Modeling to Learn,

Test. Don't guess.

Helping Care Teams Improve Implementation of Medication Assisted Therapies for Alcohol and Opioid Use Disorders



@LZPhD

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Workshop Learning Objectives

Time	Workshop	Focus
1:00 PM – 2:30 PM	Modeling to Learn (MTL) Helping Care Teams Improve Implementation of Medication Assisted Therapies for Alcohol and Opioid Use Disorders	mtl.how

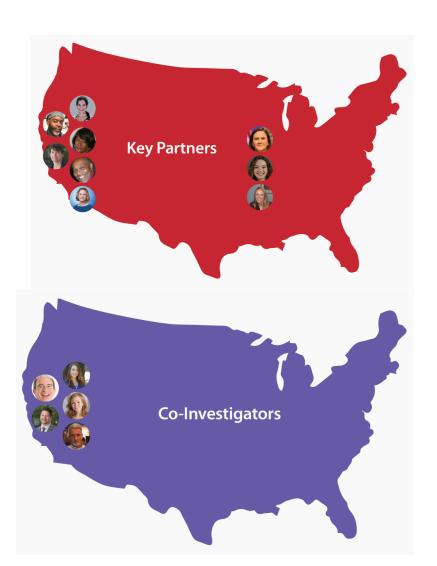
- 1. Introduce principles of **systems science** that can be applied to improve **implementation** of evidence-based pharmacotherapy (EBPharm or MAT).
- 2. Describe why **participatory learning** from simulation improves team's decision-making related to MAT.
- 3. Demonstrate the **Medication Management (MM)** module of *MTL*.
- 4. Illustrate how simulation learning, using hyper-local team data, helps to identify the best way to **optimize local MAT implementation resources**.



mtl.how/team

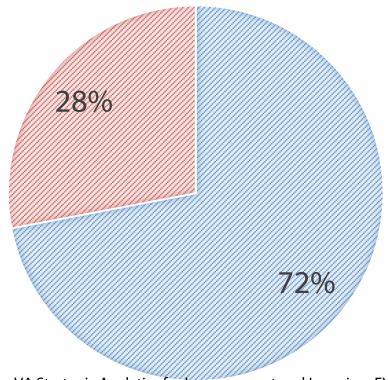






The problem of EBP reach in teams: How can we reach more patients with our highest quality care?

Other services
Evidence-based practices



Source: VA Strategic Analytics for Improvement and Learning, FY 2017



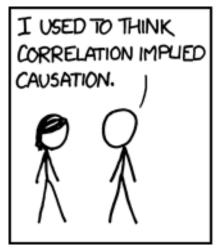
Veterans Health Administration

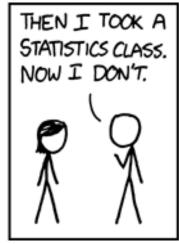
Model of a US National Health Care System

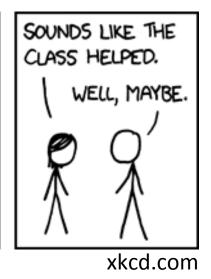
American J. Public Health 97, 2007

- VA innovates with national dissemination efforts to train providers in evidence-based mental health practices
- 2. Enterprise-wide quality measures
- 3. Clinical practice guidelines and mandates for evidence-based care
- 4. National electronic health information system
- 5. Mental health care coordinated in multidisciplinary teams

What works to improve EBP reach, why, and under what conditions?







Understanding causes of EBP reach, in local context, is critical to our stakeholders.

Our aims.

- develop a systems understanding of VA mental health services and the limited reach of evidence-based mental health care.
- empower mental health stakeholders to make locally optimized quality improvement decisions.



Target State: Lean SMART Goal

By April 2015, 40% of patients newly seen in outpatient mental health at Menlo Park for depression, PTSD, or anxiety disorders will have two psychotherapy visits completed within 28 days from time of intake assessment.

Specific. Measurable.

Actionable: if never achieved morale may suffer. Realistic: with the available resources.

Time frame: A due date.

Local clinic strategies are needed to address local differences.

Clinic 1	Clinic 2
3548 unique patients/year	2043 unique patients/year
Lower caseload per provider	Higher caseload per provider
Rare wait for initial appointment	Occasional waitlist to get into clinic
5.2 psychiatrists per 9 EBPsy providers	3.0 psychiatrists per 4 EBPsy providers
Higher EBPsy providers/MD ratio	Lower EBPsy provider/MD ratio
Higher EBPsy base rate	Higher EBPharm base rate
Providers often self refer for EBPs	Referrals to other providers by necessity
Multiple on-site specialty programs	Only telehealth specialty care
Training program site multiple disciplines	No trainees providing care
Most groups "open" (ongoing enrollment)	Most groups "closed" (infrequent opening)
Shorter time to next available appointment	Longer time to next available appointment

MTL focuses on learning among frontline

teams making	EBP-related care decisions.
Drawn from Hovmand 2014 & Scaccia et al., 2015	

Scientific Model

General

Capacity

EBP Specific

Capacity

Problem

Learning

- Coordination
- **Analysis** Restructuring

Why problems persist

Stakeholders cannot or do not learn and adapt to their situation.

Conflict or lack of stakeholder consensus.

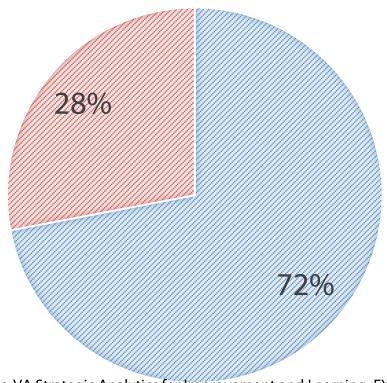
Policies are inconsistent with

the real system constraints.

The underlying structure of the system prevents workable solutions.

We define limited EBP reach among our patient population as a system behavior.

Other services
Evidence-based practices



Source: VA Strategic Analytics for Improvement and Learning, FY 2017

Saturation achieved during structural behavioral validity testing.

Direct Structure Tests

- Empirical
- Structure-confirmation
- Parameter-confirmation

Systems Theory Tests

- Structure-confirmation
- Parameter-confirmation
- Direct extreme-condition
- Dimensional consistency

Stakeholders & Literature

- Reviews and evaluations
- Exemplar SD Models

Barlas, 1996

Structure-oriented Behavior Tests

- Extreme-condition test
- Behavior sensitivity test
- Modified-behavior prediction
- Boundary adequacy test
- Phase relationship test
- Qualitative features analysis
- Turing test



EBP Reach Behavior pattern tests

STRUCTURE

BEHAVIOR

National Center for PTSD

VA Employee Education Services

Office of Mental Health & Suicide Prevention

OUR STAKEHOLDERS

VA policy-makers, patients, and providers from psychiatry, psychology, social work, nursing & certified peer support specialists

Veteran Patients (VAPOR)

Office of Healthcare Transformation

Directors of Outpatient Mental Health & VISN MH Leads

Core Modeling Group of Frontline Staff

Frontline Teams



VAPOR introduces Modeling to Learn

Our PSD approach – Participatory Research:

Community
Based
System
Dynamics

A partnership approach to research that equitably involves stakeholders in all aspects of the research process and in which all partners contribute expertise and share decision-making and ownership.

Participatory Research is an epistemology.

- Scientific inquiry that that actively considers the scope of current knowledge, its limits and validity.
- Participatory research asks, what knowledge is privileged or absent?

Modeling to Learn,

Test don't guess.

Adm Policy Ment Health DOI 10.1007/s10488-016-0754-1



Administration and Policy in Mental Health

Mental Health Services Research

Volume 43 • Number 5 • September 2016







ORIGINAL PAPER

Participatory System Dynamics Modeling: Increasing Stakeholder Engagement and Precision to Improve Implementation Planning in Systems

Lindsey Zimmerman^{1,2} · David W. Lounsbury³ · Craig S. Rosen^{1,4} · Rachel Kimerling¹ · Jodie A. Trafton^{4,5} · Steven E. Lindley^{4,6}

Modeling to Learn

Test don't guess.

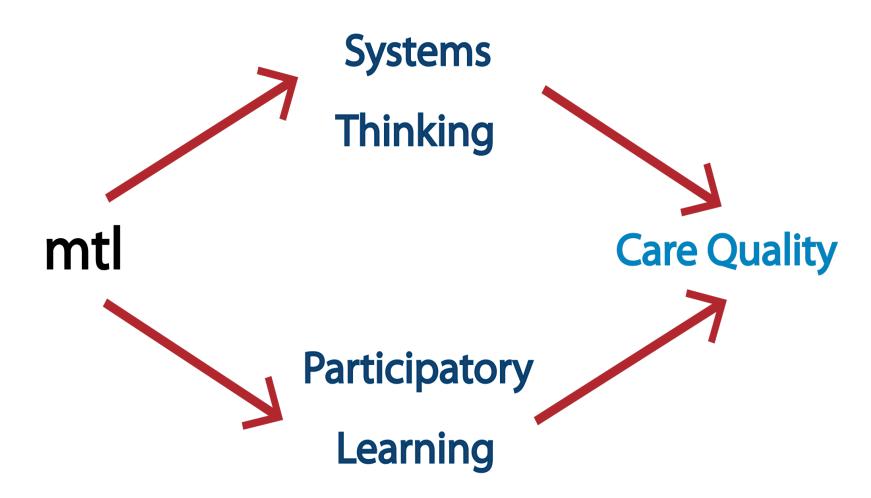
Virtual Facilitation

Transparent Local Data

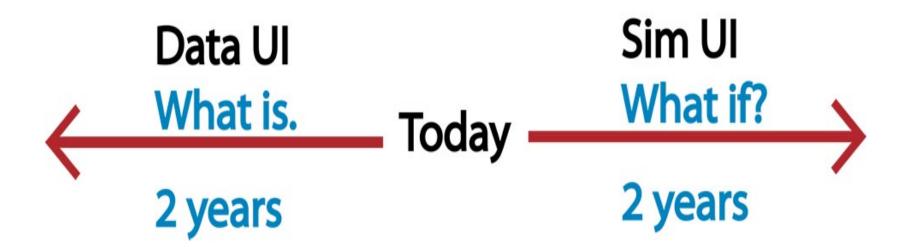
Real-time Simulation

- 1. Equitable access to resources.
- 2. Mutual learning.
- 3. Shared decision-making.

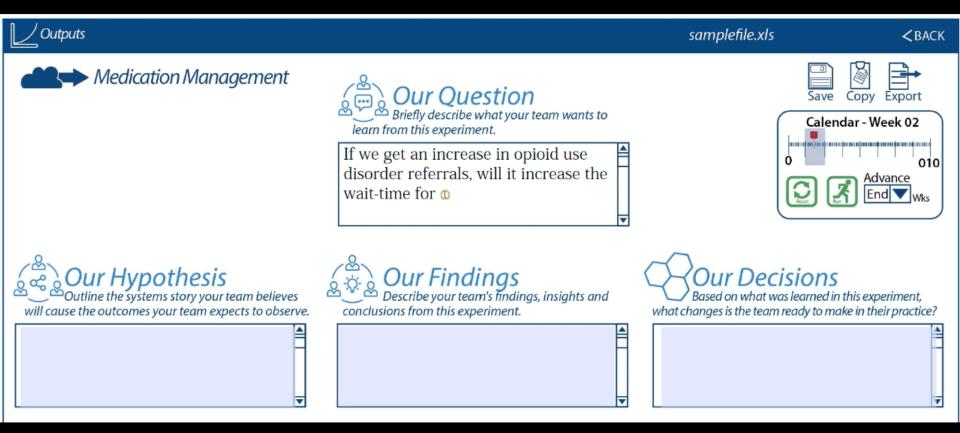
Modeling to Learn Theory of Change



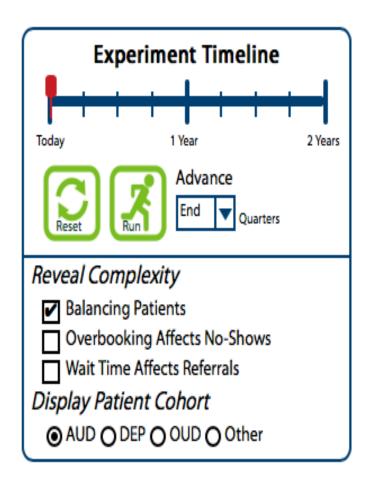
MTL resources help teams look back two years and look ahead two years.



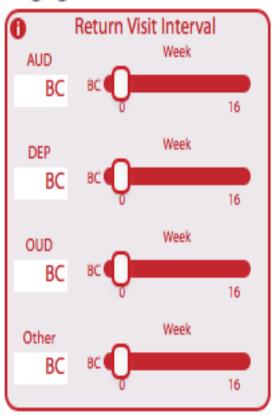
Why is PSD effective? Participatory learning to develop 'Systems Thinking.'



"Staff" and "Time" costs as dynamics.



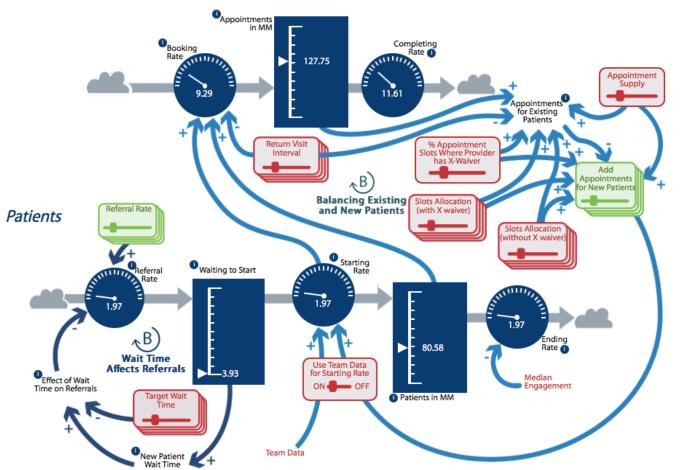
Engagement Pattern





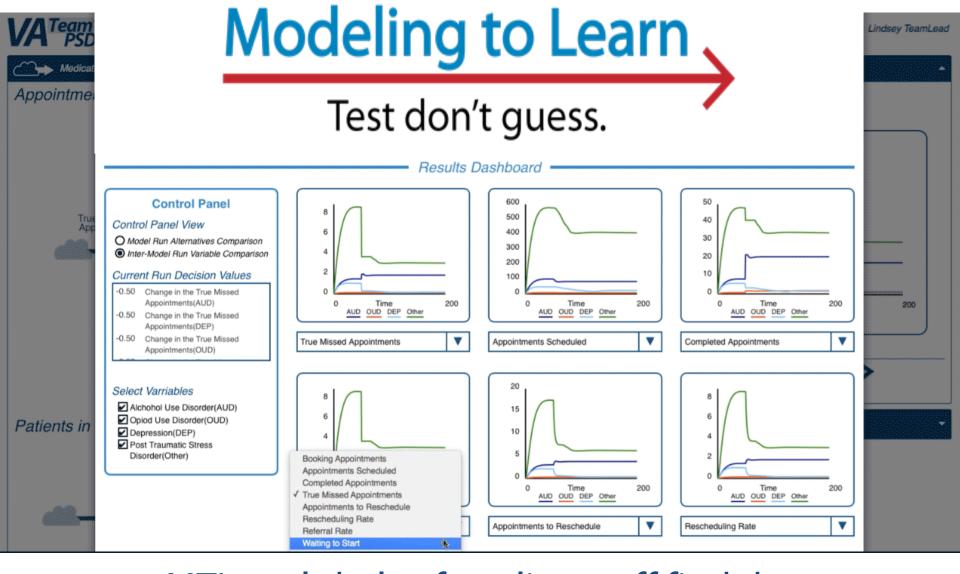


Causal mechanisms (dynamics) are made transparent for local learning.



Red =

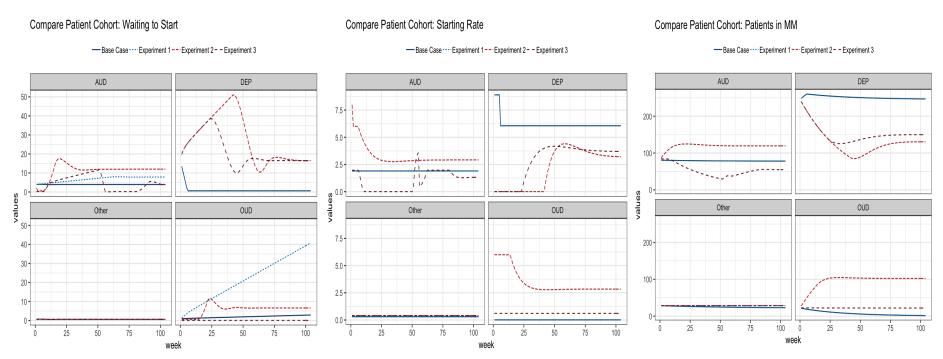
- Read in from existing team data
- Standardized



MTL tools helps frontline staff find the best local changes faster.

mtl.how



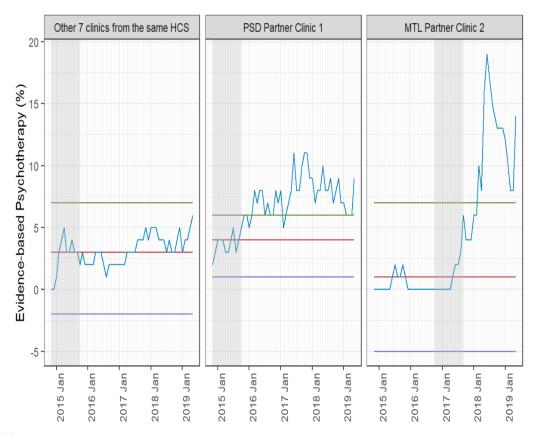


MTL shows whether things may get better before worse or worse before better.

Is PSD/MTL effective for improving EBP reach? Strong signal in R21 pilot clinics.

LCL (first year) — average (first year) — UCL (first year)

*HCS = Regional health care system



36 mos. sustained sig. improvement $+ 3 SD (\alpha = .003)$

20 mos. sustained sig. improvement $+ 3 SD (\alpha = .003)$

Key:

Green = Upper control limit (UCL)

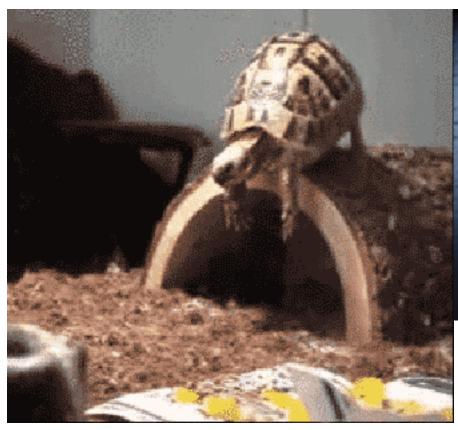
Red = 12-month pre-PSD EBP proportion

Purple = Lower control limit (LCL)

SD = standard deviations

Modeling to Learn

Test don't guess.





Measure twice cut once.









Principles of the open science movement:

- collaborative
- free and open
- transparent and reproducible science.

Modeling to Learn

Test. Don't guess.

Modeling to Learn Links

- 1. www.mtl.how/live Modeling to Learn Live Sessions Adobe Connect Room
- 2. www.mtl.how/data Team Data User Interface **Internal for VHA Providers Only
- 3. www.mtl.how/demo Simulation Demonstration Self-Registration
- 4. www.mtl.how/sim Simulation User Interface for Teams in MTL Live
- 5. www.mtl.how/menu Modeling to Learn Menu RedCap Survey of Team Needs/Priorities
- 6. www.mtl.how/facilitate MTL Facilitator Dashboard at Forio Epicenter
- 7. www.mtl.how/github This page MTL GitHub Repository of Resources
- 8. www.mtl.how/video MTL "How To" videos at YouTube
- 9. www.mtl.how/team Team Participatory System Dynamics The MTL Research & Development Team
- 10. www.mtl.how/lzim MTL and Team PSD Lead Lindsey Zimmerman, PhD
- 11. www.mtl.how/tms VA TMS 2.0 Learning System for Accreditation
- www.mtl.how/refs MTL References
- 13. www.mtl.how/pubs Publications & Presentations on MTL by Team PSD



You can review *Modeling to Learn* session guides at mtl.how



Session guides, links, and cheatsheets.

mtl1.7_models	Update README.md
mtl1.8_models	Update README.md
session01	Update README.md
session02	Update mtl_session02_see.md
session03	Update mtl_session03_see.md
session04	Add files via upload
session05	Update mtl_session05_see.md
session06	Update mtl_session06_see.md
session07	Update mtl_session07_see.md
session08	Update mtl_session08_see.md
session09	Update mtl_session09_see.md
session10	Update mtl_session10_see.md
session11	Update mtl_session11_see.md
session12	Add files via upload
■ LICENSE	Initial commit
README.md	Update README.md



Participatory Learning to develop Systems Thinking.

MTL Fidelity Checklist for 12-session Plan

Sesson Summaries across MTL Modules

session 01. Today we're modeling to learn how to align our team vision.

session 02. Today we're modeling to learn how to check our patient data and team trends.

session 03. Today we're modeling to learn how to produce team data for simulation.

session 04. Today we're modeling to learn how to prioritize team needs.

session 05. Today we're modeling to learn how to log-in to our team world.

session 06. Today we're modeling to learn how to tell a systems story.

session 07. Today we're modeling to learn how to evaluate our base case of no new decisions.

session 08. Today we're modeling to learn how to test a dynamic hypothesis.

session 09. Today we're modeling to learn how to compare alternatives.

session 10. Today we're modeling to learn how to use systems thinking.

session 11. Today we're modeling to learn how to make future team decisions.

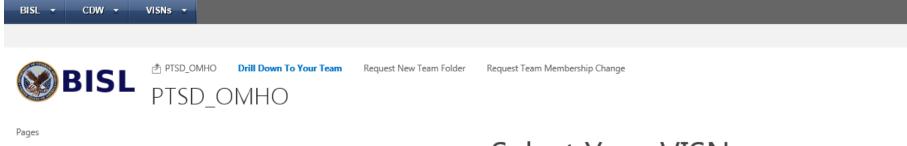
session 12. Today we're modeling to learn how to turn team learning into a team plan.







We developed a secure website for reviewing team trends over time.

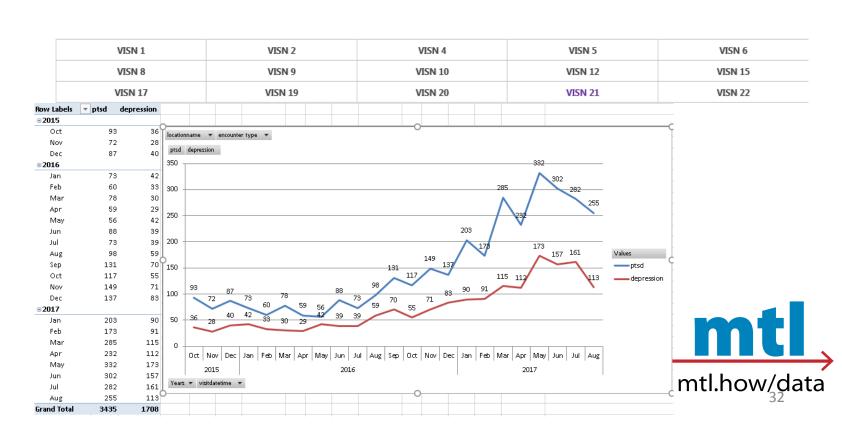


Administrative
User Guide

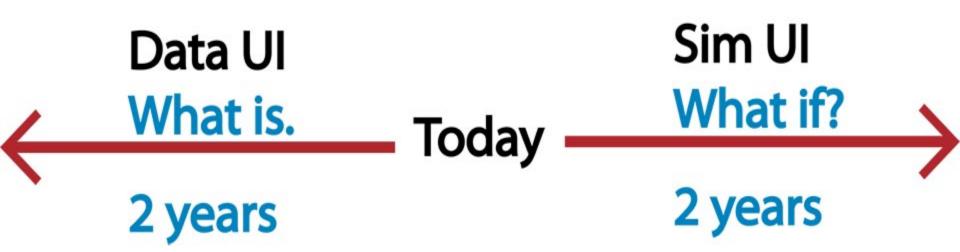
Contact Us

Site Contents

Select Your VISN

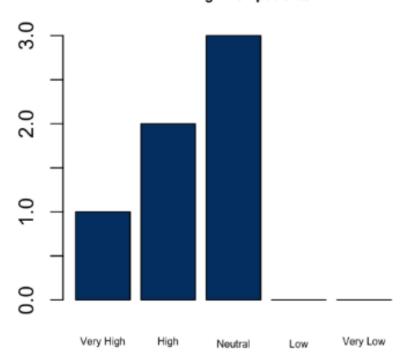


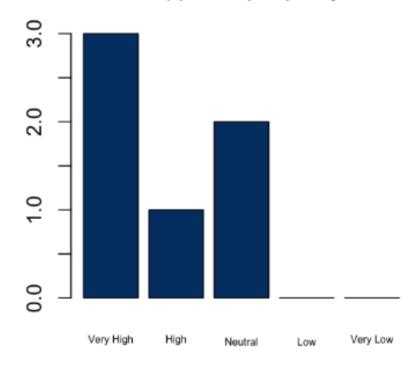
MTL resources help teams look back two years and look ahead two years.



Suicide Prevention - How to manage high risk patients.

Stepped Care - How to decide when to step patients up to specialty care.

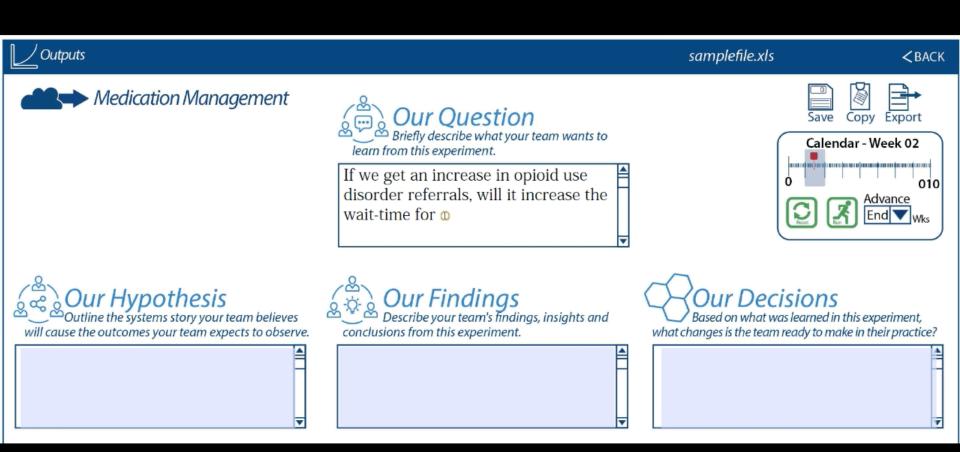








Why is *Modeling to Learn* effective?



Start a new Medication Management Session.



- 1. Review the team data and "i" information.
- Zoom in/out to review system stories and complexity reveals for each care setting.
- Run, examine the output, and save a base case of no new decisions.

Select an MM Learning Mode.

Learning Mode

Refer to the **Medication Management (MM)** tab of the **Team Data Table for Sim UI** at **mtl.how/data**, for data regarding the team's **New Patient Start Rates** and **Return-to-Clinic Intervals**.

Existing Patient Return-to-Clinic Visit Interval

Choose this learning mode to prioritize the **Existing Patients Return-to-Clinic Visit Interval (RVI)** estimated from team data, and only start new patients in remaining open slots after the existing patient RVI, appointment supply, and missed appointments are all taken into account.

O New Patient Start Rate

Choose this learning mode to prioritize the **New Patient Start Rate** estimated from team data, and only see existing patients in remaining open slots after the new patient start rate, appointment supply, and missed appointments are all taken into account.

- Starting a new simulation will stop the previous session for all team members. Session decisions and results may not have
- been saved.

Start

Cancel

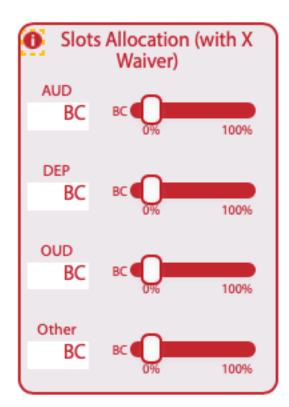
Modeling to Learn helps teams manage tradeoffs within existing staff resources.

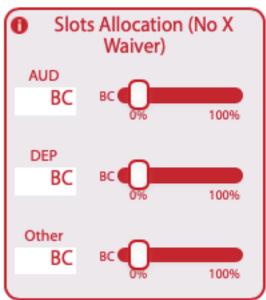
				\otimes			
If Existing Patient RVI is selected, then New Pt. Starts =							
	AUD	DEP	OUD	Other			
pts/wk	1.97	6.06	0.61	0.4			
If New Patient Start Rate is selected, then RVI =							
	AUD	DEP	OUD	Other			
wks	10.47	23.42	0.11	9.56			

Can we increase the number of patients with OUD that receive EBPharm without increasing wait-times for other patient needs?

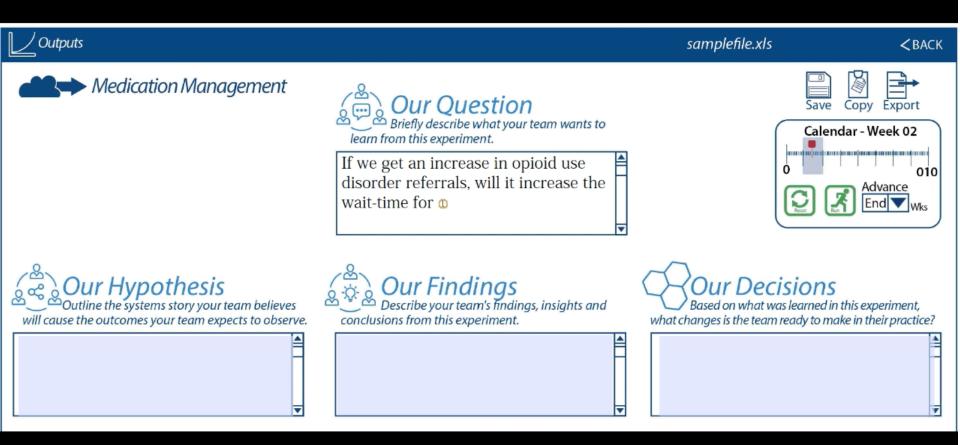
1 Team Data					
Appointment Supply (75th Percentile) (appt/wk)				60	
Appointment Slots % (with X waiver)				50	
	AUD	DEP	OUD	Other	
Engagement Duration (median) (wks)	41	40.5	36	70.5	
Return Visit Interval (median) (wks)	11	12	11	14	
True Missed Appointment (%)	20	21	0	23	
Start Rate (mean) (pts/wk)	1.9	8.9	0.01	0.3	
New Patient Wait Time (median) (wks)	2	2.7	1.1	1.5	
Slots Allocation % (with X waiver) (%)	16	71	10	3	
Slots Allocation % (without X waiver) (%)	26	63	0	11	
🖰 🚳 Learning Mode					

Not all medication management staff resources are the same.



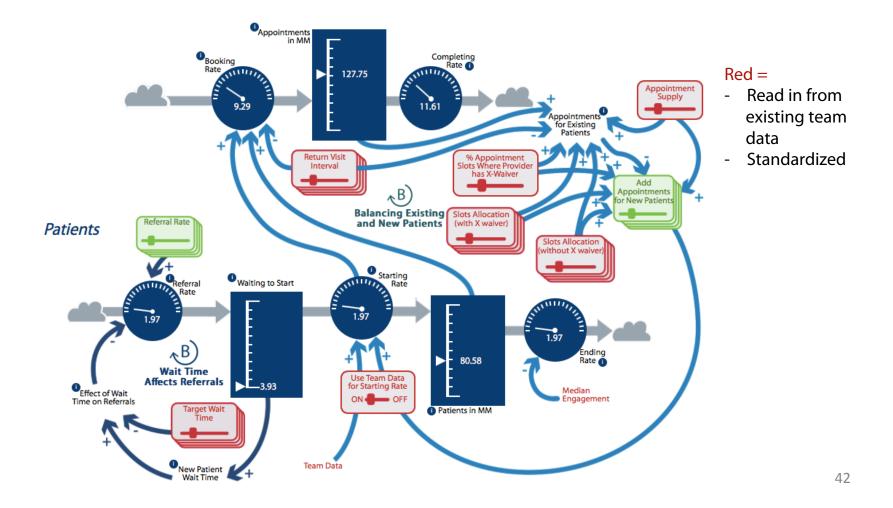


<u>Team Question</u>: Can we increase the number of patients with OUD that receive EBPharm without increasing wait-times for other patient needs?



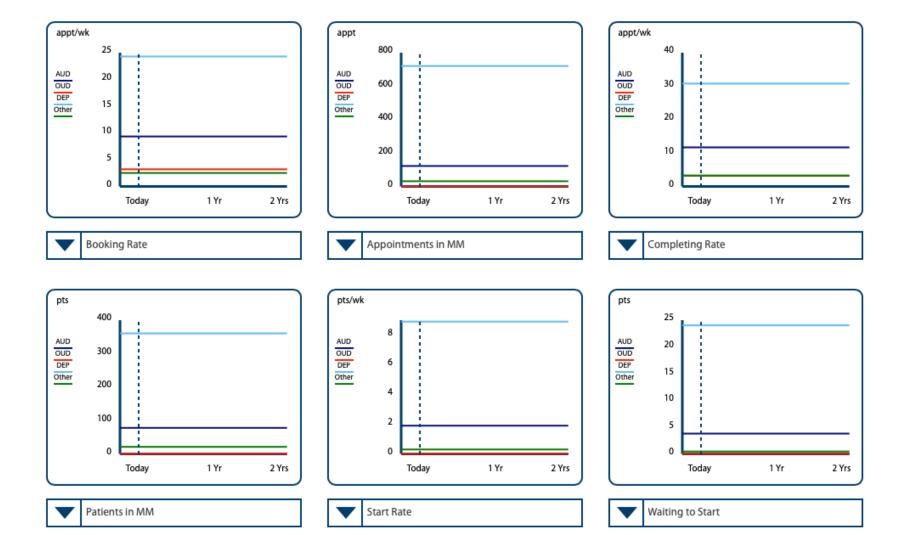
Hypothesis: If we make no new decisions in our team, then...

1. we will continue to underserve patient in our community who may benefit from OUD EBPharm, and 2. we won't help as many patients toward recovery as we would like.



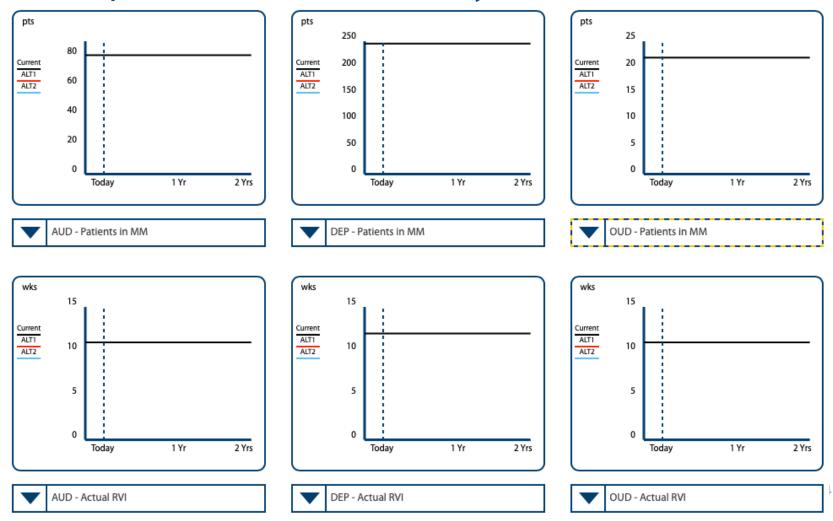
What if we made no new decisions?

Basecase



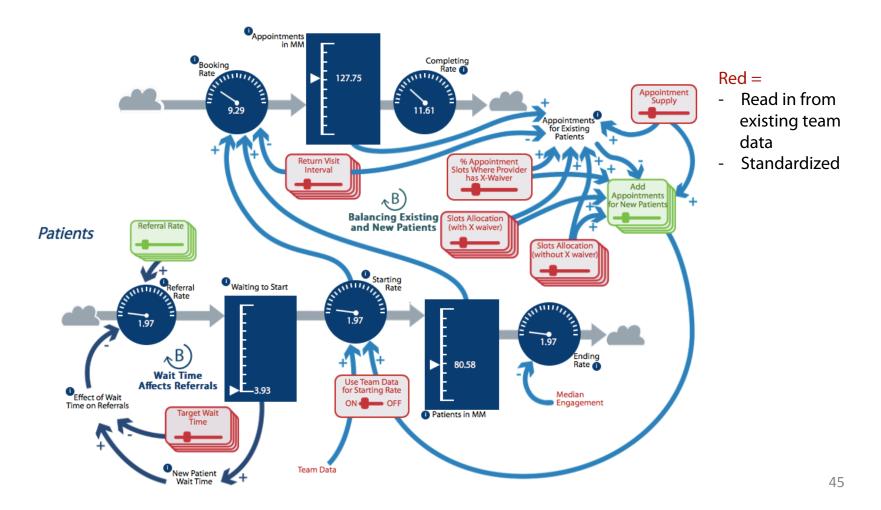
<u>Findings</u>: If we make no new decisions, then...

- 1. we continue to primarily serve patients with depression
- 2. all patients come back every 10-11 weeks

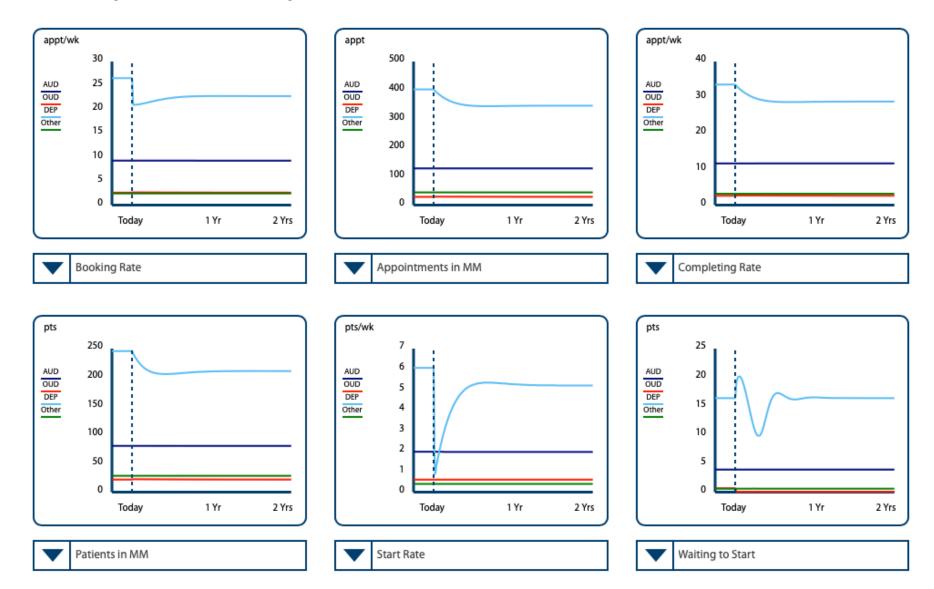


Hypothesis about Re-allocating X-waiver slots:

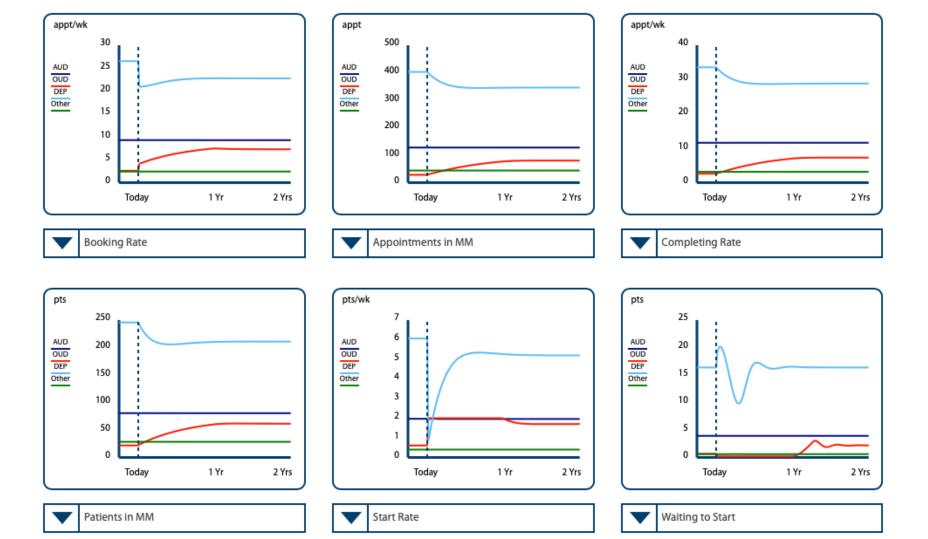
- we could start more patients with OUD on EBPharm
- but we expect more patients with depression and AUD will be waiting to start than in the base case.



Re-allocating 20% of x-waivered slots from patients w/depression to patients w/OUD levels out over time.



With two new referrals each week we can triple the number of patients with OUD in our team who receive EBPharm over the next two years.



MTL focuses on improving systems thinking among frontline teams making care decisions.

Systems Thinking

Definition

improvement rate), or two or more settings (primary

Better understanding of change over time

(e.g., worse before better, better before worse).

Forest not trees. Relationships among two or more variables (wait times,

Complex

care, general mental health).

Feedback

Time

Trends over time. Systems cause their own behavior through feedback.

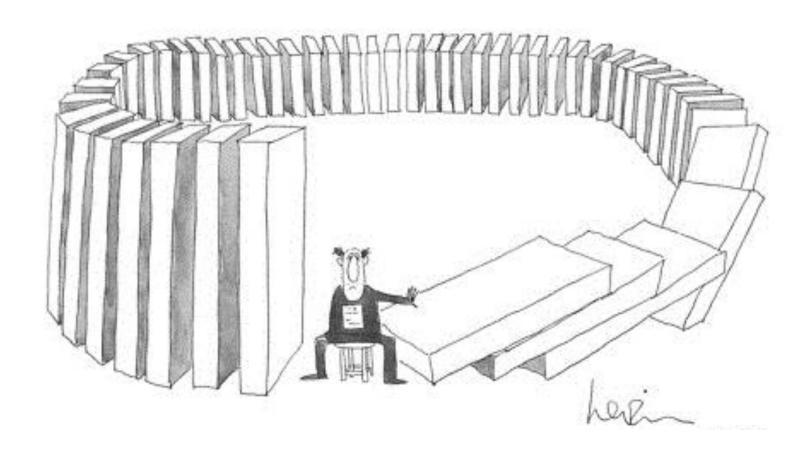
System Behavior

Short and long term.

Loop not line. Not simple cause and effect. The end of the story often influences the beginning, and is strengthened (reinforcing) or reduced (balancing) around the loop. Movie not snapshot.

Decisions based on *Modeling to Learn* experiments:

Something that we think is outside of our control may actually be the accumulated result of our own decisions.





NIDA R01DA04665, R21DA042198, HSRD I01HX002421 Pl: Zimmerman

R21 Co-Investigators

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Resources and Help



Session guides, links, and cheatsheets.

Self-registration for simulation demo. *Course code: ahsr2019*



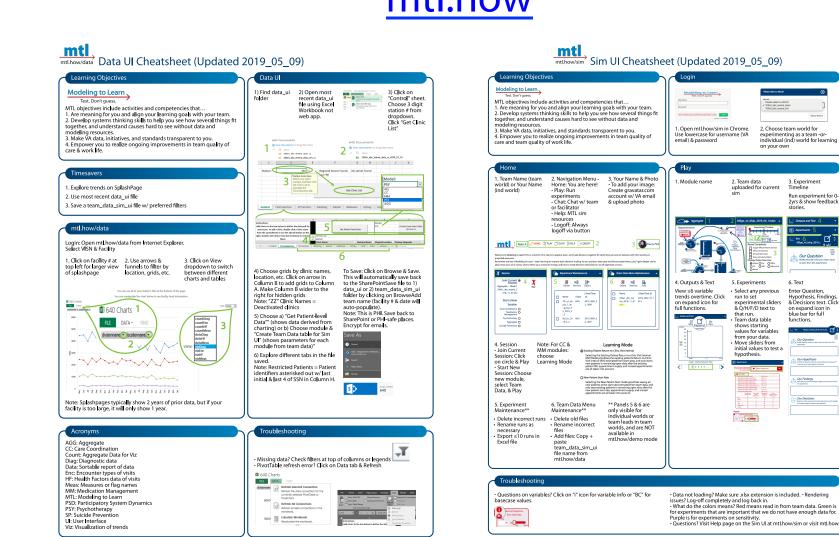


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Download 1-page *Modeling to Learn* Cheatsheets at mtl.how



Modeling to Learn (MTL) is an ongoing quality improvement project by VA Team Participatory System Dynamics. Info: mtl.info @va.gov Updated

2. Choose team world for

experimenting as a team -or

individual (ind) world for learning

Run experiment for 0-2yrs & show feedback

Enter Question,

Hypothesis, Findings, & Decisions text. Click

on expand icon in

functions.

You can check out our demonstration simulation website.



Session guides, links, and cheatsheets.



Self-register

Course Code: ahsr2019

Run Your Test



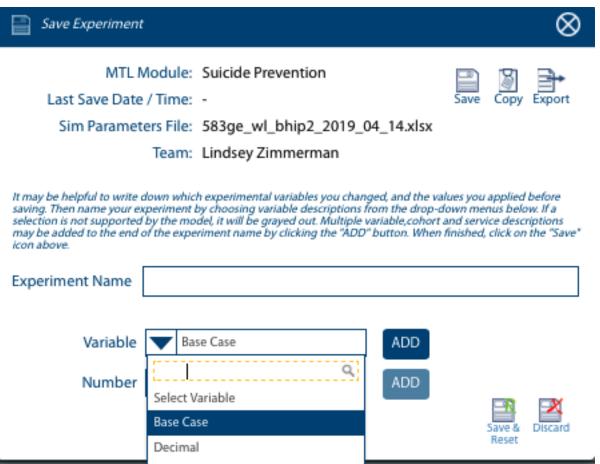
Click the icon to run your own simulation.

	provide some information			
	e can send you a login			
credentials. You will have according to Course Code. Unless you cho	mail and password will only be used to create your login ess to the simulation for 5 days, unless you were given a pose to continue to receive updates about the Modeling system will erase your information after 5 days.			
*First Name?				
*Last Name?				
*Your email?				
*Create a Password?				
*Confirm your password?				
*Your Institution?	ease select			
If other please specify >				
*Your Role? Please se	elect			
If other please specify >				
*Your Discipline?	ease select			
If other please specify >				
*How did you find us?	Please select			
If other please specify >				
Enter your course code >				
	I do not need a course code. However, some users may be course codes for specific trainings.			
Yes No Would you like email updates about <i>Modeling to Learn</i> quick tips and new releases?				
	Run			





*Once registered go to: mtl.how/demo_login



Help is available in top navigation bar.

















Model Diagram	Experiment Timeline	Outputs	Experiment
The blue header at the top shows the module and data file chosen. The rates (circles) and stocks (rectangles) update dynamically with changes in the experiment variables. Throughout the model diagram, there are "I" icons to explain how the variable is calculated.	Use reveal complexities to look at balancing and reinforcing feedback systems stories. R B B B In the systems stories, there are two kinds of arrows. Plus signs mean trends move in the same direction. Minus signs mean trends move in the opposite direction.	Text or Q/H/F/D Enter Question, Hypothesis, Findings, and Decisions text for each experiment. Expanded Outputs View Q/H/F/D Text and Results Dashboard at once Expand Results Dashboard View trends over time for ≤6 variables. Compare ≤2 experiments against current run.	Select Experiment Select previous experiments to cue up experiment values and q/h/f/d text from previous experiments. Team Data Table Shows initial starting values of experimental variables based on team data. Experiment Adjust experiment sliders to test different values in the sim by dragging the slider. Appointment Supply OOD BC O Appts/Week 200



Five ways to improve MTL usefulness.

Email: mtl.info@va.gov

Subject line: Learning

- 1. MTL Live Team/Clinic
- 2. Pilot Review EES materials (e.g., Video, Guides)

Design

- 3. Data User Interface (mtl.how/data)
- 4. Simulation User Interface (mtl.how/demo)

Research

5. Advisory Board and other opportunities

References

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