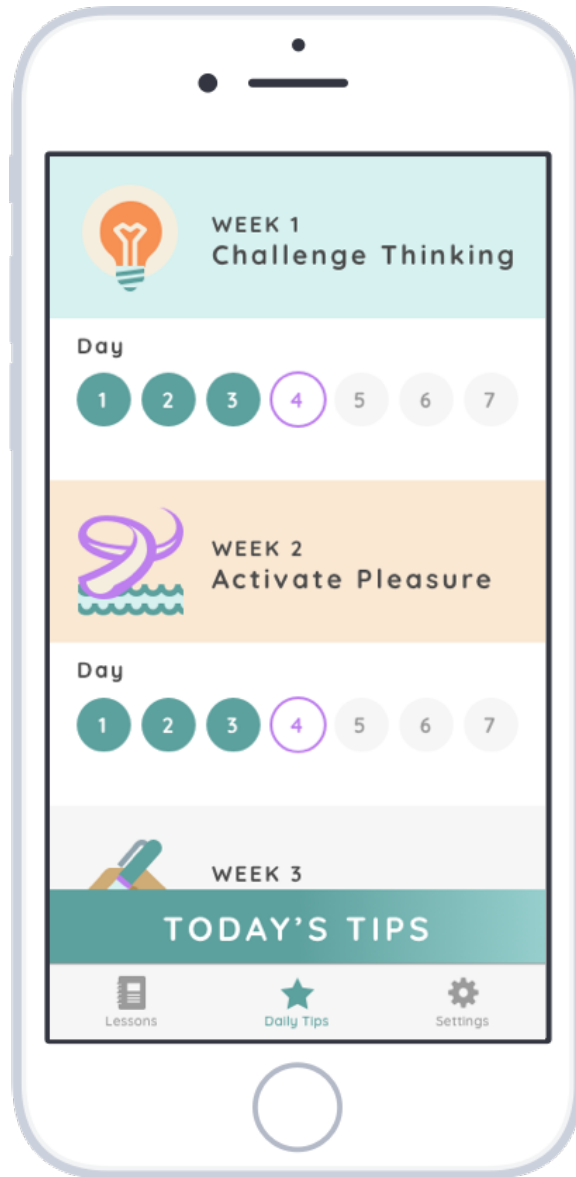
Elemental
Skills-Based
Eclectic



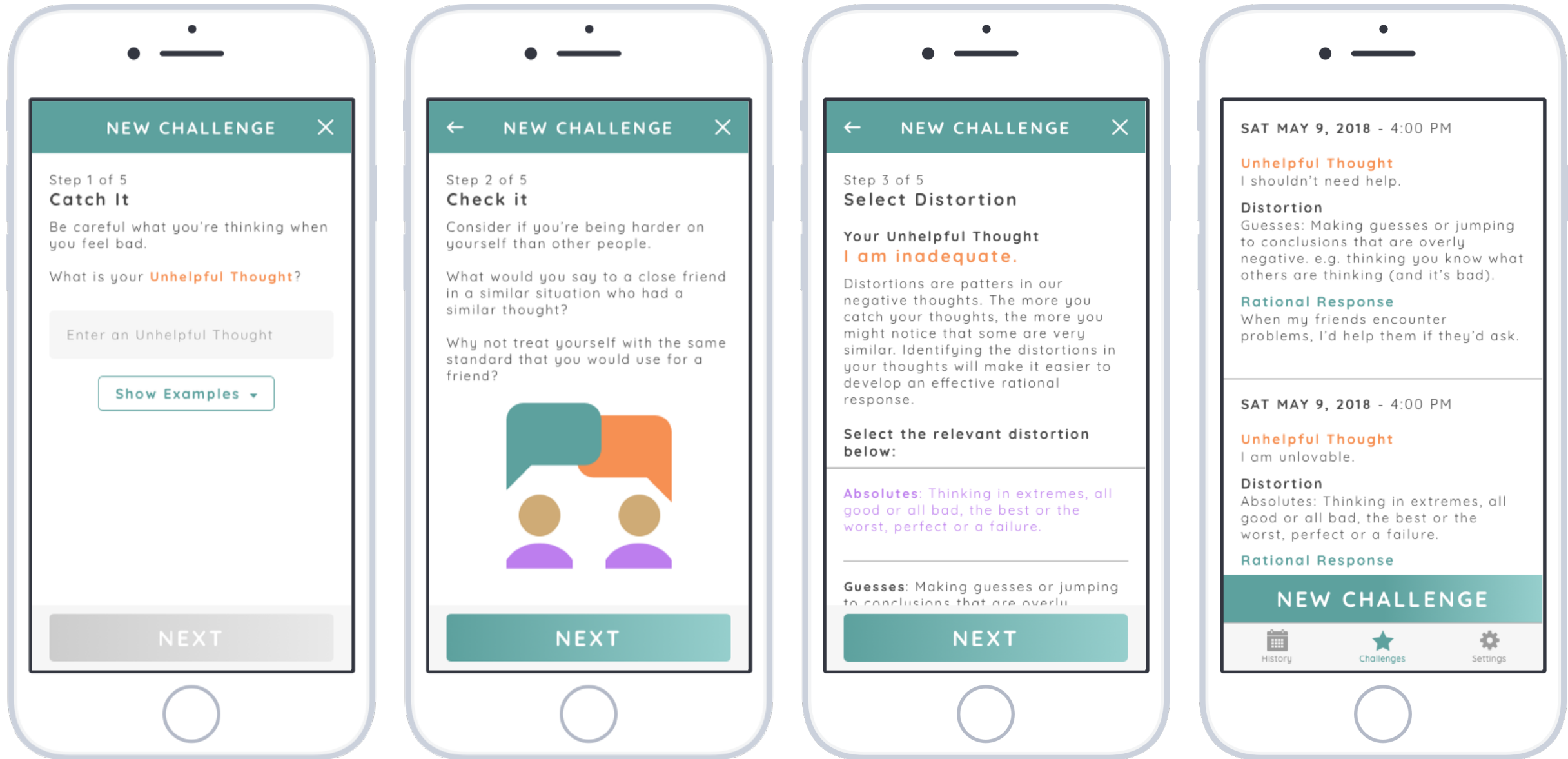
Daily Feats



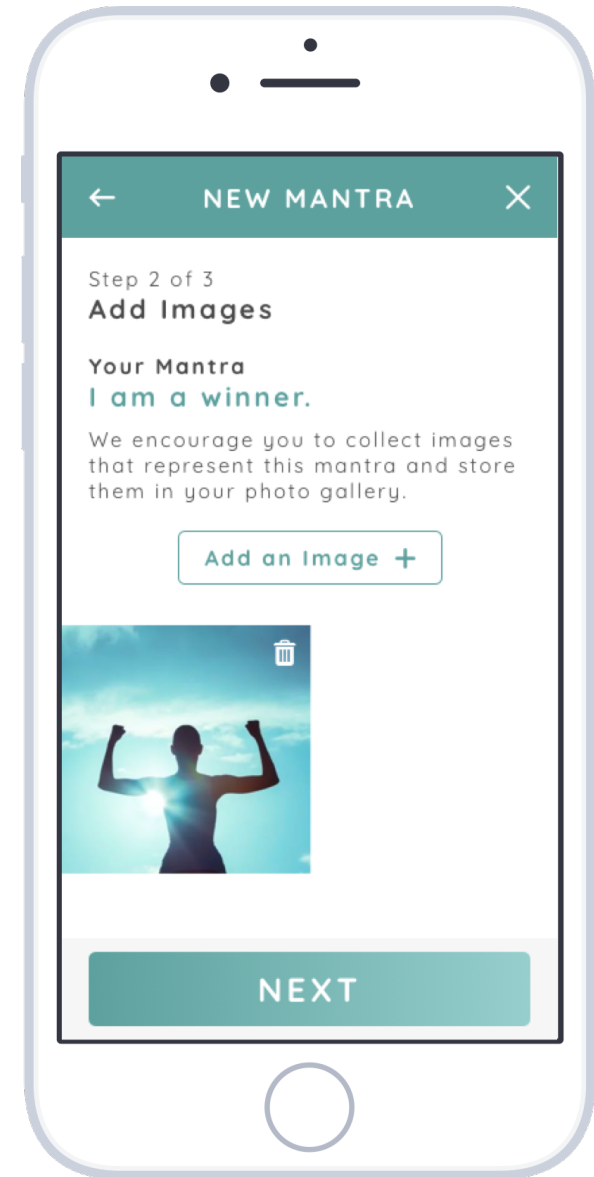
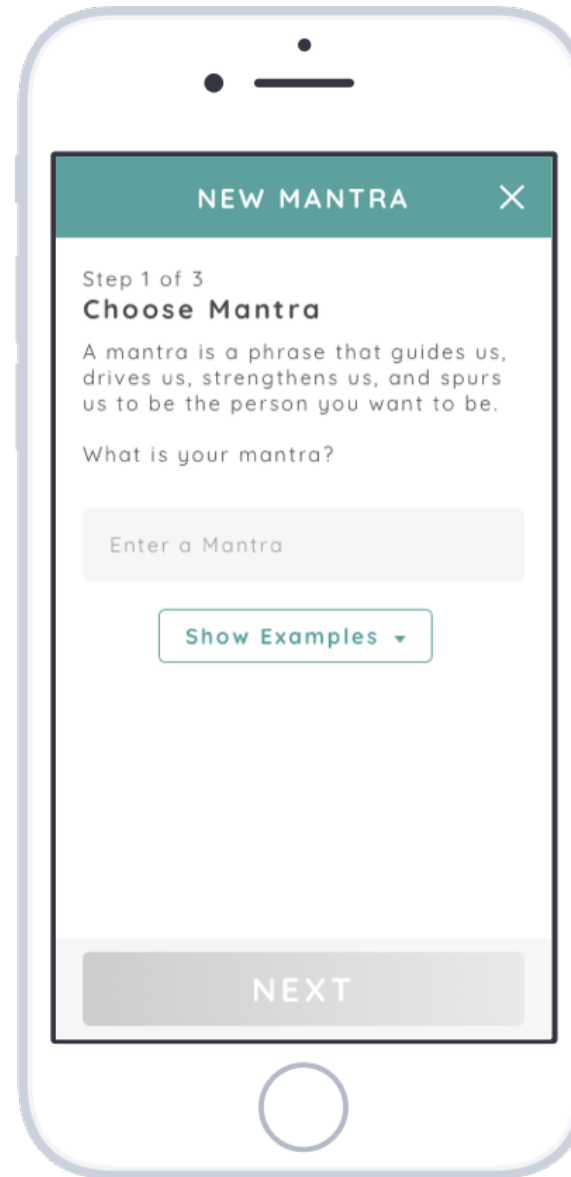
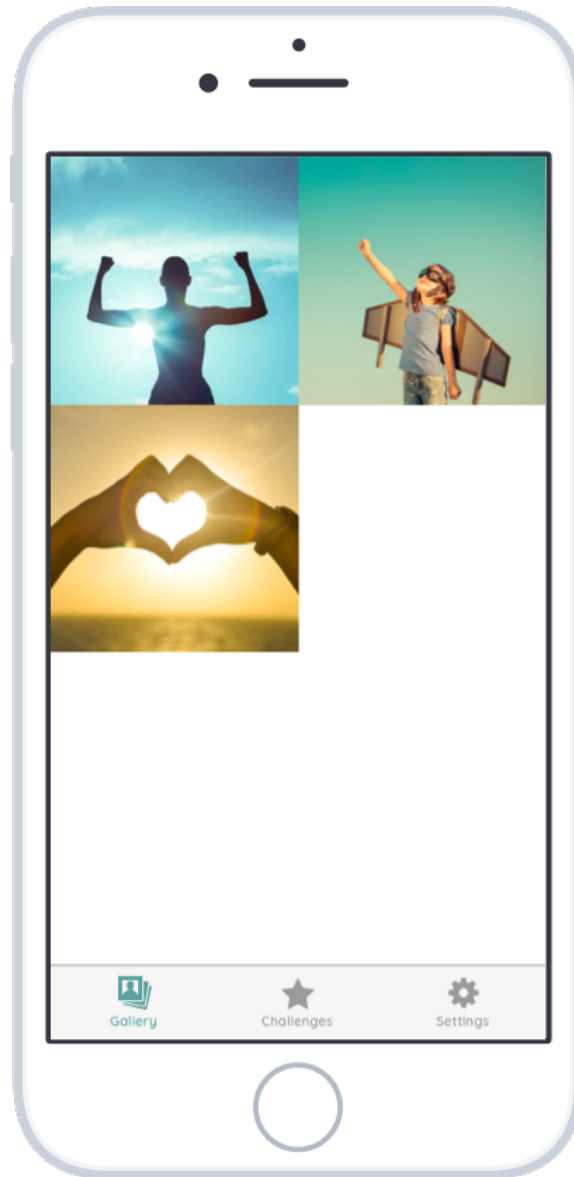
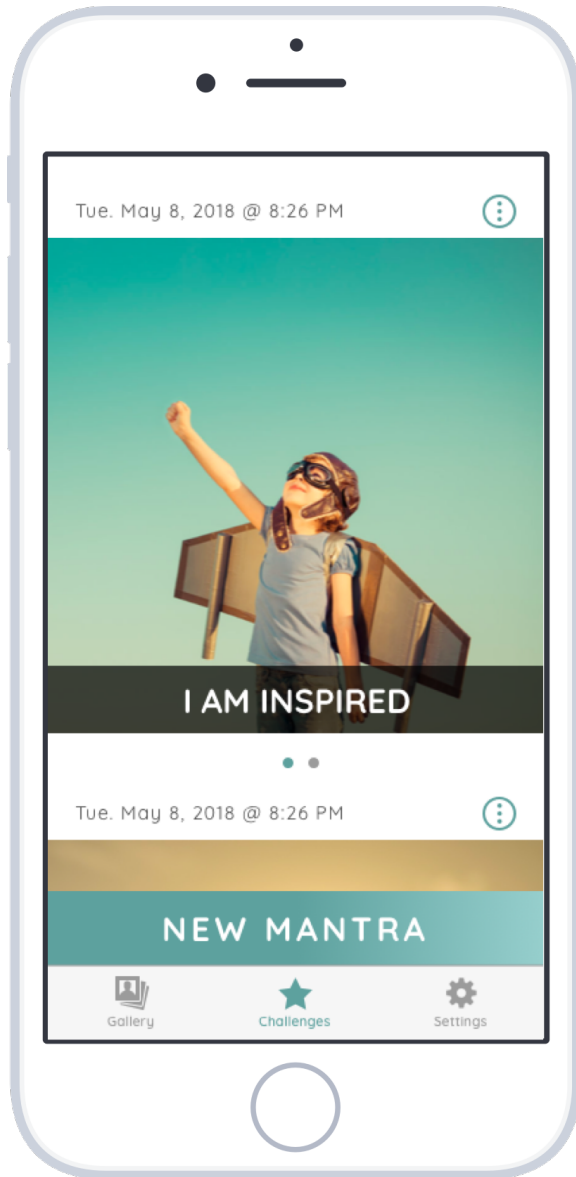
Day to Day



Thought Challenger



My Mantra



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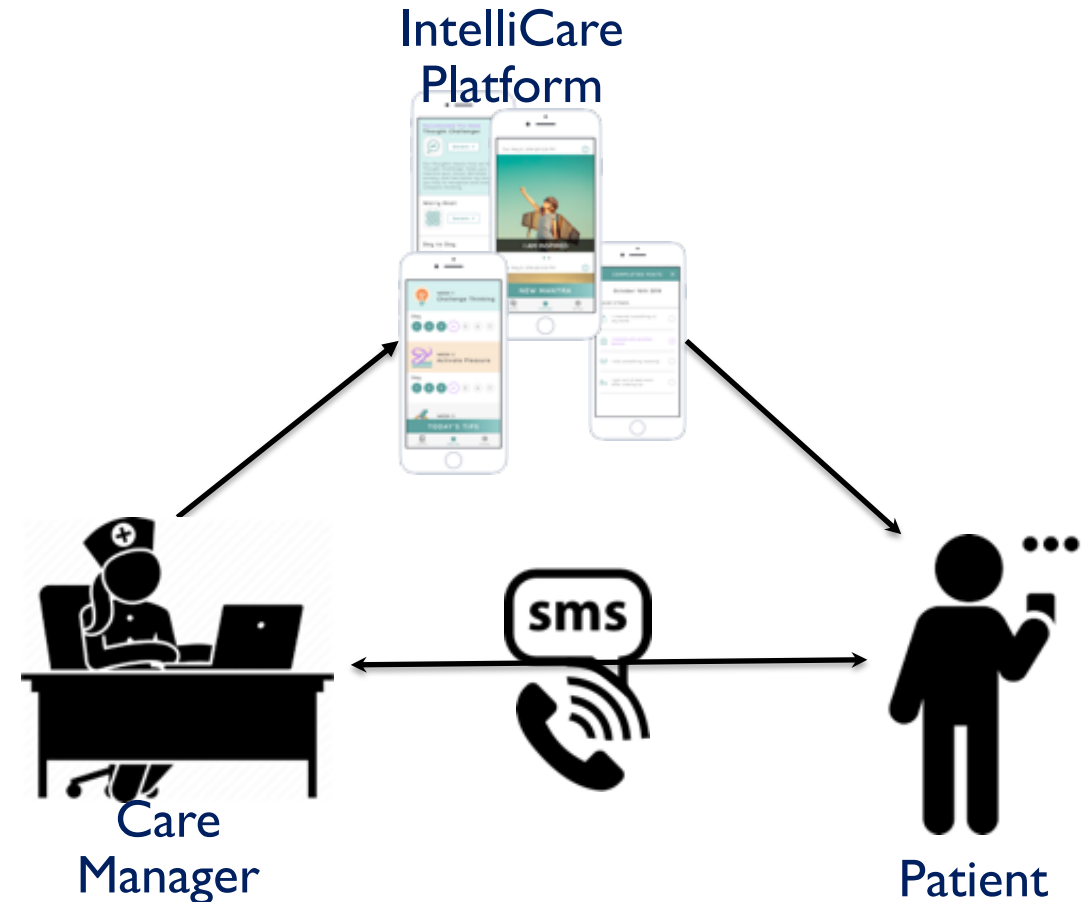
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Implications for Digital Mental Health Research

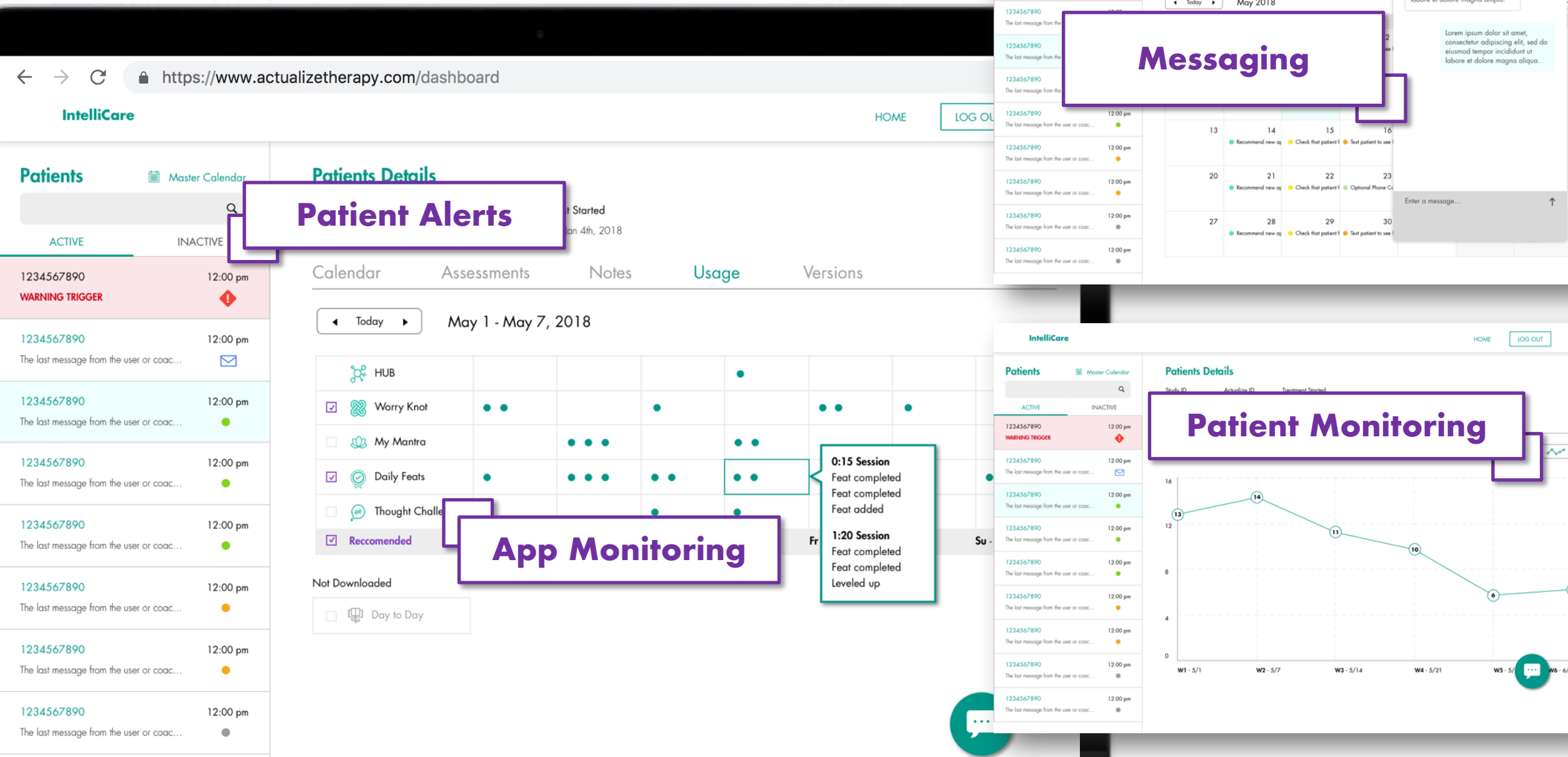
Summary

Trial Treatment Protocols

- 8-week treatment period
- 30 Min Engagement call
- Text message communication
 - CM sends 3 weekly texts
 - Patient can text any time
- Mid point call if desired
- Weekly app recommendations through Hub app



Care Management Dashboard

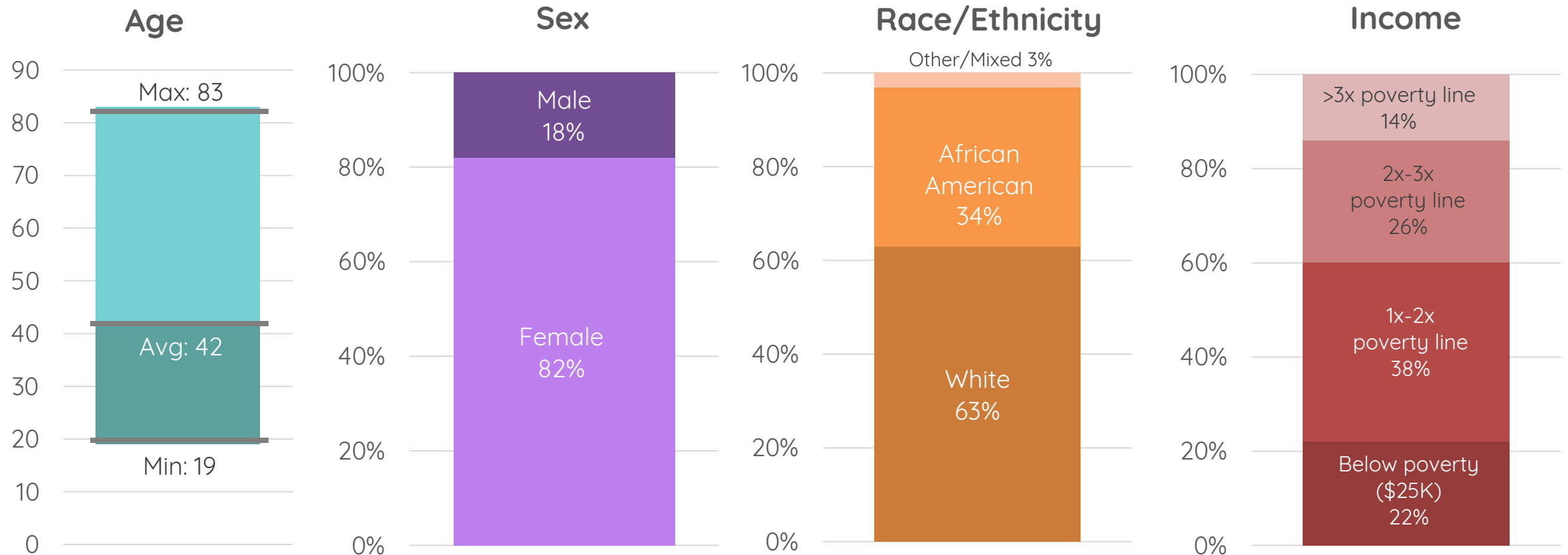


Randomized Controlled Trial University of Arkansas Med Ctr. Primary Care

Preliminary Results (do not disseminate)

148 Patients: PHQ-9>10 or GAD-7>8

Graham...Mohr. *JAMA Psychiatry*. 2020;77:906-914



Primary Outcomes

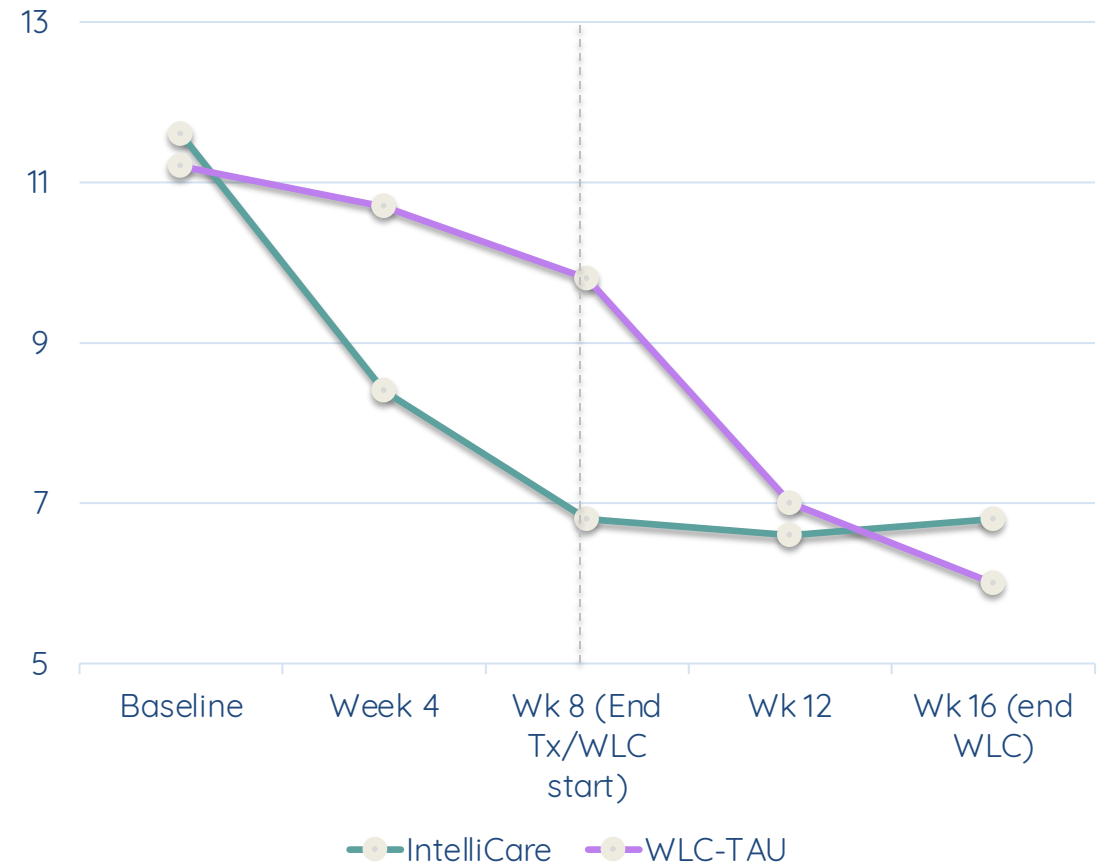
Depression Outcomes (PHQ-9)

$p < .001$



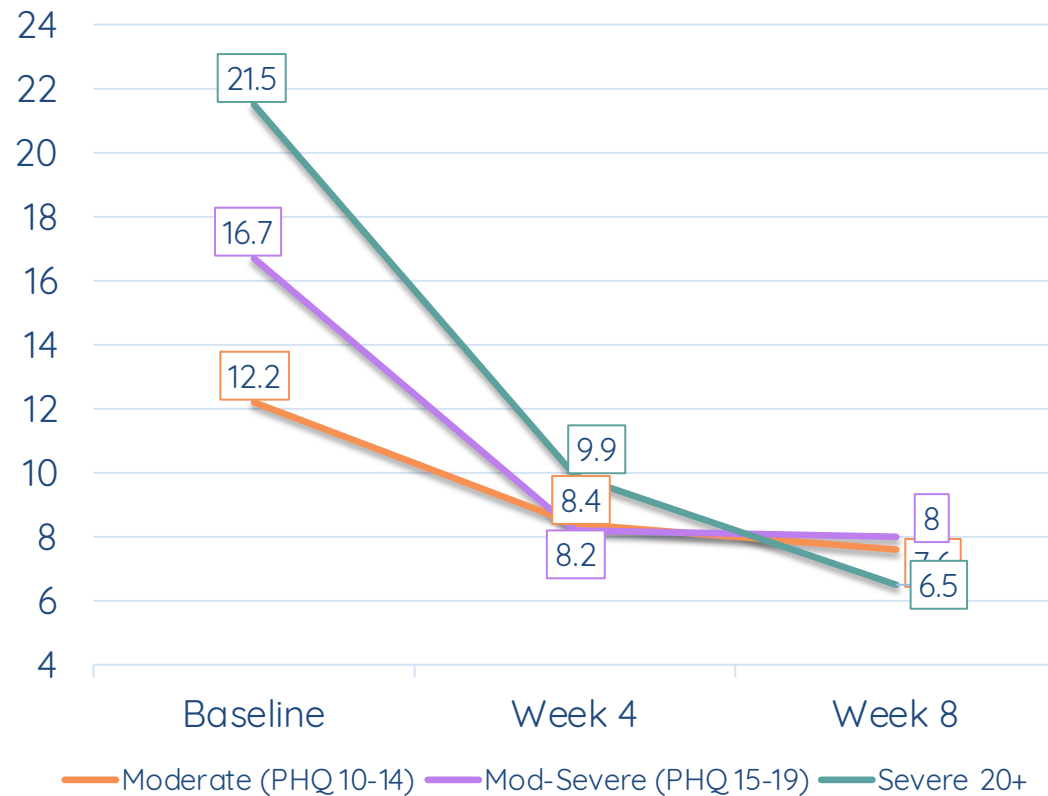
Anxiety Outcomes (GAD=7)

$p < .001$

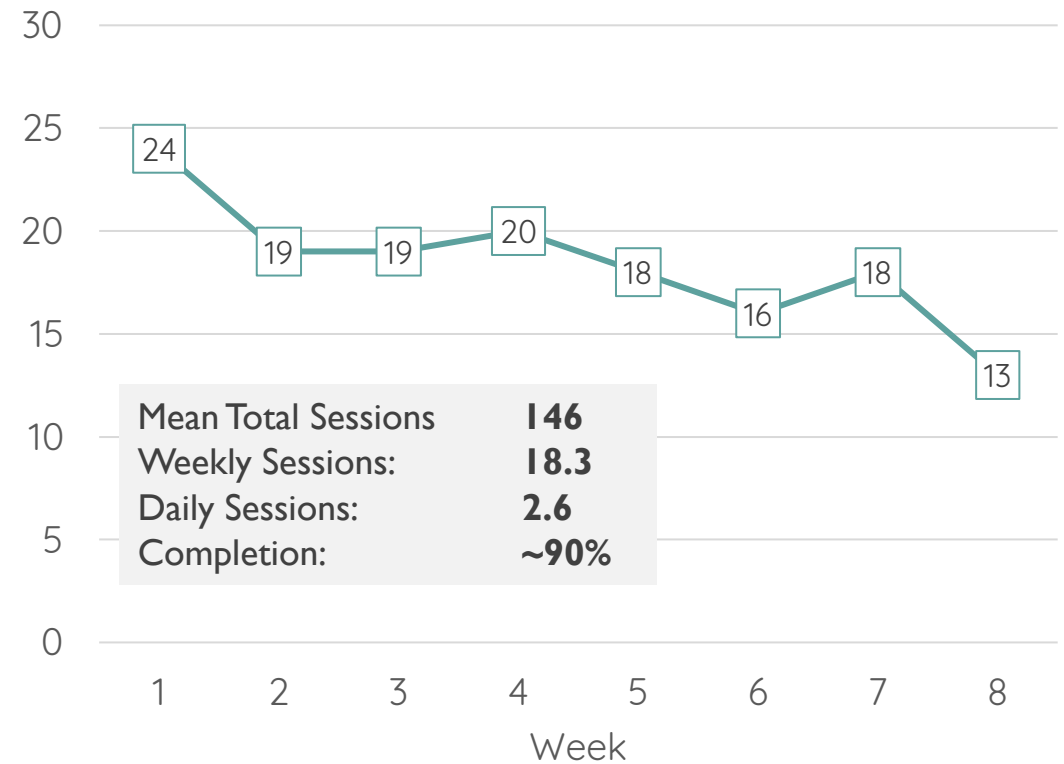


Effectiveness by Severity & User Engagement

Depression Outcomes (PHQ-9)



Average App Launches by Week



Assistance in Bundling Apps

Clusters of IntelliCare App Use

Kwasny, ...Mohr. JMIR. 2019;6:e11572

Thinking	Calming	Checklists	Activity	Junky apps
Thought Challenger Day to Day iCope My Mantra	Purple Chill Slumbertime	Aspire Daily Feats	BoostMe Move Me	MeLocate Social Force Worry Knot

Assistance in Bundling Apps Recommender System

Initial results

- ~30% use of any app in a week
- 55% with commodity recommender engine
- 64% with current stage recommendation algorithms

ROI MH109496

RECOMMENDER SYSTEM ALGORITHMS

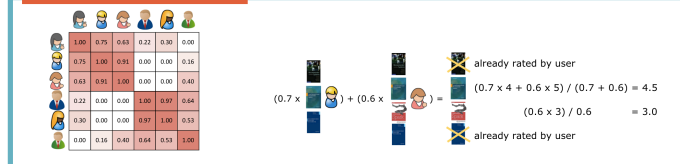
INPUT

	4	3		5	
	5		4		4
	4		5	3	4
		3			5
			4		4
				2	4
					5

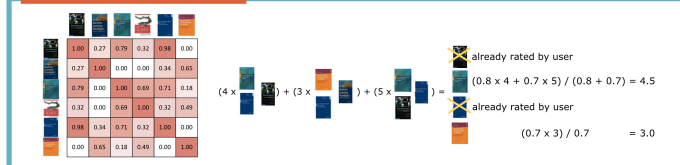
Introduction to Recommender Systems
Machine learning Paradigms
Social Network-based Recommender Systems
Learning Spark
Recommender Systems Handbook
Recommender Systems and the Social Web

FOUR RECOMMENDER ALGORITHMS ARE FED THE SAME INPUT AND PRODUCE DIFFERENT OUTPUTS. THEIR DEFINITIONS OF USER AND/OR ITEM SIMILARITY ACCOUNT FOR MOST OF THE DIFFERENCES.

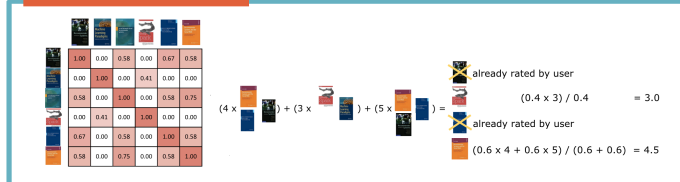
USER-BASED COLLABORATIVE FILTERING



ITEM-BASED COLLABORATIVE FILTERING



CONTENT-BASED FILTERING



HYBRID



OUTPUTS



Randomized Trial Evaluating Coaching & Recommender

Mohr, et al. *J Med Internet Res*. 2019;21(8):e13609

302 participants with $\text{PHQ-9} \geq 10$ or $\text{GAD-7} \geq 8$

Recruited from healthcare systems & Internet

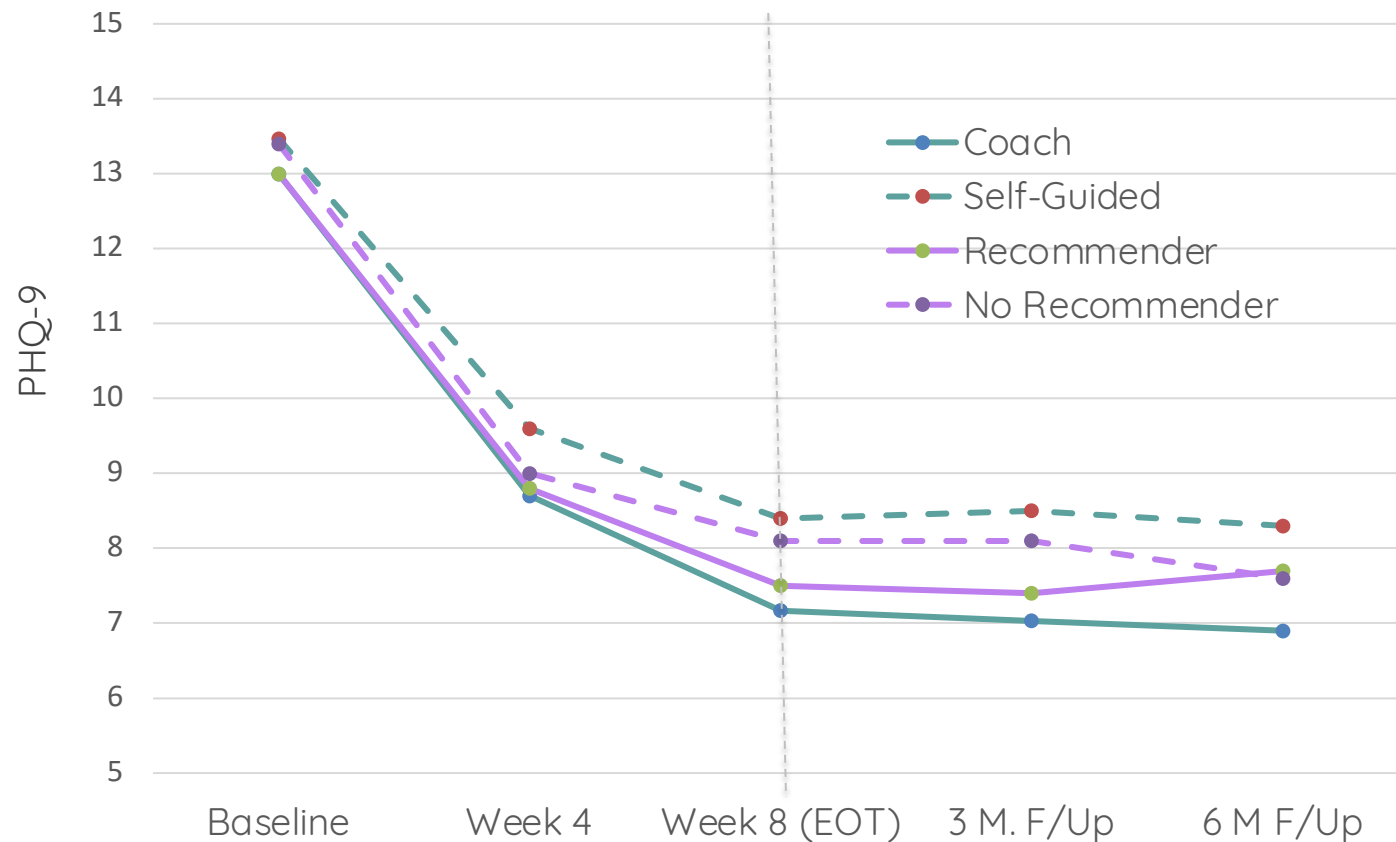
8 weeks of treatment; 6 months follow-up

		Coaching	
		Coached	Self-Guided
Recom-mender	Rec.	75	75
	No Rec.	76	76

Trial 3

Randomized Trial Evaluating Coaching & Recommender

Significant reductions in PHQ-9 & GAD-7 ($p < .0001$)



PHQ-9 (Depression)

- Coached vs. Self-Guided: $p = .06$
- Recommendations: $p = .04$

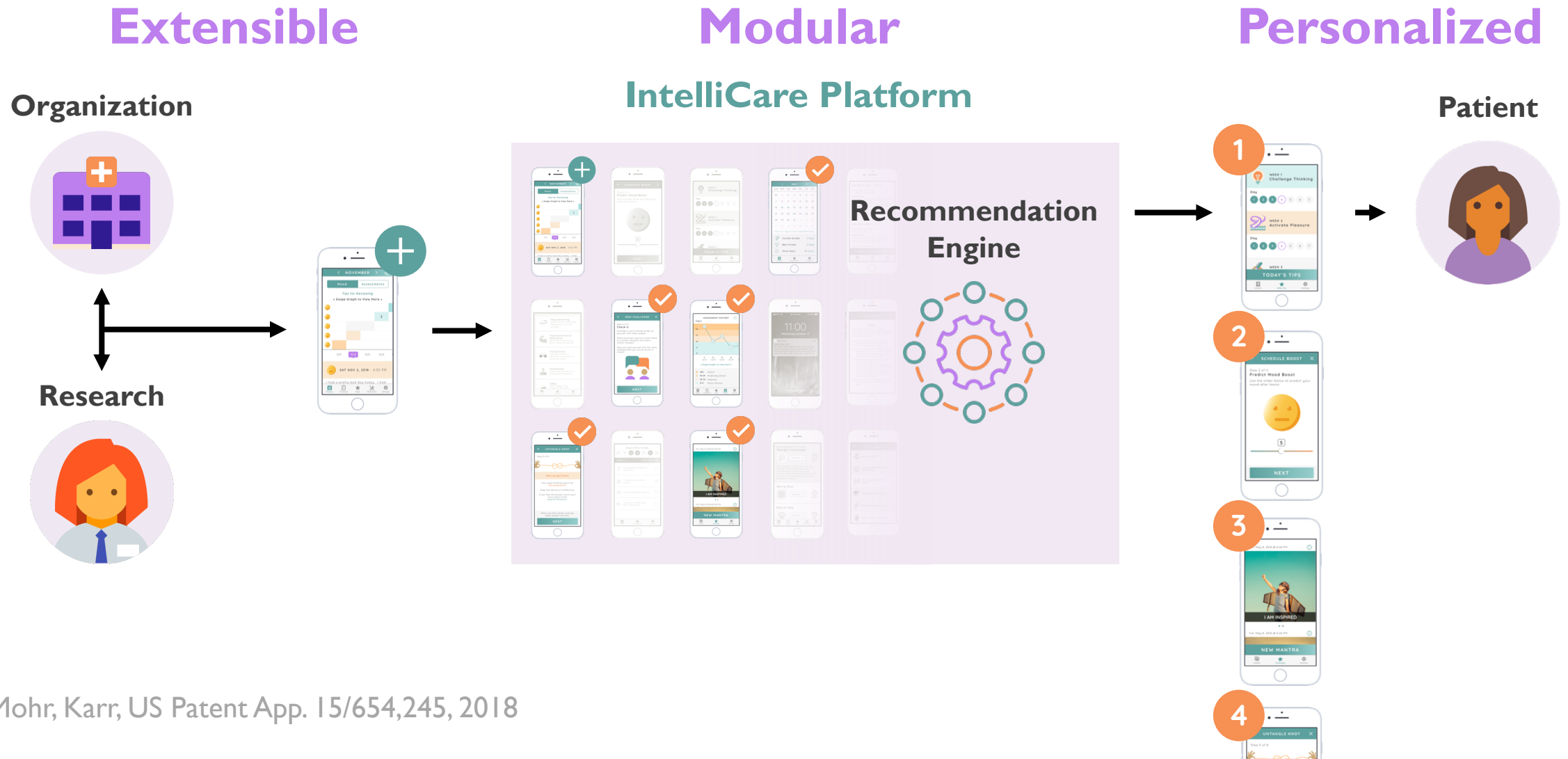
GAD-7 (Anxiety)

- Coached vs. Self-Guided: $p = .03$
- Recommendations: $p = .62$

App use:

- App sessions during treatment $M = 216$
- App use per day $M = 3.8$
- Participants continuing after tx 84%
- Days from EOTx to last use $M = 92$

IntelliCare Platform Overview



Other Published Research

Trials

Open Deployment of IntelliCare from App Stores

Lattie EG, et al. *Internet Interv.* 2016;4(2):152-158.

Coached IntelliCare for Depression/Anxiety

Mohr DC, et al. *J Med Internet Res.* 2017;19(1):e10.

IntelliCare for Distress during Breast Cancer

Chow, et al. *JMIR Cancer*, 2020;6:e16476

Mechanisms

Use Patterns

Kwasny M, et al. *JMIR Ment Health.* 2019;6(3):e11572

User Behavior

Chen AT, et al. *J Biomed Inform.* 2019;94:103187.

Recommendation Engine

Cheung K, et al. *J Am Med Inform Assoc.* 2018;25(8):955-962.

Current Research Projects

Breast Cancer

Chow, R01 CA248434



UVA Cancer Center

Perinatal Depression

Mohr, P50 MH119029



Student Mental Health

Lattie, K08 MH112878



Northern Illinois University



RUSH UNIVERSITY
MEDICAL CENTER

Collaborative Care

Mclver, R44 MH114725

Outline

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IntelliCare

Efficacy & Engagement Data

Implementation

Implications for Digital Mental Health Research

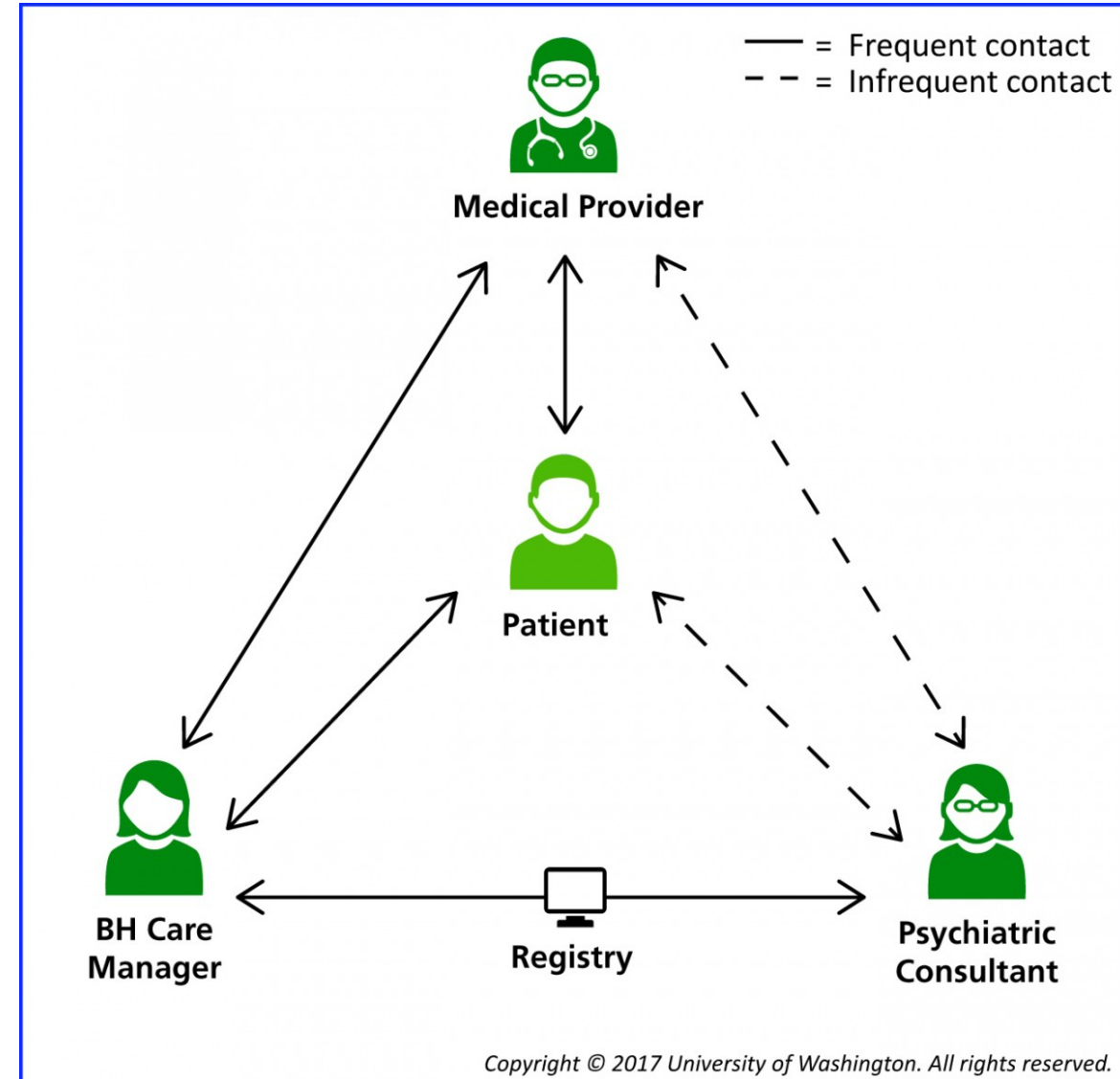
Summary

Rush University Medical Center

Collaborative Care

NIMH R44 MH114725

- ~250,000 primary care patients
- ~90% annual PHQ-9 screening rate in primary care
- ~5,000/year CoC uptake;
~2,000 to Rush CoC group
- 3 social workers



Care Managers

Ringland, Reddy, Mohr (2019) *ACM CSCW (Computer-Supported Cooperative Work and Social Computing)*.

Burgess, Reddy, Mohr (2019) *ACM CSCW (Computer-Supported Cooperative Work and Social Computing)*.

- Overwhelmed by patient flow
- Poor compliance with ongoing depression symptom monitoring
- Poorly defined objectives, no consistent workflow
- Emphasize social work over depression treatment
 - Case management primarily focused on medications
 - Care managers use informal consultation
- Concerns that technology will surface non-compliance, restrict clinical judgement





Care Managers

Satisfaction

- Deeper engagement with patients
- Knowledge they have helped
- Interactions with colleagues

Opportunities

- Lots of time spent trying to contact patients
 - Phone is only contact method; many people do not answer.
- Difficulty obtaining critical information, such as depression screening, medication adherence, referral uptake.

Patients

Eschler, et al., (2020) Proceedings of the Association for Computing Machinery (ACM)
Special Interest Group on Computer-Human Interaction (CHI)

Kornfield, et al., (2020) Association for Computing Machinery (ACM) Special Interest Group
on Computer-Human Interaction (CHI)

- Many patients don't know who their care managers are
- Unclear of purpose of care management
- Variety of preferences for communications medium (text, email, phone, voice mail)
- Using technology to support mental health – just not what we think.



Organizational Issues

Leadership needs

- Clear protocols
- Quality assurance
- Treatment outcome data
- Revenue generation

Organizational Tensions

- Population focus, not just patient
- Workplace politics: “Surfacing non-compliance”
- Central vs. local control of policies

Patient Management Platform

- Integration into EHR (Epic) & CoC registry
 - Care Management tools within Epic
 - PHQ-9, GAD-7
 - Alerts
 - Suicide risk
 - Worsening of symptoms
 - Treatment target reached
 - Medication adherence/side effects
 - Referral uptake
 - Noncompliance (e.g. PHQ-9 completion; app use)
- Communication through app
- Extend IntelliCare monitoring for 12 months
 - Identify relapses after treatment target reached

EHR

Full Integration into Epic



Pragmatic Trial

Year	2020												2021												2022			
Month	1	2	3	4	5	6	7	8	9	10	11	12	1	2	3	4	5	6	7	8	9	10	11	12	1	2	3	4
CM 1	TAU								IntelliCare										Sustainment									
CM 2	TAU										IntelliCare										Sustainment							
CM 3	TAU												IntelliCare								Sustainment							

Outcomes

- **Implementation** (CM productivity; medication adjustments, referral uptake, cost, etc)
- **Effectiveness** (PHQ-9)
- **Economic** (cost; billing – PHQ-9 administration)



Policy: Banbury Forum

Mohr, et al, in press, *Psychiatric Services*

-
- Stakeholders from payers, care systems, policy, employers, DMHT companies, NIMH, researchers, patients, international representatives.
 - Consensus statement:
 - Guided DMHTs should be available to all Americans
 - DMHT services and products should be reimbursable
 - Evidence standards framework required
 - Next steps towards reimbursement...

Agenda

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IntelliCare

Efficacy & Engagement Data

Design for HealthCare Systems

Implications for Digital Mental Health Research

Summary

Addressing the Research-to-Practice Gap



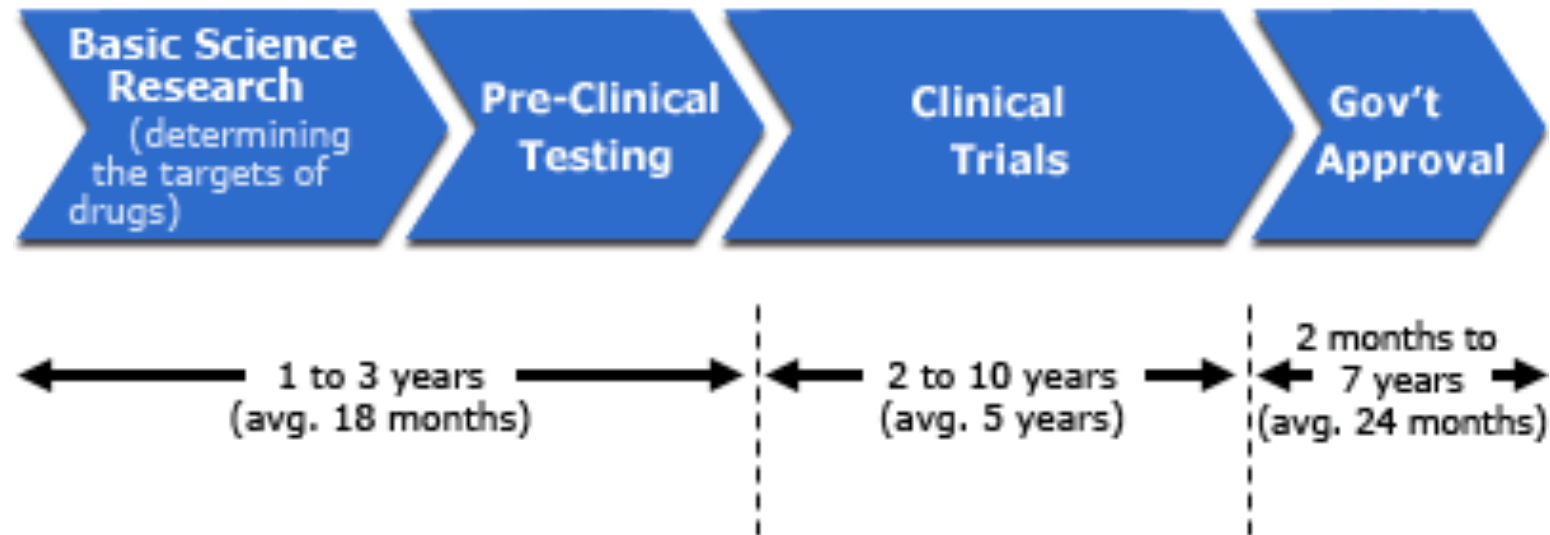
RCTs protect stakeholders (patients, providers, payers)

Methods derived from pharmaceutical trials

Contributors to Research-to-Practice Gap

Failure in Research Methods

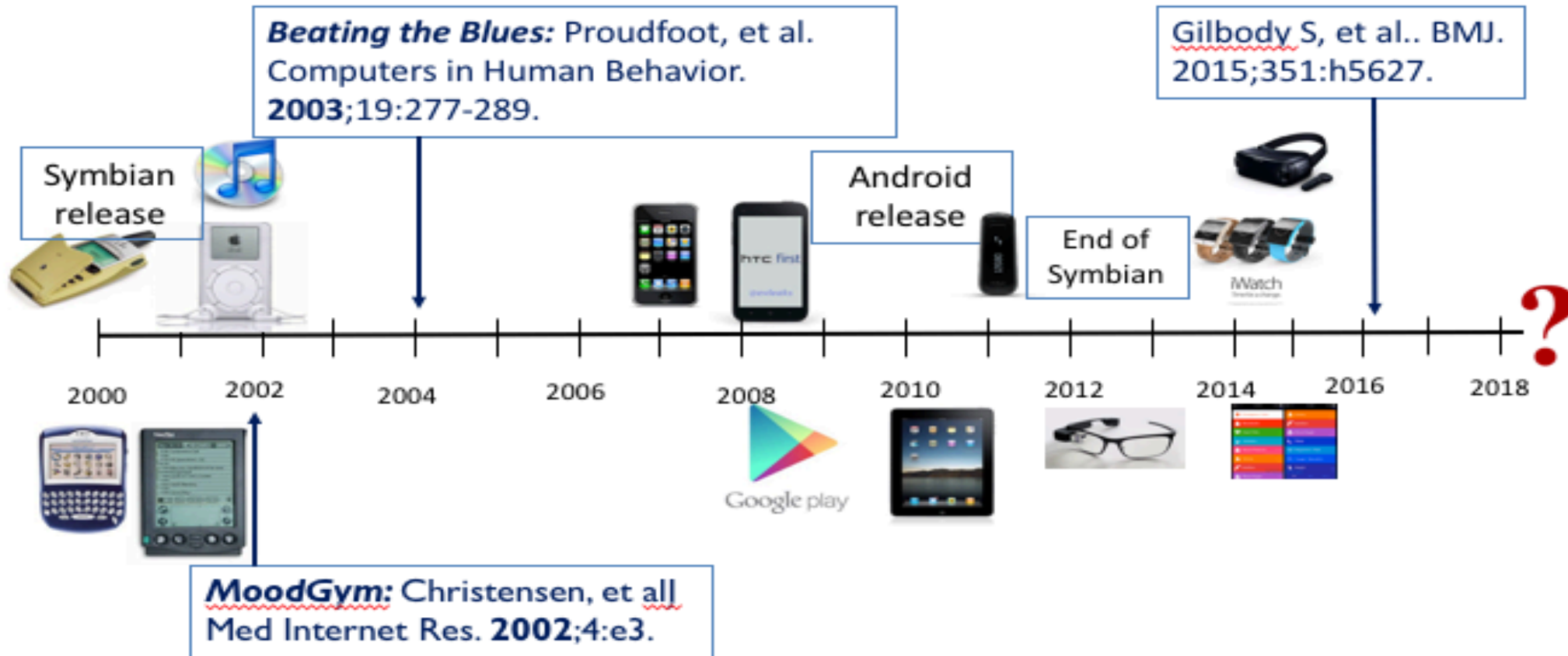
17 years to turn 14% of original research to the benefit of patient care



Balas, Boren.. in *Yearbook of medical informatics* 2000. Edited by van Bommel, McCray., Schattauer; 2000. pp. 65-70.

Contributors to Research-to-Practice Gap

Failure in Research Methods



Obsolescence: Changing technologies and healthcare organization

Contributors to Research-to-Practice Gap

Denominator Problem

$$\frac{\# \text{ Enrolled/completed}}{\# \text{ In recruitment pool}}$$

CBITs (recent trials)

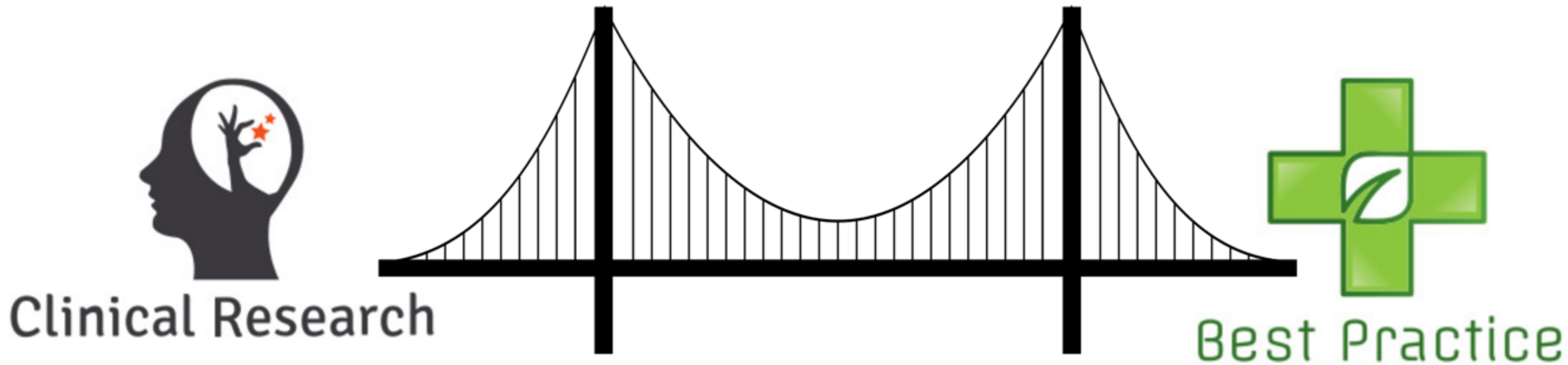
- >27,000 web screens
- ~ 700 recruited
- 2.6% of completed screenings recruited
- Very small ratio for views
- Direct recruitment of depressed clinic patients: **0.02%-1.3%**

Subthreshold depression & prevention

- 8.7 million Barmar Insurance members
- 1 year to recruit 406 participants
- Recruitment rate: **.00005%**

Buntrock, Ebert, et al. JAMA.
2016;315:1854-1863.

Implementation



Is it possible to build an implementation bridge?

Solution Focused Research

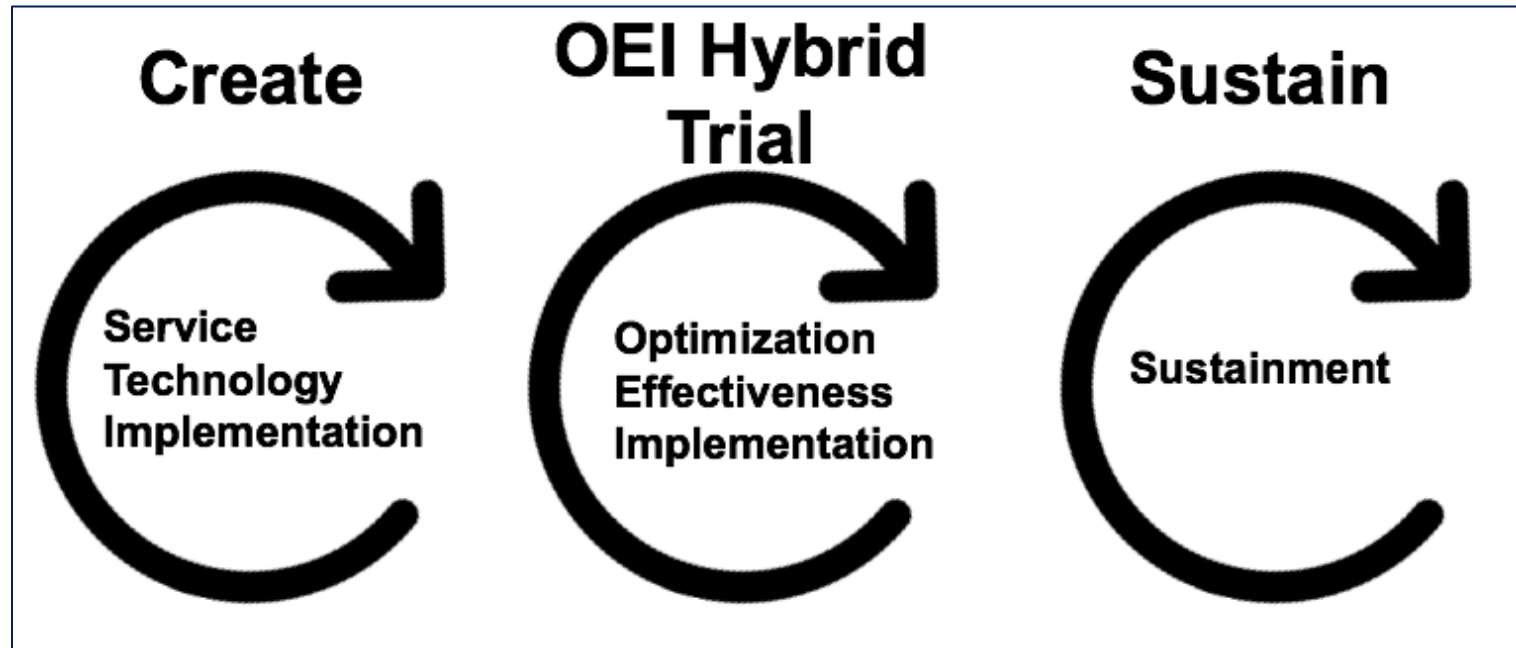
Accelerated Creation-to-Sustainment (ACTS) Model

Mohr, et al. *J Med Internet Res*. 2017;19:e153

Mohr, et al. *JAMA Psychiatry*, 2018;75:113-114

Research needs

- Solution-focused
- Design in care setting
- Iterative & Adaptive
- Rapid
- Sustainment

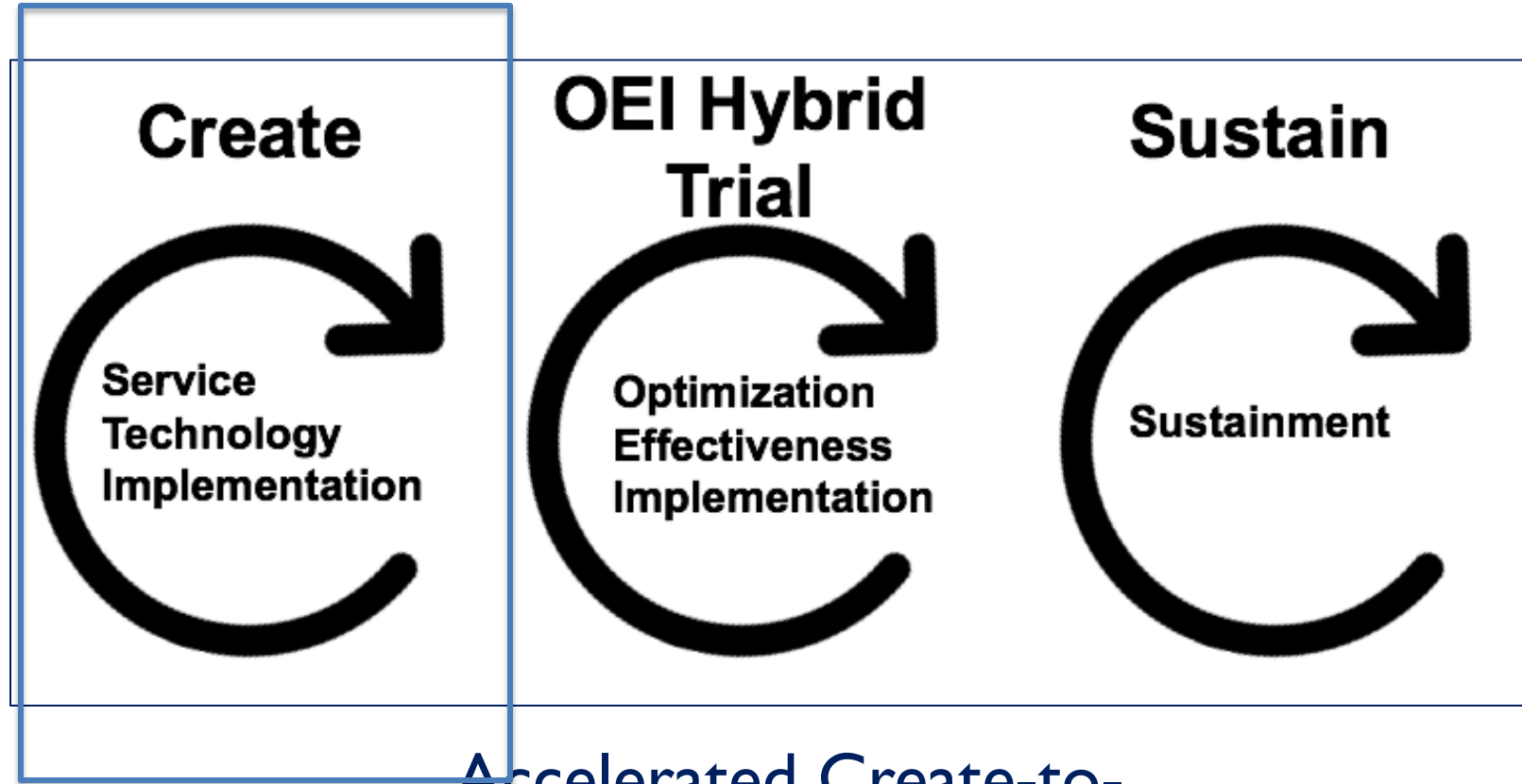


Create/Adapt

In the setting where it will be used!

Stakeholders

- Design in setting where it will be deployed
- Design with all stakeholders
 - Service Protocol
 - Technologies
 - Implementation Plan

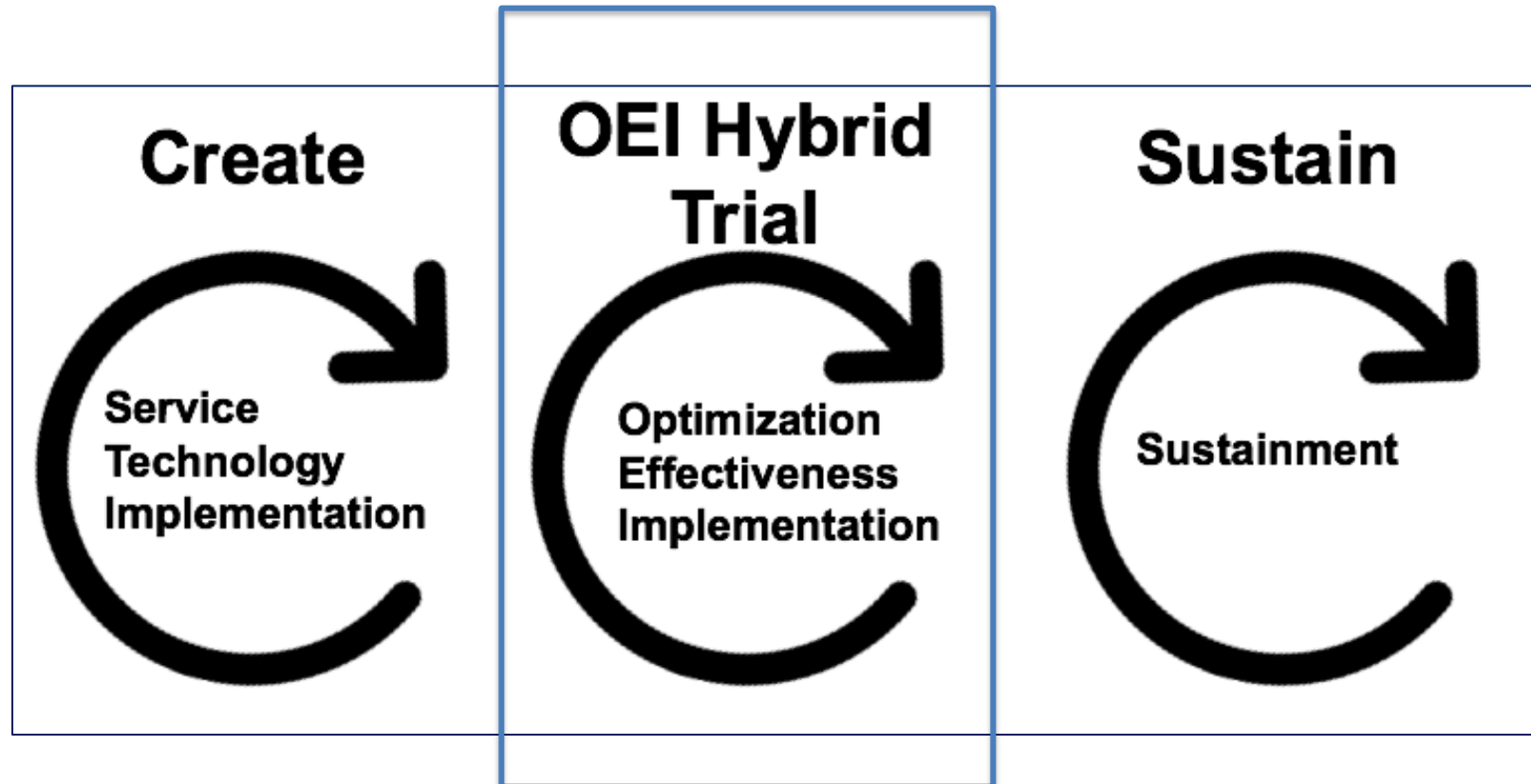


Accelerated Create-to-Sustainment (ACTS) Model

Optimization, Effectiveness, Implementation (OEI) Hybrid Trials

OEI Components

- *Evaluate effectiveness & implementation*
- *Optimization*
 - Technology
 - Service
 - Implementation



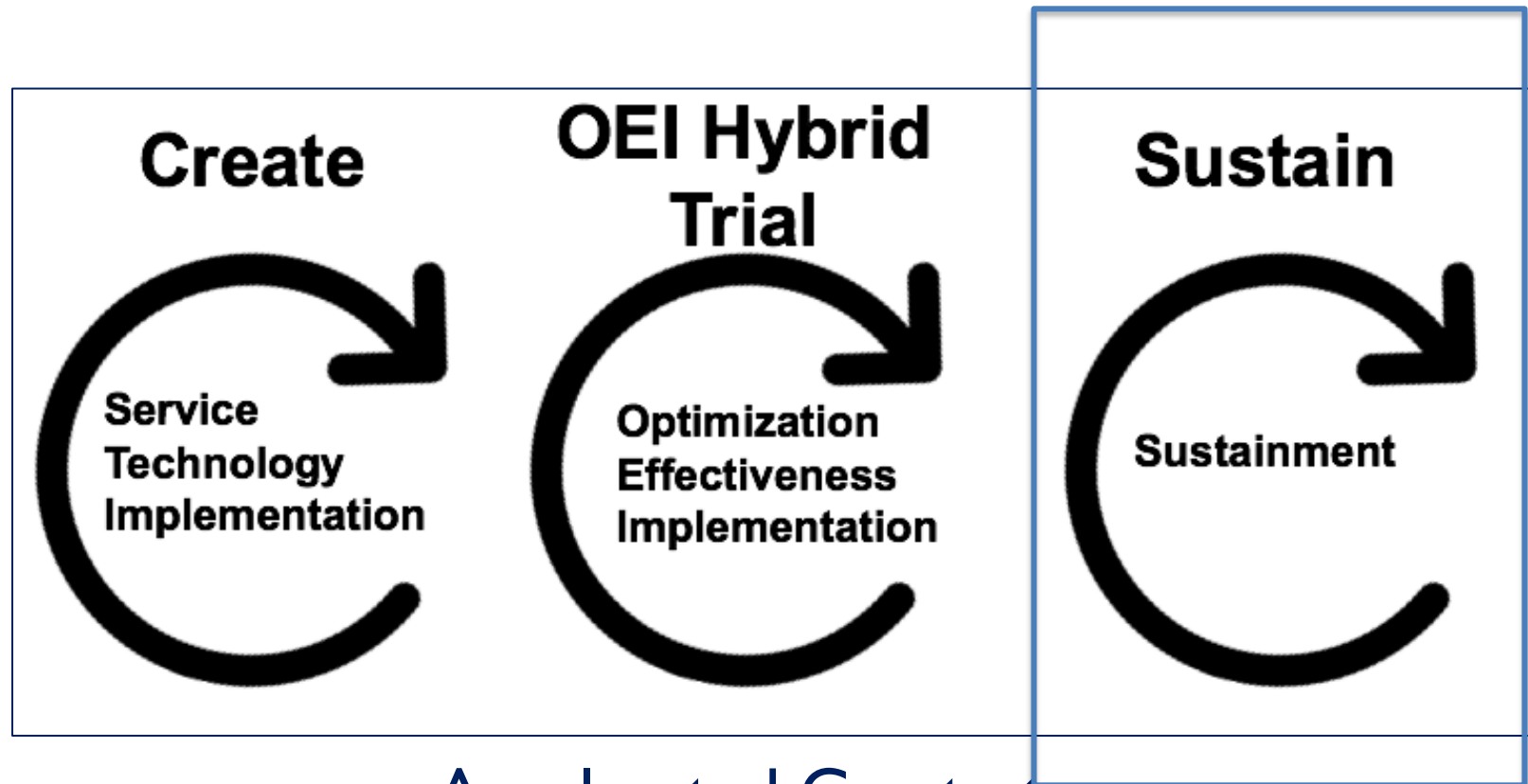
Accelerated Create-to-Sustainment (ACTS) Model

Sustainment

Chambers, et al. *Implementation science* : IS. 2013;8:117.

Aims and Methods

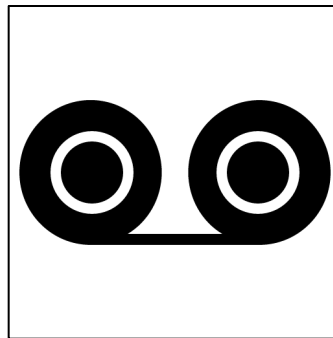
- Remove research support
- Low intensity data collection
- Quality Improvement



Accelerated Create-to-Sustainment (ACTS) Model

The Future of Digital Mental Health

Skeuomorph: A derivative object that retains design cues that were necessary in the original



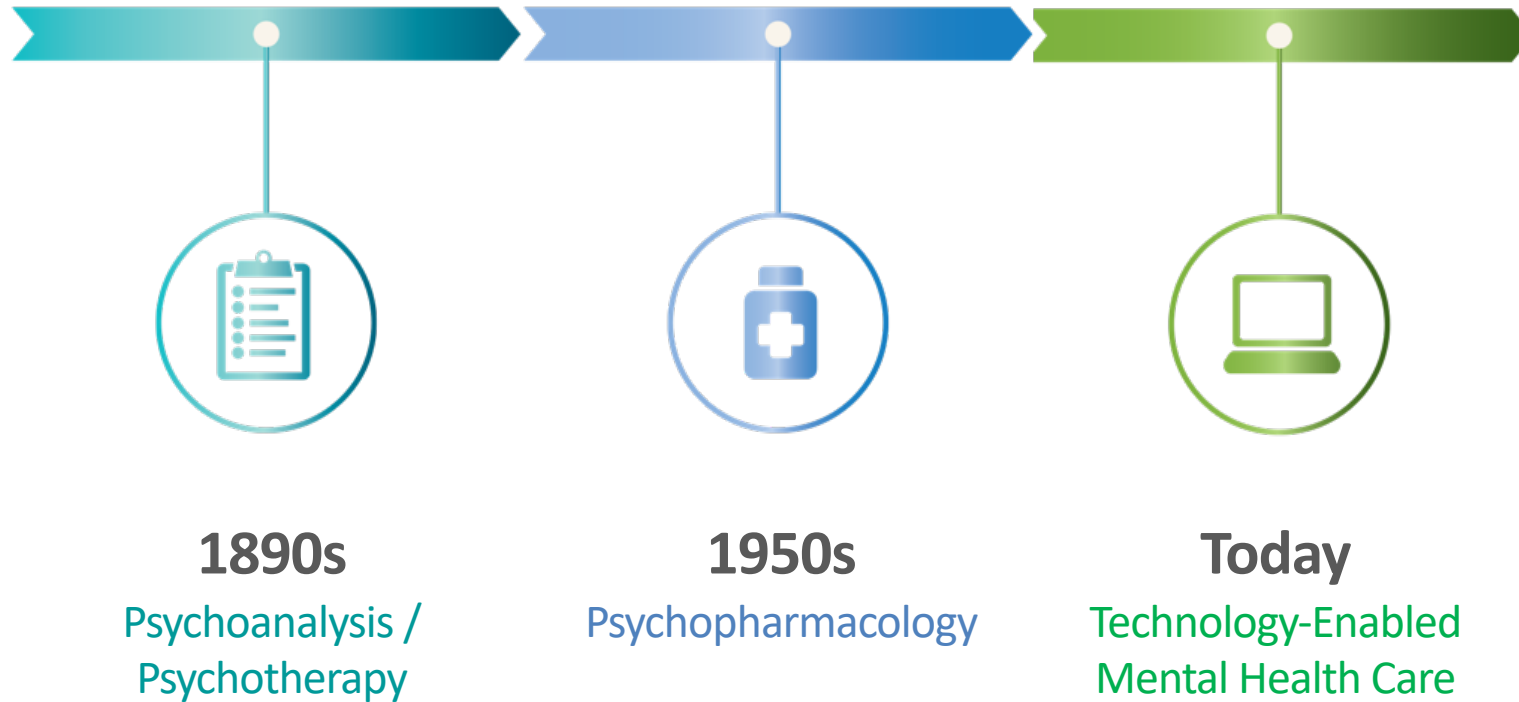
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Summary

- **Mindset:** Stop thinking about digital mental health as another form of delivery psychotherapy
- **Design:** Technology-enabled services that are *usable* and *useful* for end users (patients and clinicians) and healthcare systems.
 - Adapt to patients: Tools can be bundled by individuals
 - Adapt to setting needs
 - Adapt over time: Incorporate new components, technologies
- **Research Methods:** Rapid, Responsive, Relevant
- **Goals:** Sustainable implementation

Revolution in Mental Health Care

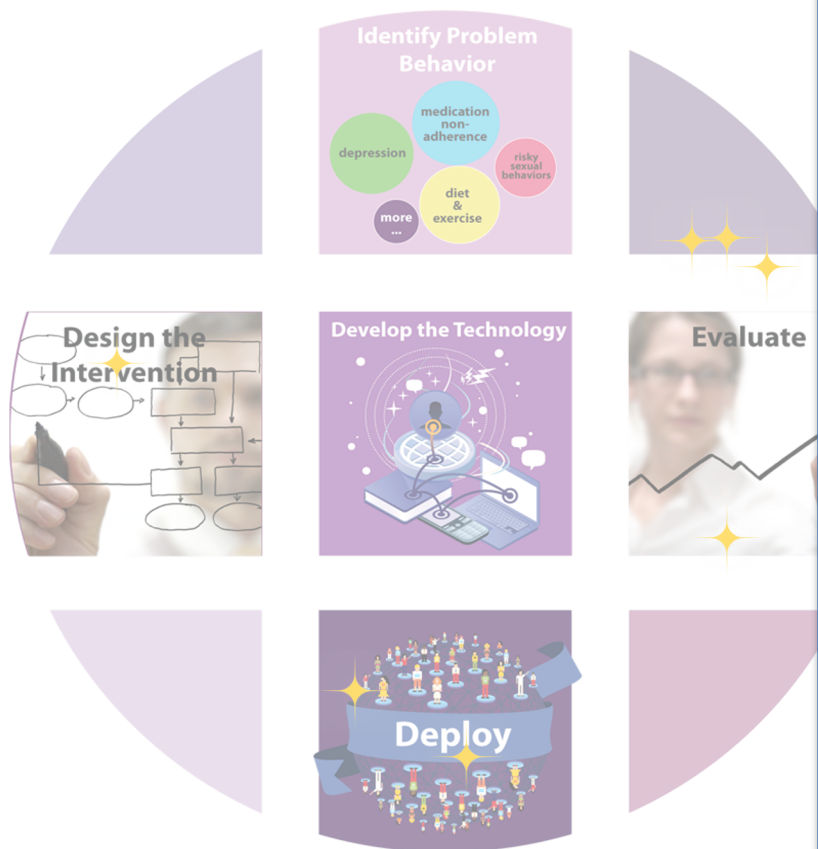


Thanks to our team!

www.cbits.northwestern.edu

David C. Mohr, PhD: d-mohr@northwestern.edu

 DavidCMohr



Some of our Team

Madhu Reddy, PhD
Andrea Graham, PhD
Neha Gupta, MD
Susan Kaiser, MPH
Emily Lattie, PhD
Ken Cheung, PhD
Stephen Schueller, PhD
Evan Goulding, MD
Chris Karr, MA
Mary Kwasny, PhD
Darren Gergle, PhD
Jen Nicholas, PhD
Ashley Knapp, PhD
and many, many more...

NIMH T32 Postdoctoral Training Program in Digital Mental Health

<http://cbits.northwestern.edu/training-081020/>