



# Multilevel, Adaptive, Implementation Strategies (MAISYs)

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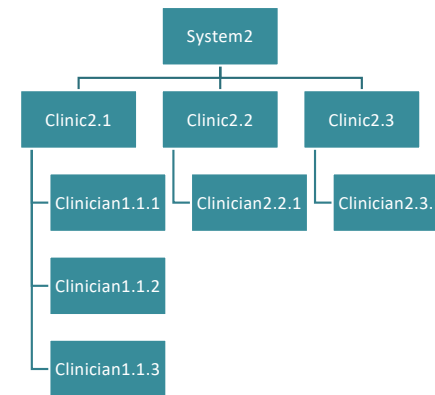
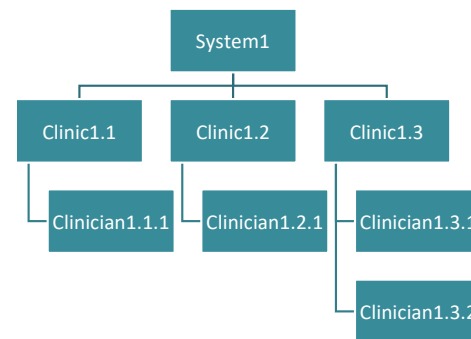


## Outline

- Implementers Have to Make Many Decisions @ Many Levels
- Multilevel Adaptive Implementation Strategies  
What? Why? Who?
- Using Randomization to Construct an Optimized MAISY

# Multiple Decision Levels

@System Level  
@Clinic Level  
@Clinician Level



# Determinants to Implementation at Multiple Levels

Evidence-based practices fail to be implemented or sustained due to barriers at multiple levels. For example,

@System Level	Ineffective communication, monitoring practices, policies
@Clinic Level	Lack of support, workflow processes
@Clinician Level	Lack of skills

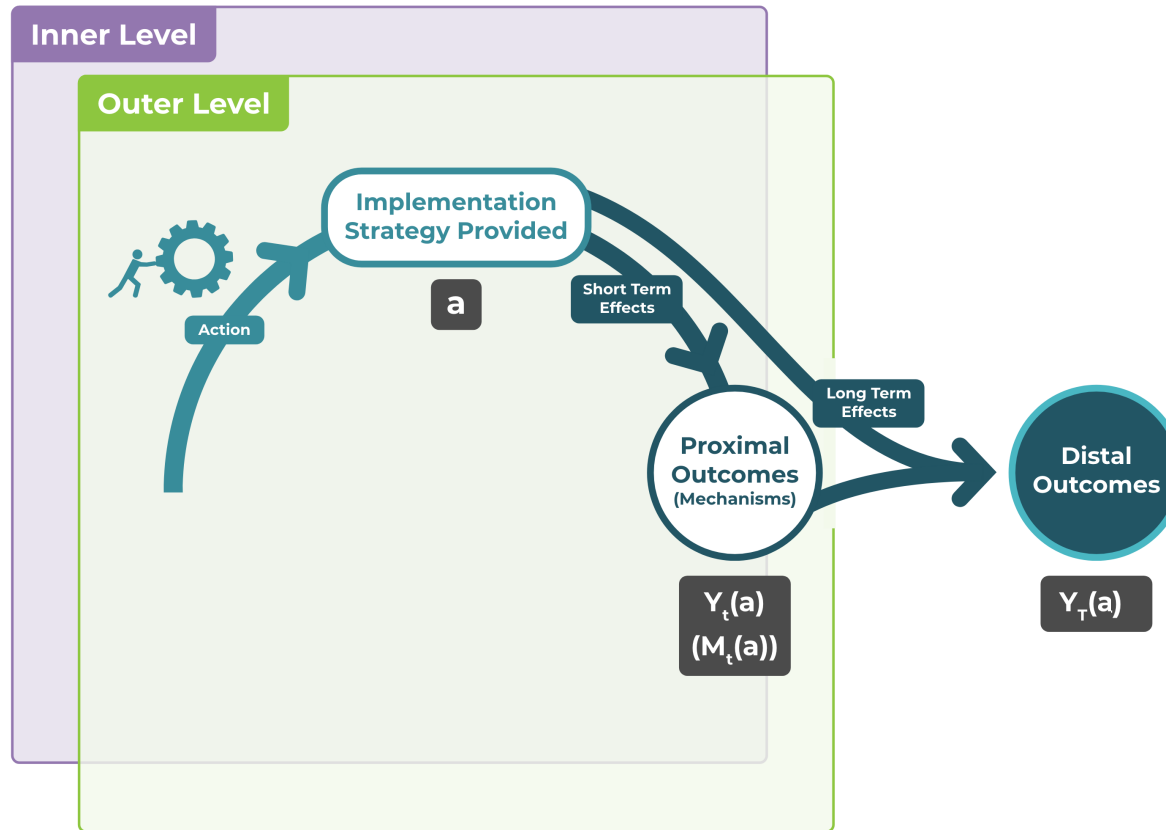
## Implementation Strategies at Multiple Levels

A growing cadre of implementation strategies can help mitigate these challenges. For example,

@System Level	<b>Audit &amp; Feedback<sub>S</sub></b>	Ineffective monitoring practices
@Clinic Level	<b>Facilitation<sub>CC</sub></b>	Lack of support
@Clinician Level	<b>Coaching<sub>CN</sub></b>	Lack of skills

# Quick Review: What is an Implementation Strategy?

- Implementer
- Levels
- Targets
- Action
- Outcomes
- Rationale



Proctor, Powell, McMillen (2013), *Impl Sci*

## From the Perspective of the Implementer

**What works for one target may not work for another target**

Between-target Heterogeneity

**What works in the short-run may not work in the longer-run, or vice-versa**

Within-target Heterogeneity

# From the Perspective of the Implementer



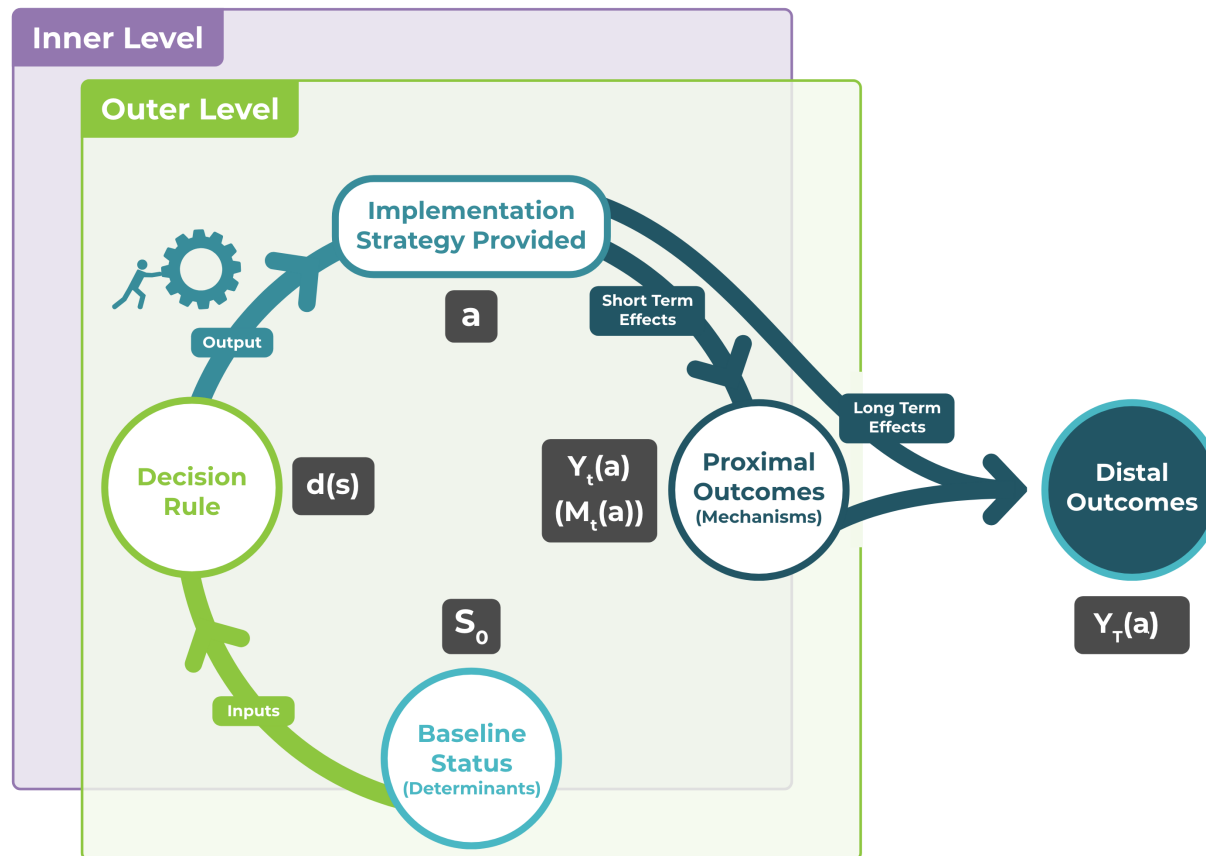


## From the Perspective of the Implementer



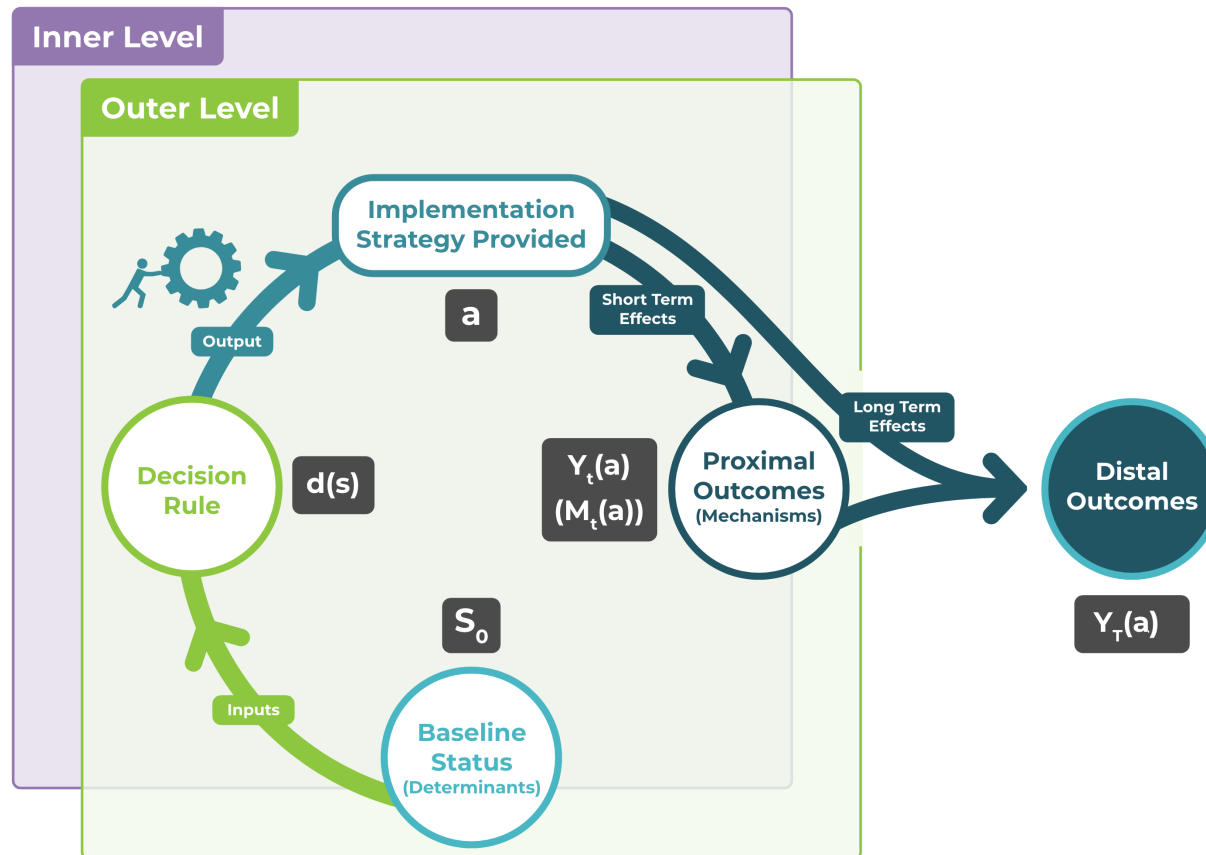
But a Decision is Not Just Any Action

# What if we do this, instead?



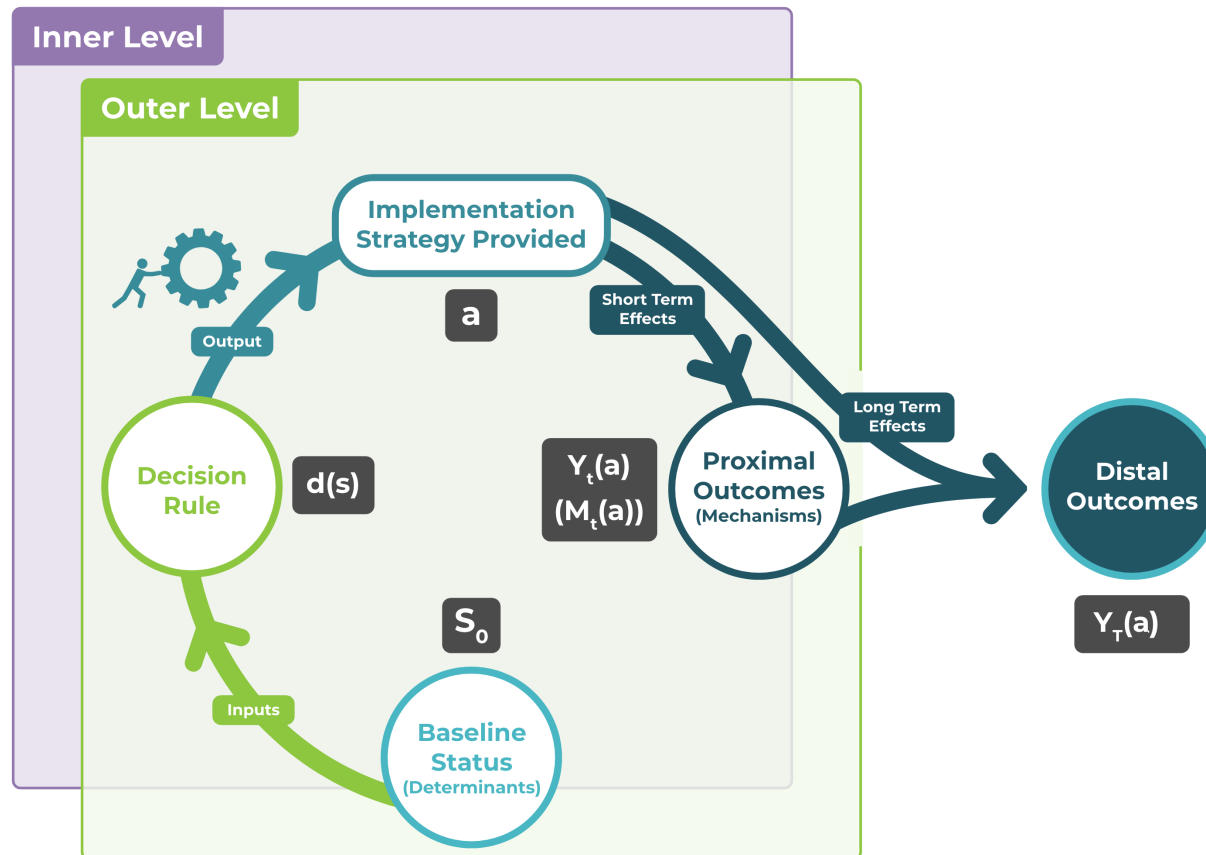
# This strategy has these extra components

- Implementer
- Levels
- Targets
- Action Options
- Baseline Status
- Decision Rule
- Outcomes
- Rationale



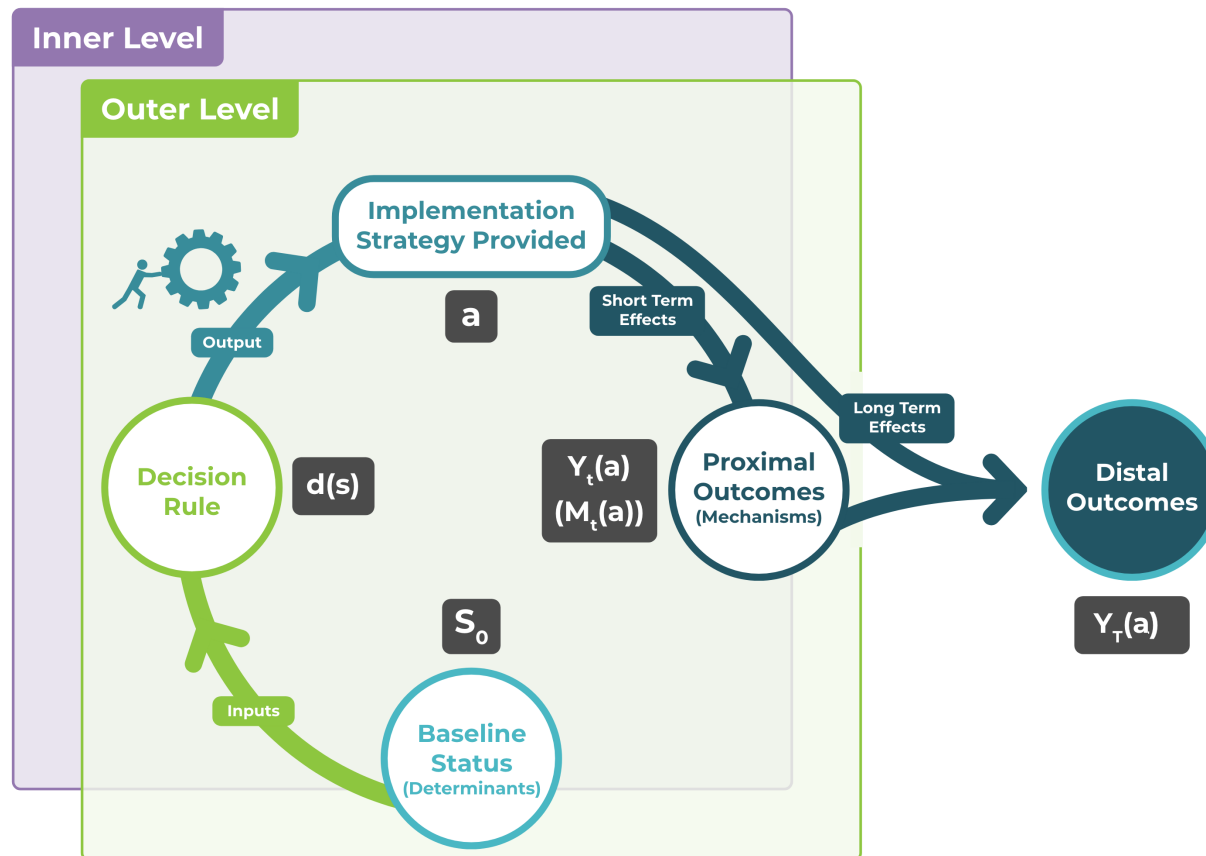
# This strategy has these extra components

- Implementer
- Levels
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- Action Options
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The “determinant” here is the measure (the variable), which takes on different values.

Between-target Heterogeneity @ Baseline ✓  
Within-target Heterogeneity ✗



In many settings, what implementers will need is

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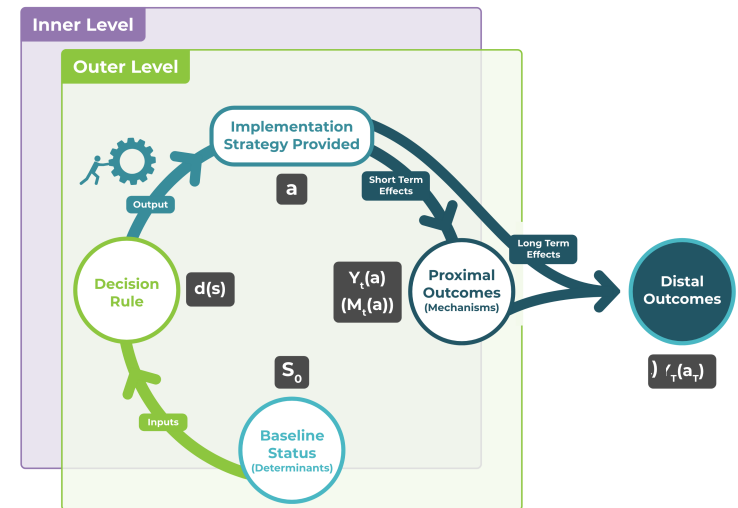
In many settings, what implementers will need is a **practical & intelligent guide** for how best to provide strategies across **multiple levels** and **multiple phases**.

One that guides how best to adjust strategies given **both baseline and ongoing needs** of targets at the multiple levels of implementation.



# Outline

- Many Decisions at Many Levels
- Let's Close this Loop using a Multilevel Adaptive Implementation Strategy
- Using Randomization to Construct an Optimized MAISY

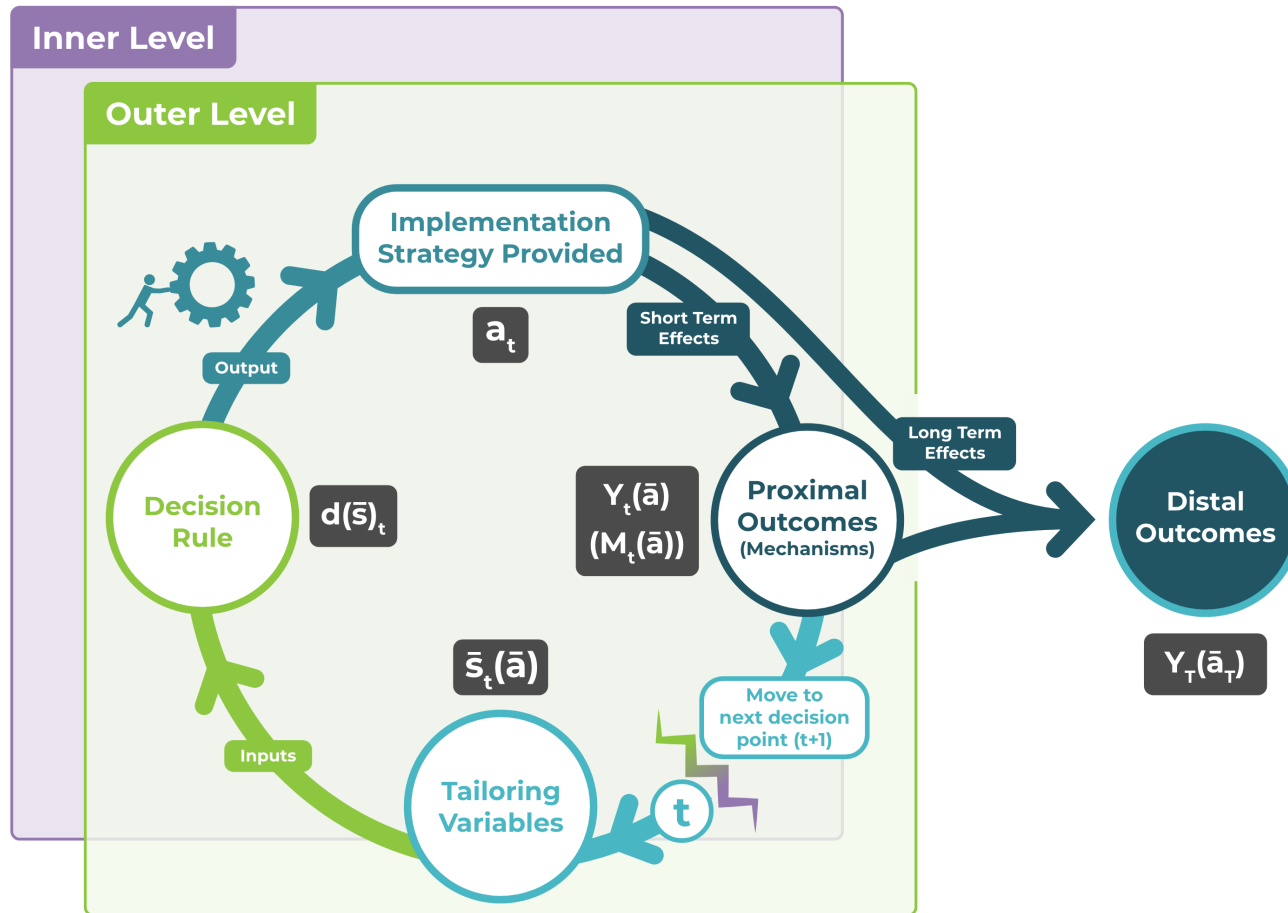


# Multilevel Adaptive Implementation Strategy (MAISY)

A MAISY is a sequence of decision rules used to guide how best to adapt the provision of implementation strategies

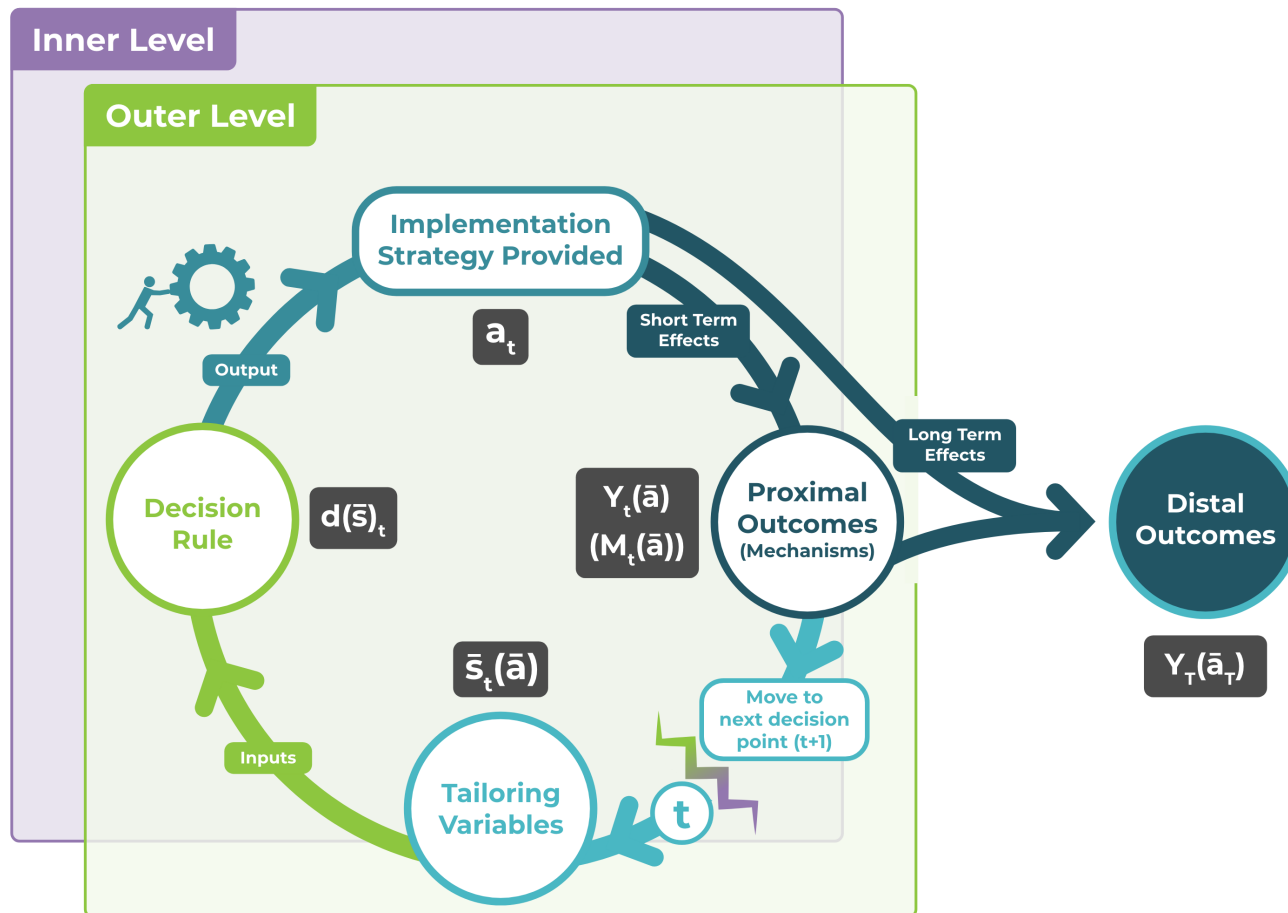
- (i) at critical decision points,
- (ii) across multiple levels,
- (iii) based on **both baseline and ongoing/changing status** of the targets in an organization.

# Multilevel Adaptive Implementation Strategy (MAISY)



# Multilevel Adaptive Implementation Strategy (MAISY)

- Implementer
- Decision Levels
- Targets
- Decision Points
- Action Options
- Tailoring Vars
- Decision Rules
- Outcomes
- Rationale

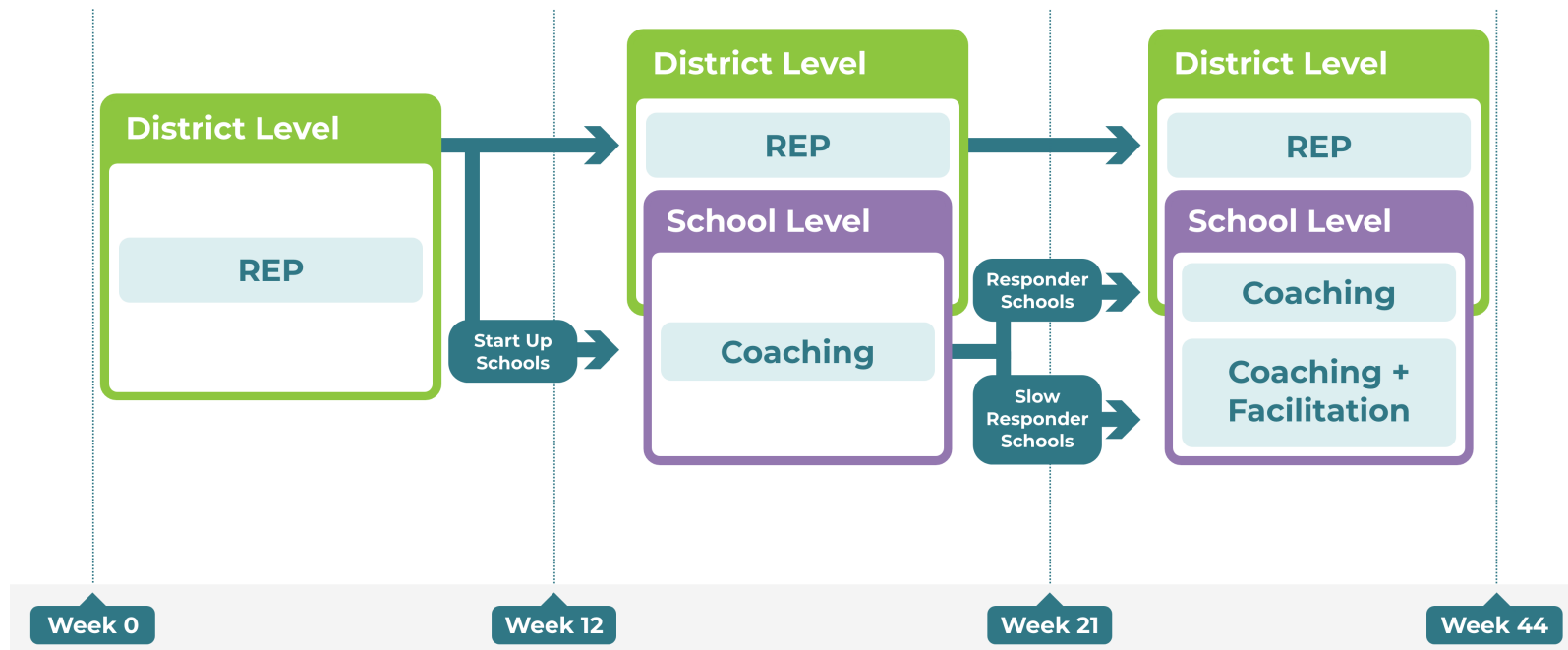


# MAISY Example #1

## Adaptive School-based Implementation of CBT (ASIC)

EBP: Cognitive Behavioral Therapy in Michigan Schools

Developer: Amy Kilbourne



### Start-up School:

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A school with school professionals who do not have training in CBT or have never provided CBT to any of their students.

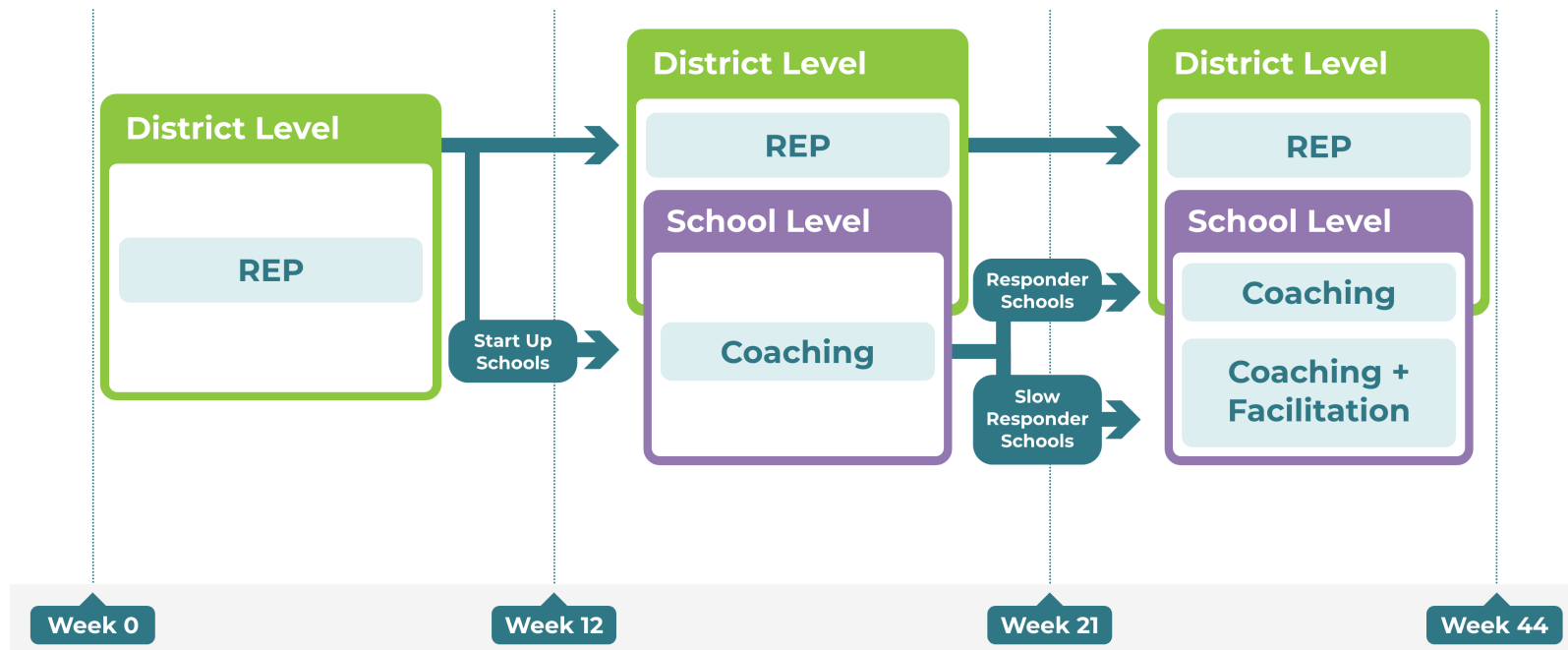


# MAISY Example #1

## Adaptive School-based Implementation of CBT (ASIC)

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### Slow-responding School:

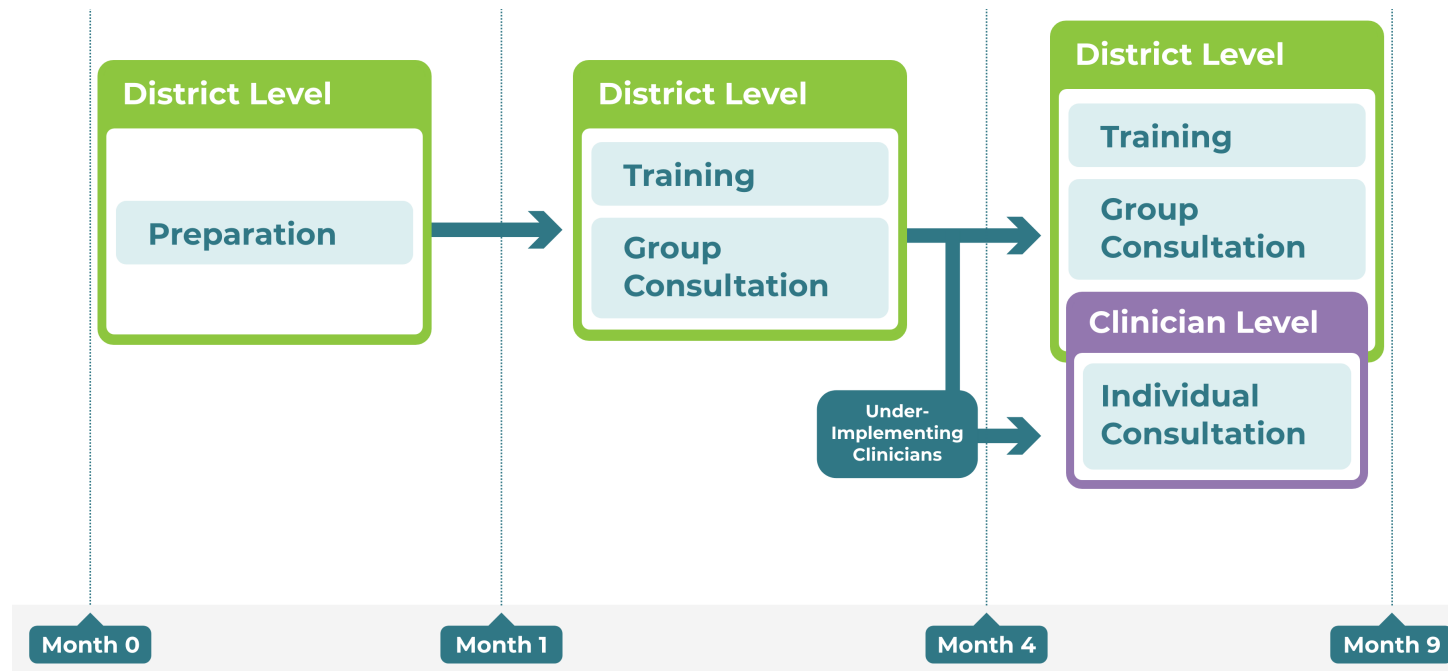
- (i) Any SP reports not providing 3+ CBT components to >10 students
- OR
- (ii) SPs report >2 barriers to CBT delivery (on average)

# MAISY Example #2

## Feedback & Outcomes for Clinically Useful Student Svcs (FOCUSS)

EBP: Measurement-based Care in Connecticut Schools

Developer: Elizabeth Connors



23

Under-implementing Clinician:  
Collected 1+ outcome measure on <40% of students served in first 4 months



## Why MAISYs?

### **Timing is important**

Speed of adoption varies; not all targets are ready to take on more

### **Strategic sequencing**

Lay a strong foundation for subsequent strategies, if needed

### **Health equity**

MAISYs are consistent with “vertical health equity” principles





## Why MAISYs?

### **Engagement is critical**

In short-run, not just about fidelity or quant. of implementation

### **Often, more is not better**

Kitchen sink strategies can lead to suboptimal implementation

### **Resource/Cost efficiency**

Step-up for targets that need it; step-down for targets doing well

## MAISYs are Guides for Implementers

- Implementation practitioners
- Community service providers
- Policy makers
- Clinical leaders
- ~~Researchers~~

## MAISYs are Guides for Implementers, **not Researchers**

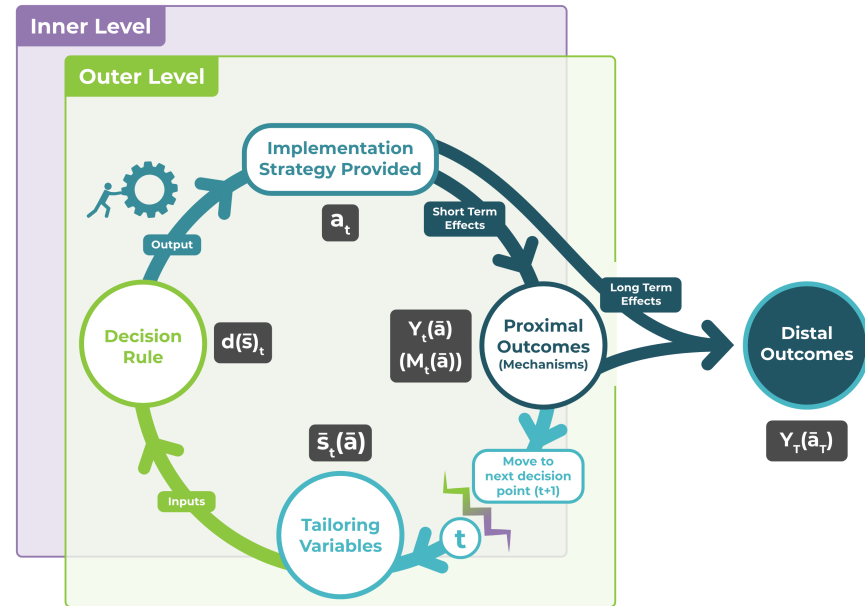
- Implementation practitioners
- Community service providers
- Policy makers
- Clinical leaders
- ~~Researchers~~ unless the Researcher happens to be in the role of the Implementer for purposes of the study, but let's hope the clinics and practitioners perceive it this way

## A MAISY is not a Research Method

- Not an experimental design
  - There are no researchers in a MAISY
  - There are no randomizations
- Not an approach to conducting pilot studies
- Not an approach to data analysis
- Not an adaptive trial design

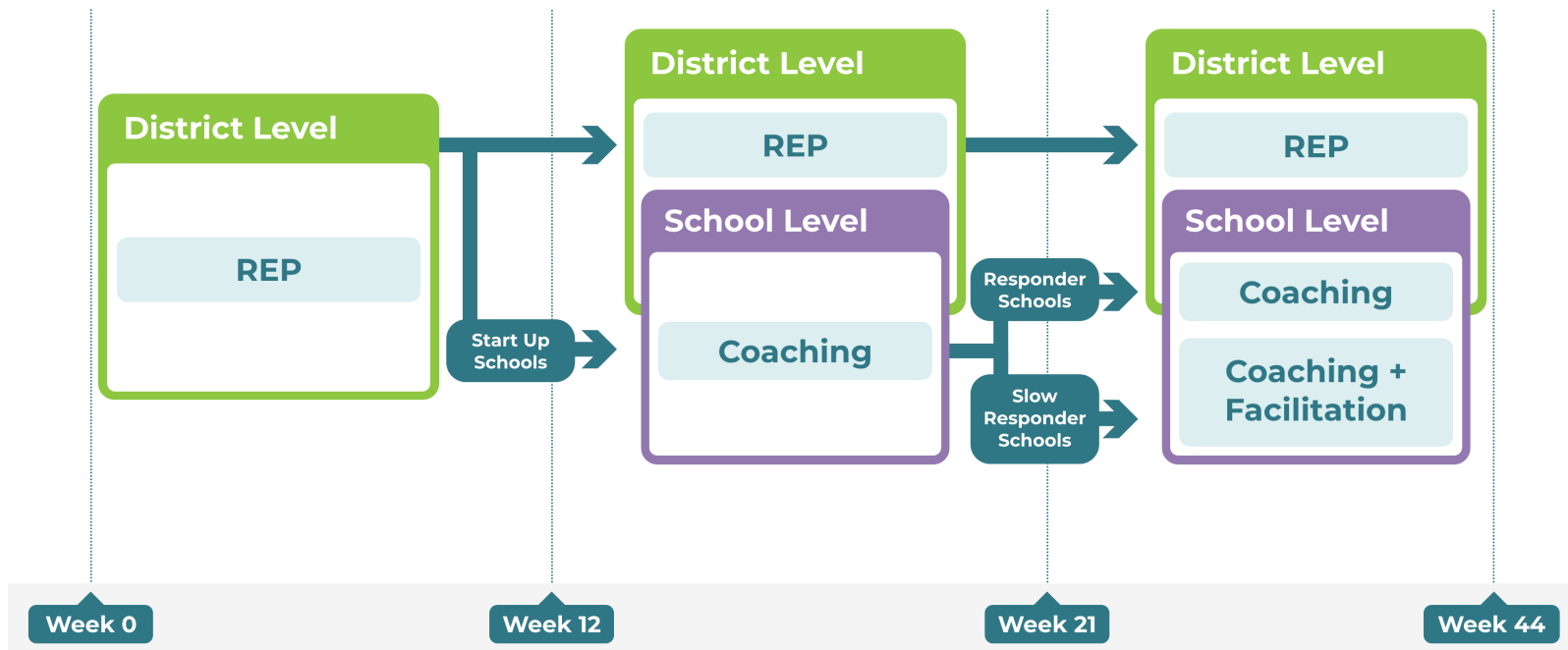
# Jargon Buster Slide (Babel)

- Special case of what Byron Powell calls a “multifaceted multilevel implementation strategy”
- Special type of “adaptive implementation strategy”
- Certainly, MAISYs falls within the realm of “precision implementation strategies”



# Recall MAISY Example 1

## Adaptive School-based Implementation of CBT (ASIC)



## Other Considerations

- Pre-specified (pre-planned)
- Mechanisms can be tailoring variables!!
- The tailoring variables are part of the MAISY
- Caution against conflating MAISYs and “adaptation”

# Outline

- Implementers Have Many Decisions to Make
- Multilevel Adaptive Implementation Strategies  
What? Why? Who?
- Developing an Optimized MAISY

**This is all about asking Optimization Questions  
I am going to show you 13 of these.**



## Outline

- Implementers Have Many Decisions to Make
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What? Why? Who?
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So put your researcher hats back on!



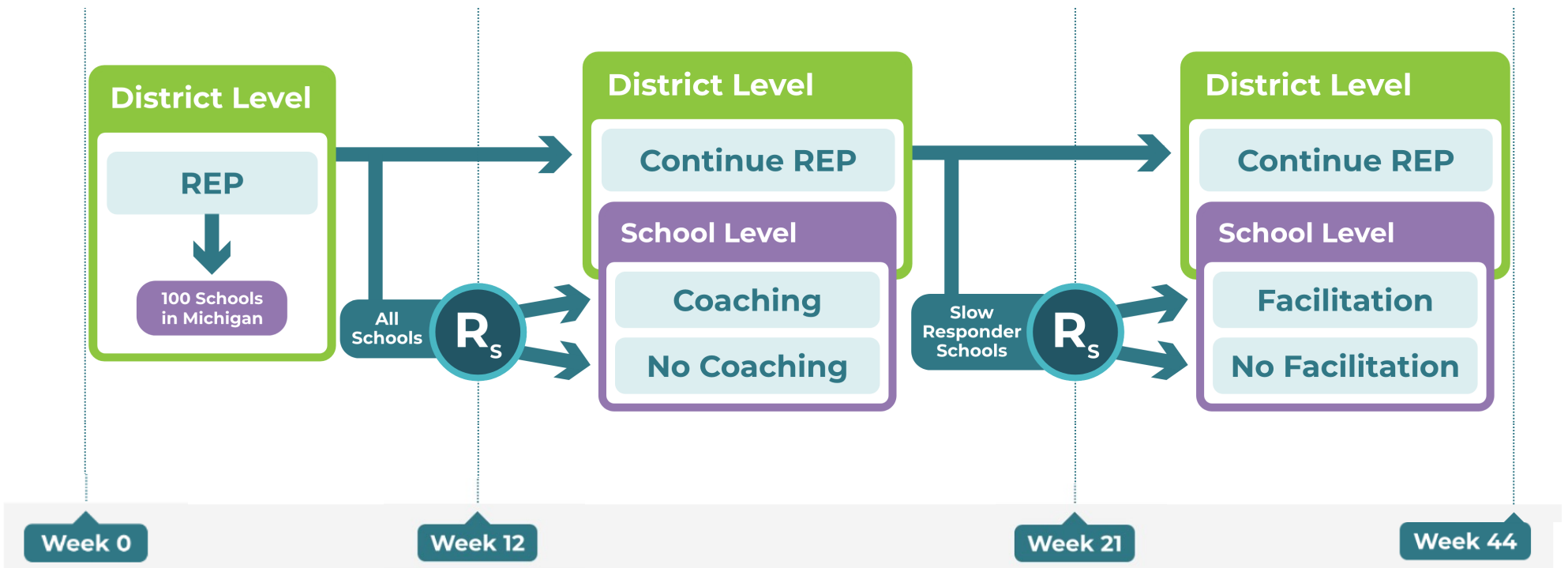
## Optimization Questions: Basic, but important

	Type	In the context of ASIC
1	First stage strategies	What is the effectiveness of Coaching?
2	Later stage strategies	What is the effect of Facilitation among schools that are slower responders?
3	Interaction	Do Coaching and Facilitation interact to produce beneficial outcomes?
4	Adaptive versus not adaptive	What is the effect of the MAISY shown earlier vs only Coaching (not adaptive)?

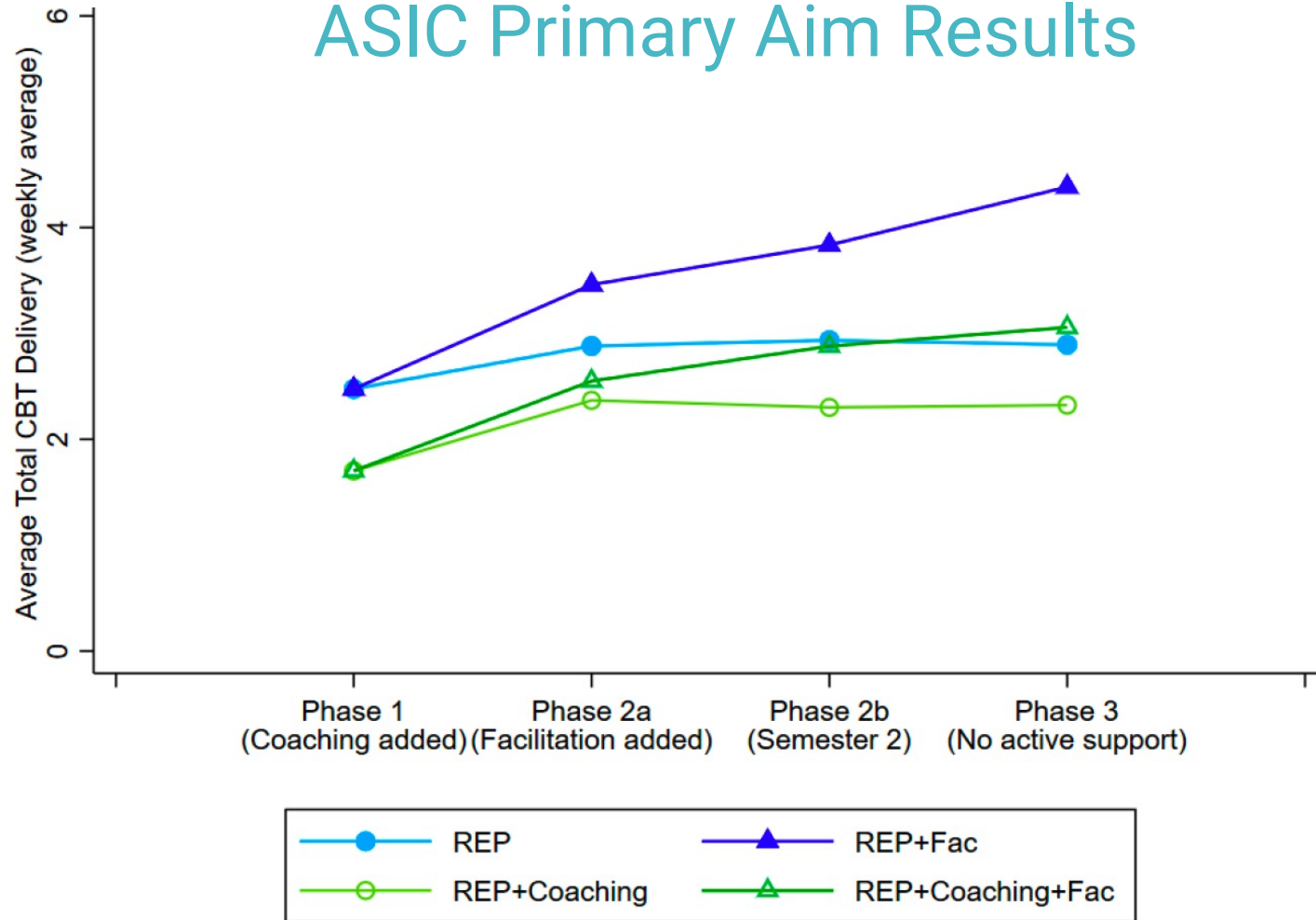
# Sequential Multiple Assignment Randomized Trial

The ASIC SMART

PI: Amy Kilbourne



# ASIC Primary Aim Results



## Optimization Questions: All about tailoring

	Type	In the context of ASIC
5	Better way to define non-response?	Should we use a more lenient definition (a lower cut-off) for “Responding School”?
6	Other baseline tailoring variables?	Perhaps only start-up schools require Coaching?
7	Other ongoing tailoring variables?	Perhaps Facilitation should only be offered to sub-optimally responding schools that did not engage in Coaching?

## Optimization Questions: More about tailoring

	Type	In the context of ASIC
8	Other multilevel tailoring variables?	Perhaps Facilitation should only be offered to sub-optimally responding schools within the lowest resourced school districts?
9	Is the putative mechanism aide in decision making?	Is Facilitation necessary in sub-optimally responding schools delivering higher-quality CBT as a result of Coaching?

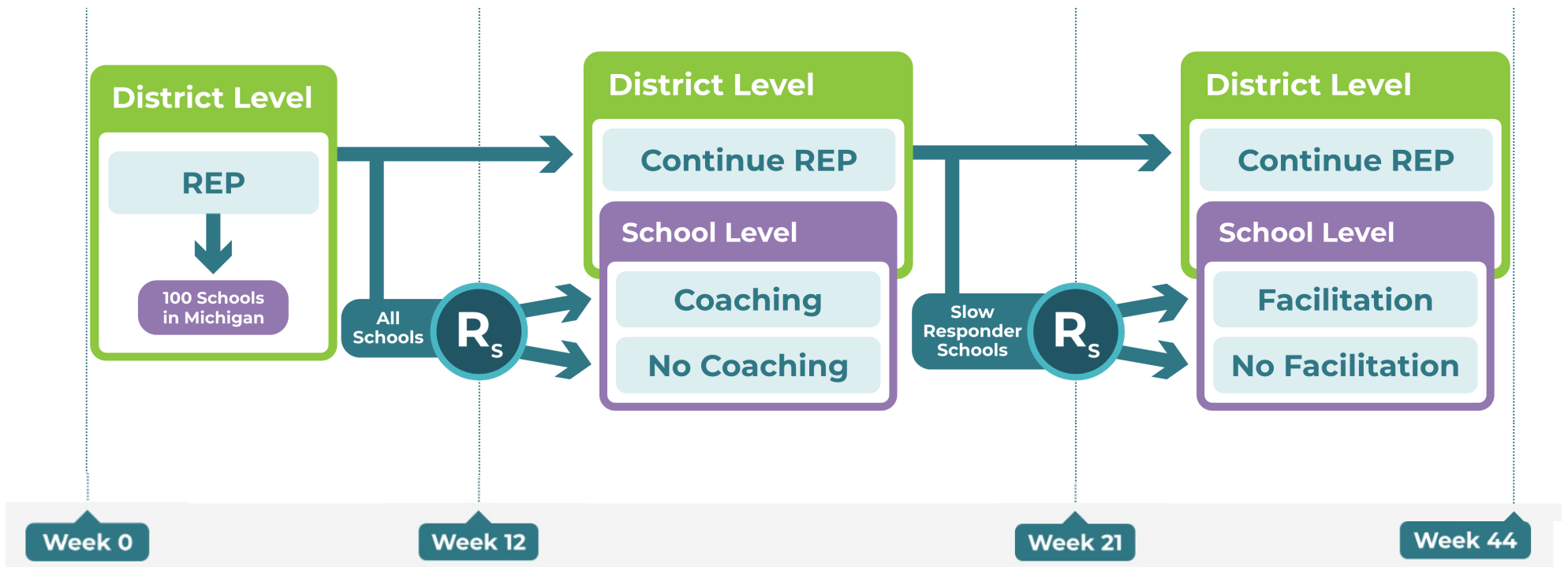
## Optimization Questions: Some novel ones

	Type	In the context of ASIC
10	Sleeper effects of prior stage strategies?	Is it possible that first-stage strategies have no effect in the short-run, but have beneficial effects in the long-run when followed by a particular second-stage strategy?
11	Prescriptive effects?	Did we learn something from Coaching that can help decide whether to do Facilitation?

# Sequential Multiple Assignment Randomized Trial

The ASIC SMART

PI: Amy Kilbourne



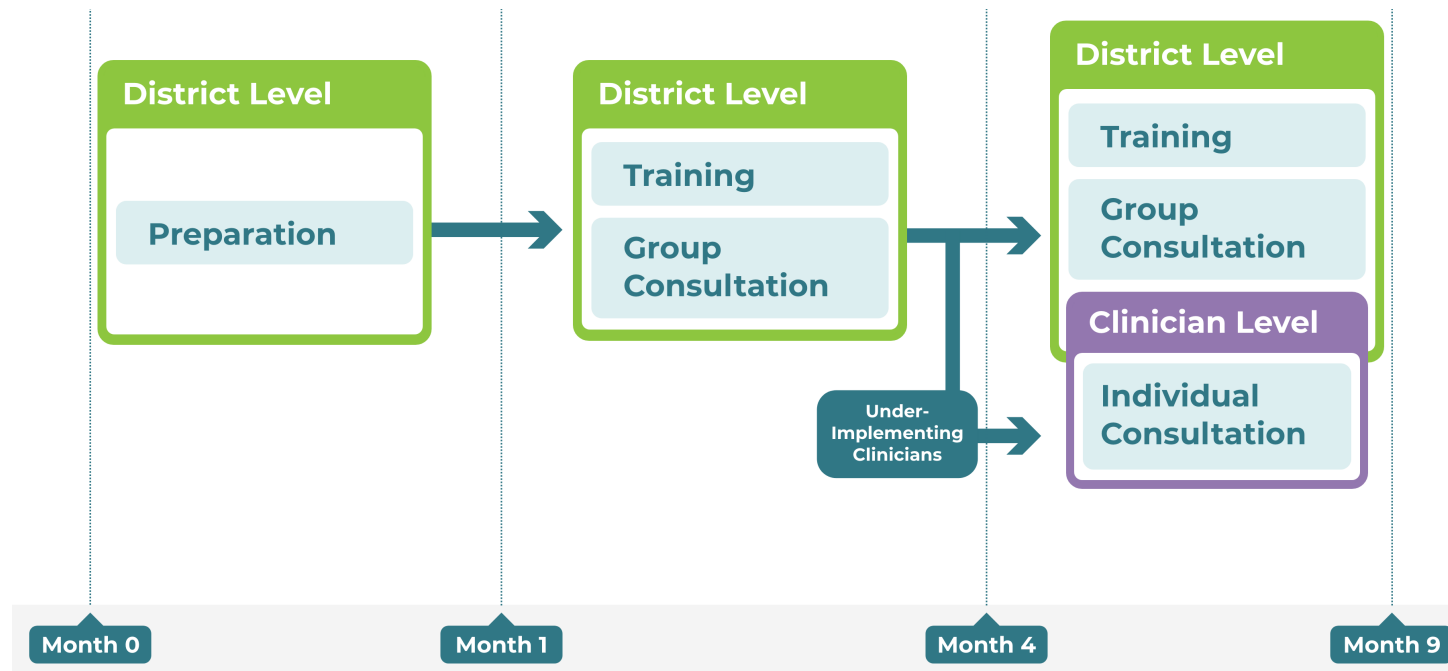


# MAISY Example 2

## Feedback & Outcomes for Clinically Useful Student Svcs (FOCUSS)

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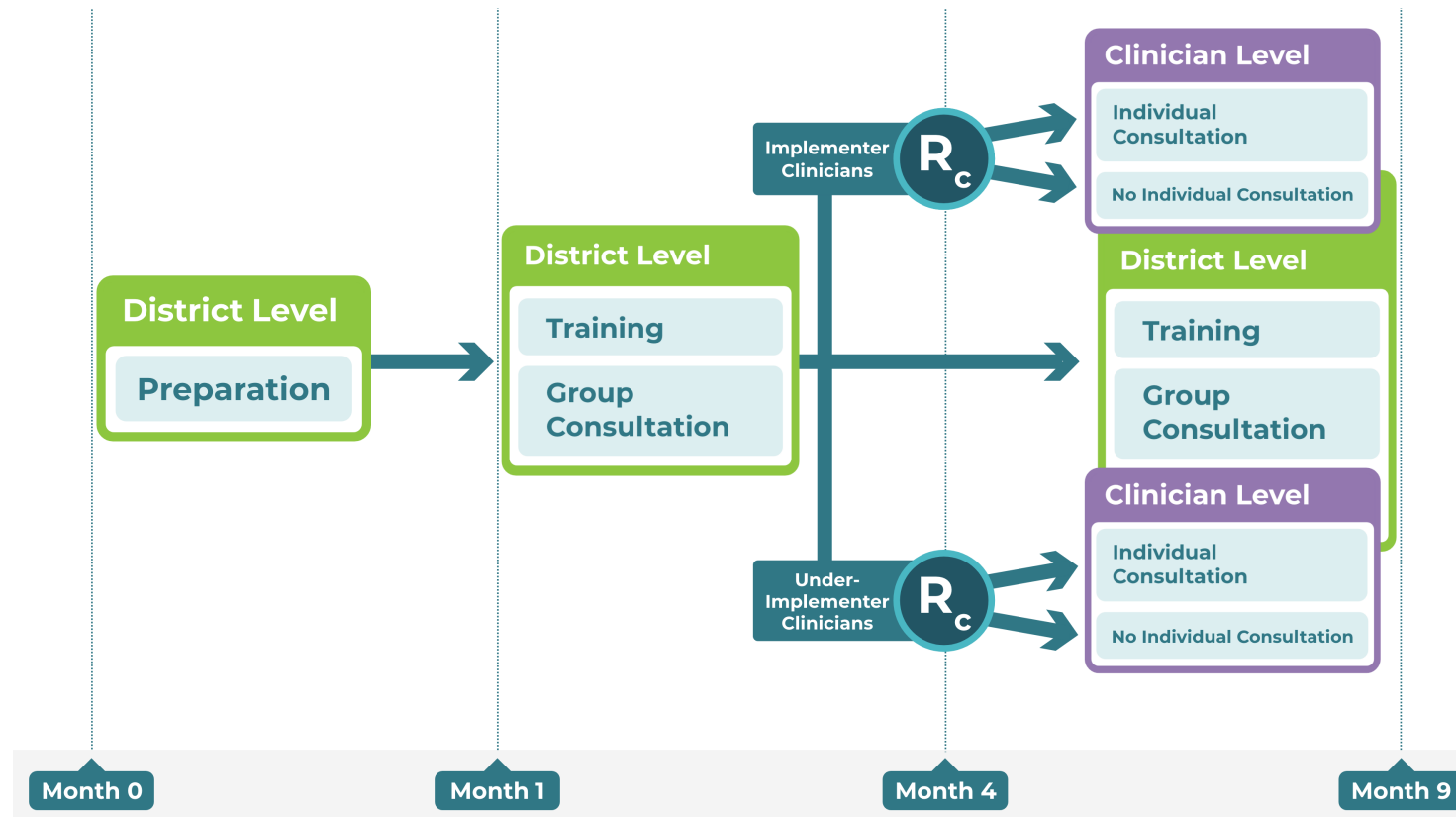
## Back to Optimization Questions #2 and #7 in FOCUS

	Type	In the context of FOCUS
2	Later phase strategies	What is the average effect of clinician-level Individual Consultation?
7	Ongoing tailoring variables	How do we define “under-implementing clinician”?

# Two-arm Optimization Randomized Trial

The FOCUSS Study

PI: Elizabeth Connors, Yale



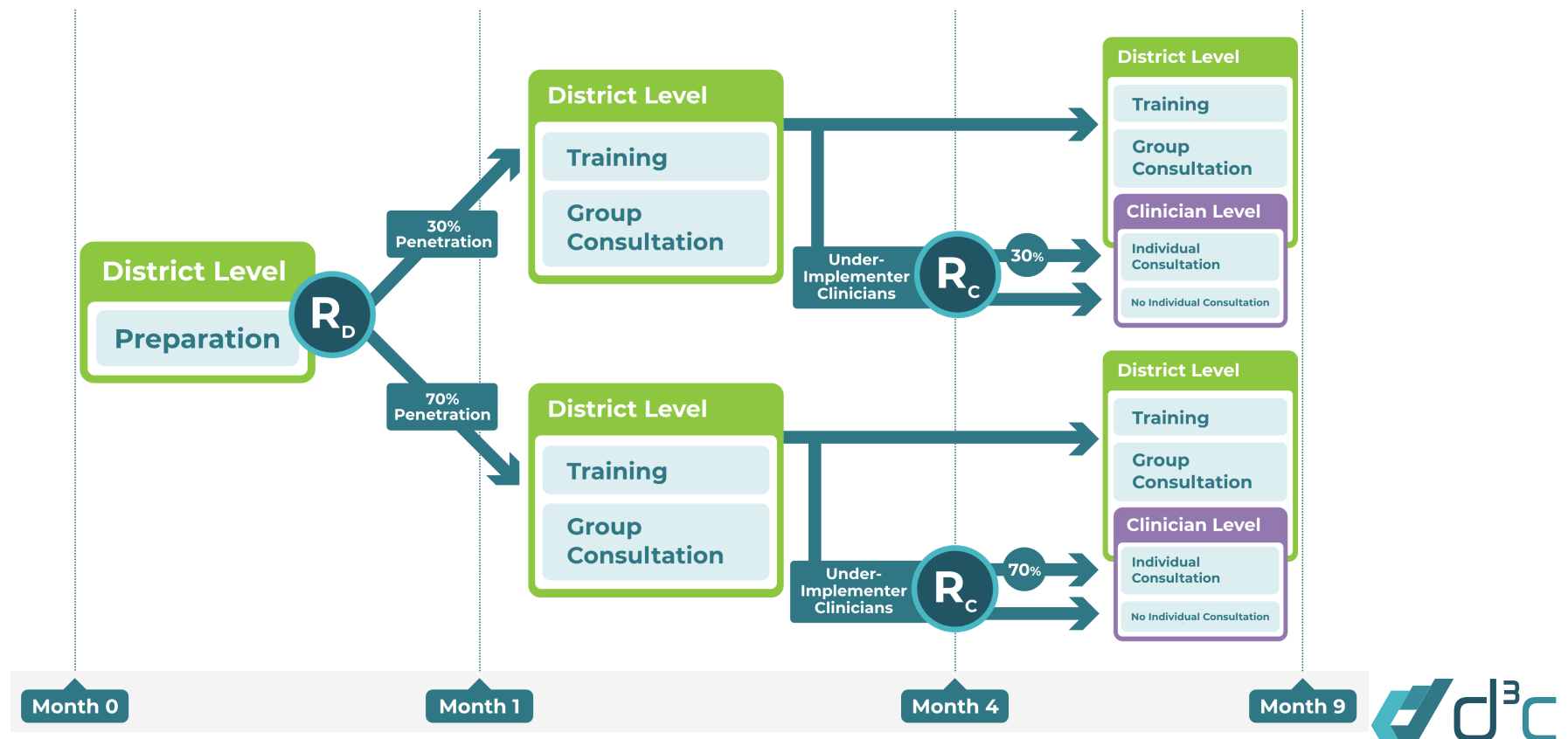
## Optimization Questions: Concerning Spillover

	<b>Spillover Questions!</b>	<b>In the context of FOCUSS</b>
<b>12</b>	Optimal tipping point effect?	Effect of providing Individual Consultation to 30% vs 70% of under-implementers in a district?
<b>13</b>	Outer level strategies that engender beneficial spillover?	Target a random 1/2 of under-implementing clinicians vs up to 1/2 on a first-come first-serve basis?

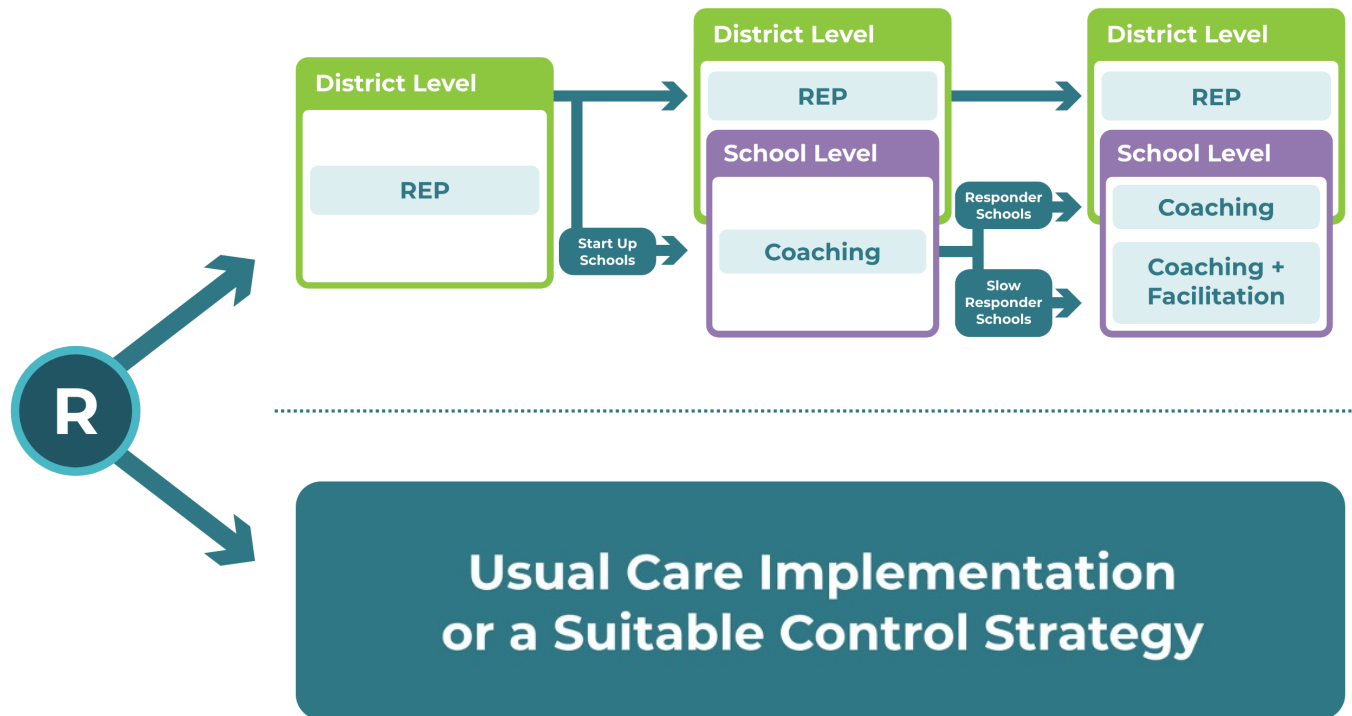
# A Hypothetical, Multilevel SMART

Illustrated using FOCUSS

PI: Elizabeth Connors, Yale



Evaluation and optimization are very different.  
This is evaluation.



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## **Institutes for Education Sciences**

R324B210001 (UMich, d3c);  
R305B210004 (UWash, A. Lyon)