

WORKSHOP

System Dynamics Modeling in Epidemiology

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Welcome and Workshop Overview

- 3 hours, 1 ice breaker, 4 hands-on exercises, some didactics, 1 short break, short post-workshop evaluation
- Handouts
 - Agenda
 - GitHub repository with instructions for downloading all conference materials
 - Stella Architect® software download instructions
 - *Words Commonly Used in SDM* (Quick Reference)
 - *Causal Loop Diagramming | Model Conceptualization Exercise*
 - Post-workshop evaluation form
- GitHub repository:



Learning Objectives

1. Understanding systems theory and the potential utility of SDM for research and action in epidemiology
2. Applying basic concepts of systems thinking: Drawing ‘balancing’ and ‘reinforcing’ loops
3. Creating, running and evaluating simple system dynamics models.

By the end of the workshop, participants will appreciate how system dynamics modeling is a novel, ‘mixed methods’ research methodology well-suited to understanding and addressing diverse, complex problems in epidemiology, public health, and medicine.



My Professional Research Foci: Three Aims

1. Participatory action research to develop and sustain high quality community-based systems of care
2. Capacity-building to reduce global health disparities, with a focus on reducing cancer burden
3. Applying systems thinking and system dynamics modeling to address complex health care problems

Modeling projects: HIV prevention and care, diabetes self-care, cancer screening policies, mental health services, school wellness programming, family planning and maternal health



What's your problem?

ICE BREAKER



Systems Theory

Hallmark

- Identifying feedback structures ('cybernetics')
- Understanding how things change ('causality')

General assumption

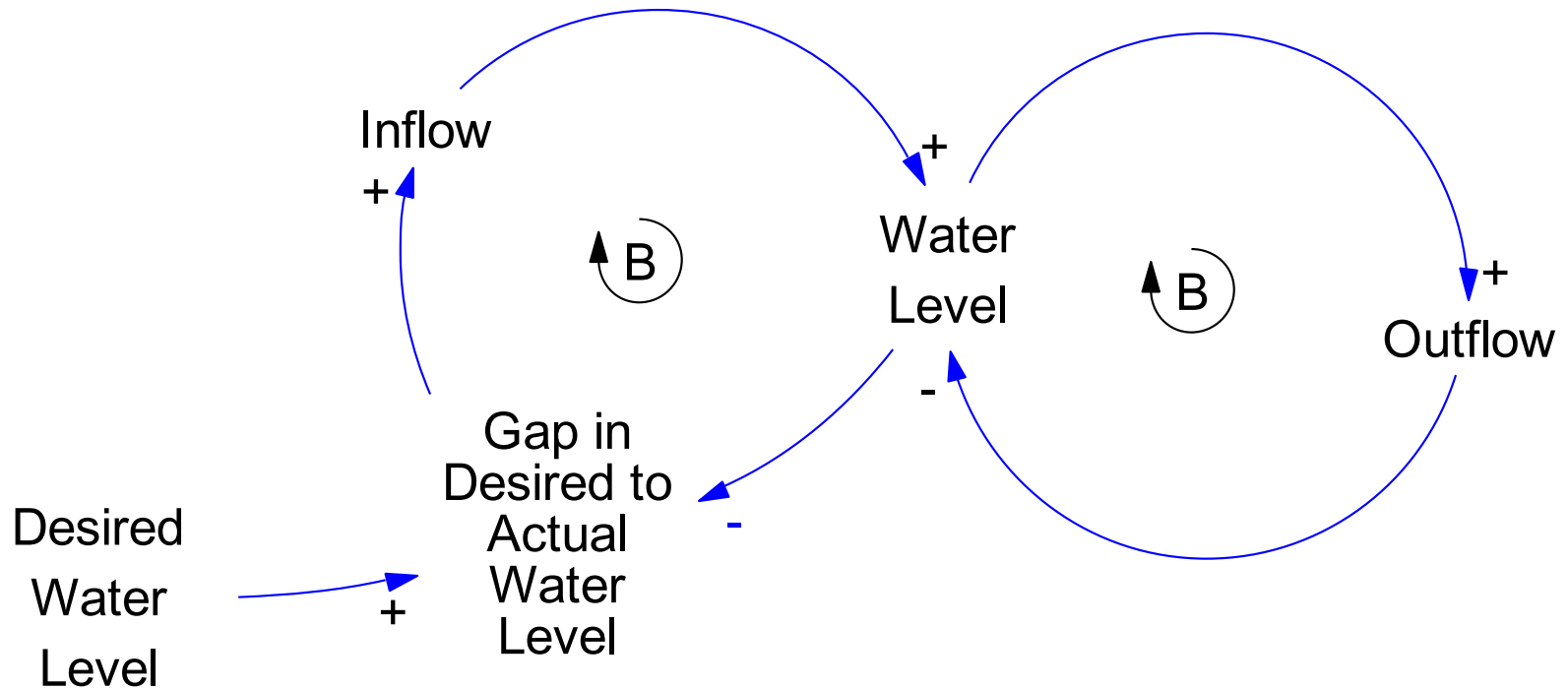
- Problems in natural and human systems have dynamic complexity
- Natural and human systems are 'goal-seeking' (i.e., gravitate towards a dynamic equilibrium; a state of homeostasis, sustainability, balance, stability)



Systems Theory: Feedback Concepts from Engineering

- Engineers characterize two kinds of feedback loops: ‘reinforcing’ and ‘balancing’
- Reinforcing feedback loop: Characteristically tends to amplify a change in any one of its elements over time.
 - Moving away from equilibrium
 - Instability
- Balancing feedback loop: Adjusts to counteract a change in any one of its elements over time.
 - Coming to equilibrium
 - A steady state: Homeostasis

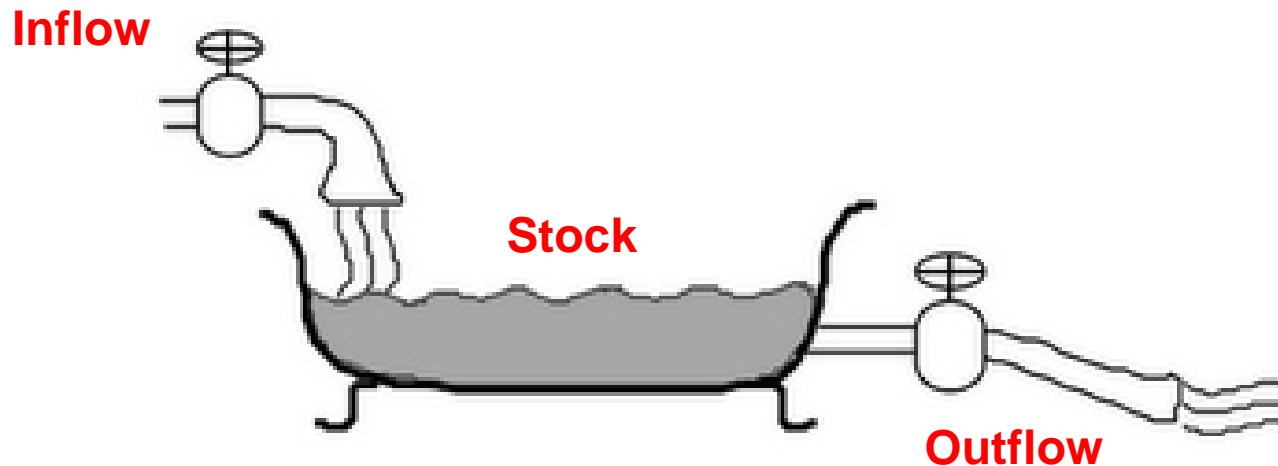
Basic Concepts is SDM: Causal Loop Diagram (CLD)



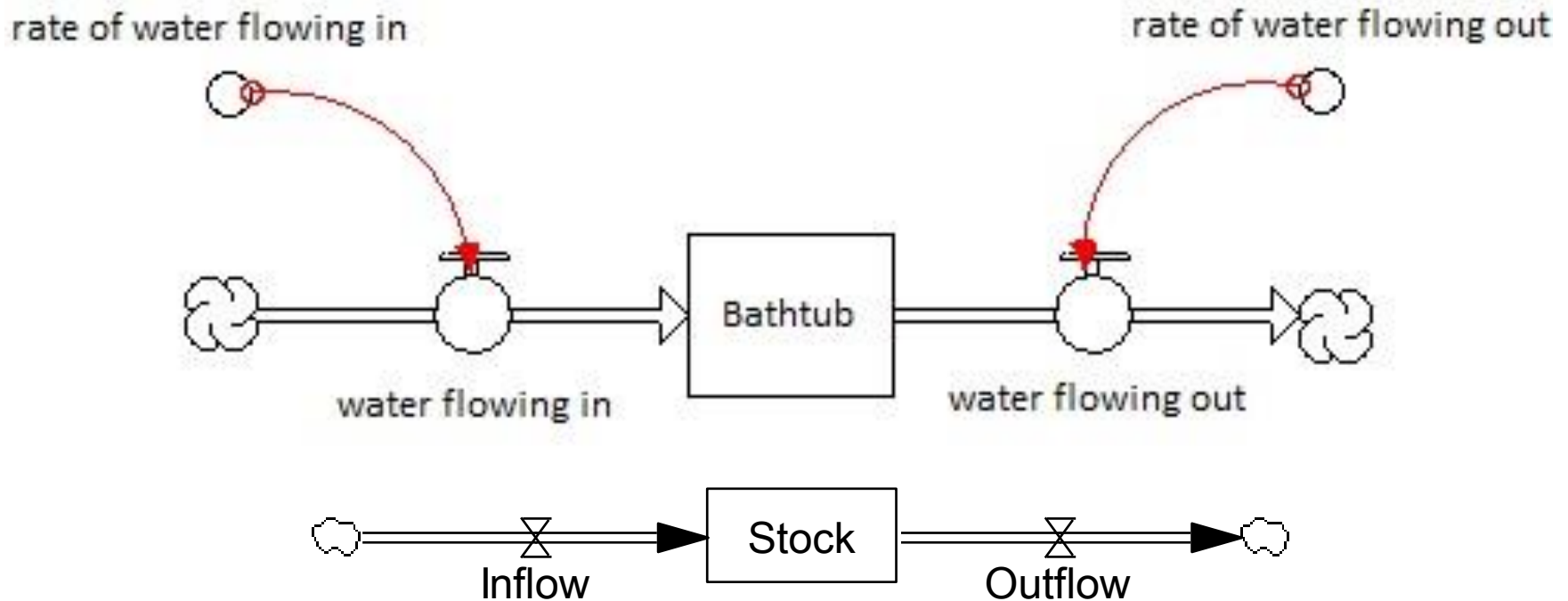
Basic Concepts in SDM: Stocks and Flows

Stock: Accumulation of units, e.g., liters of water in a tub

Flow: Movement of units into, out of, and between stocks



Another Way to Diagram Stocks and Flows



$$\frac{d(\text{Stock})}{dt} = \text{Inflow}(t) - \text{Outflow}(t)$$

[Link to demo](#)

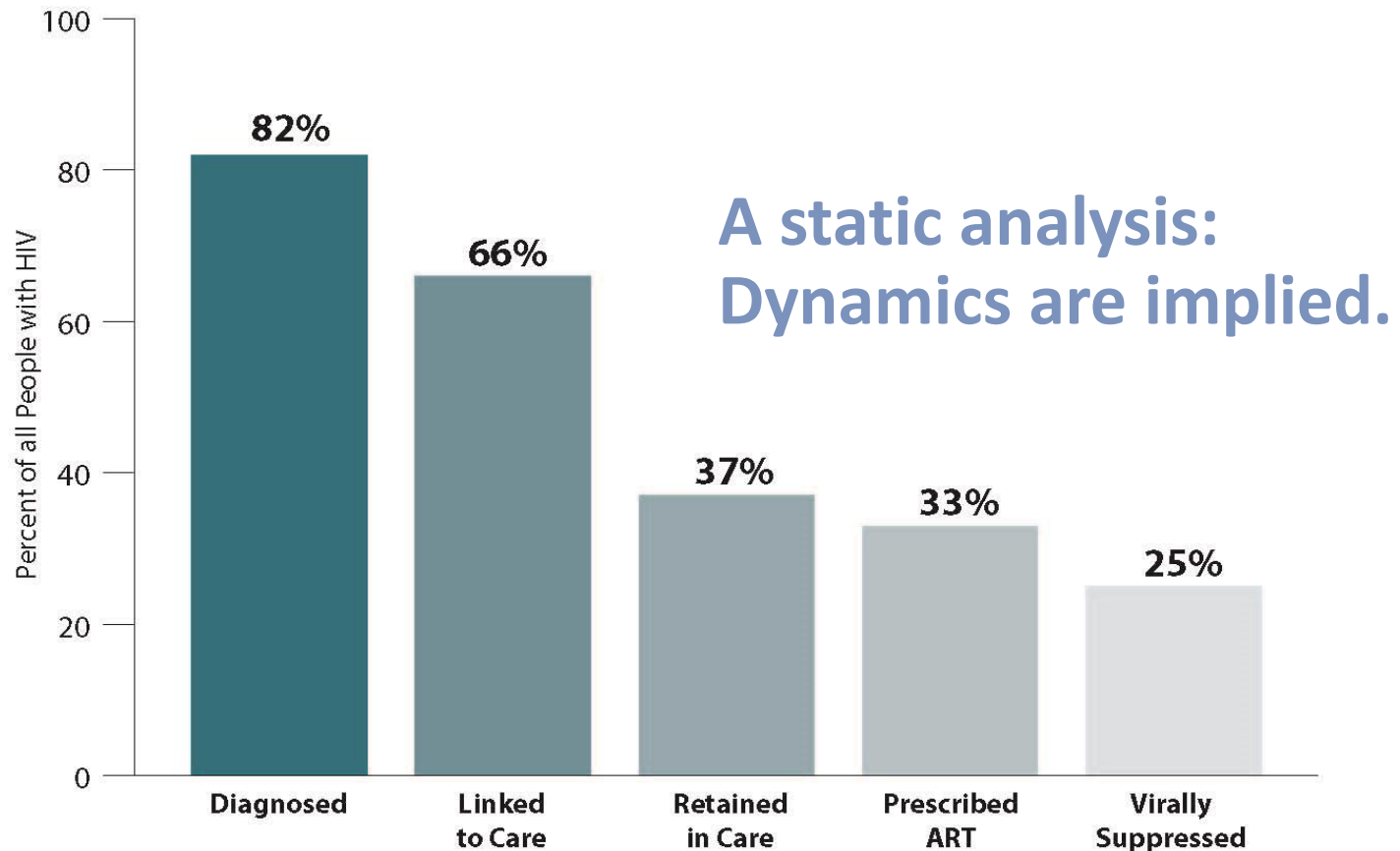
FLOW: Rate of change (derivative)

STOCK: Accumulation over time (integral)

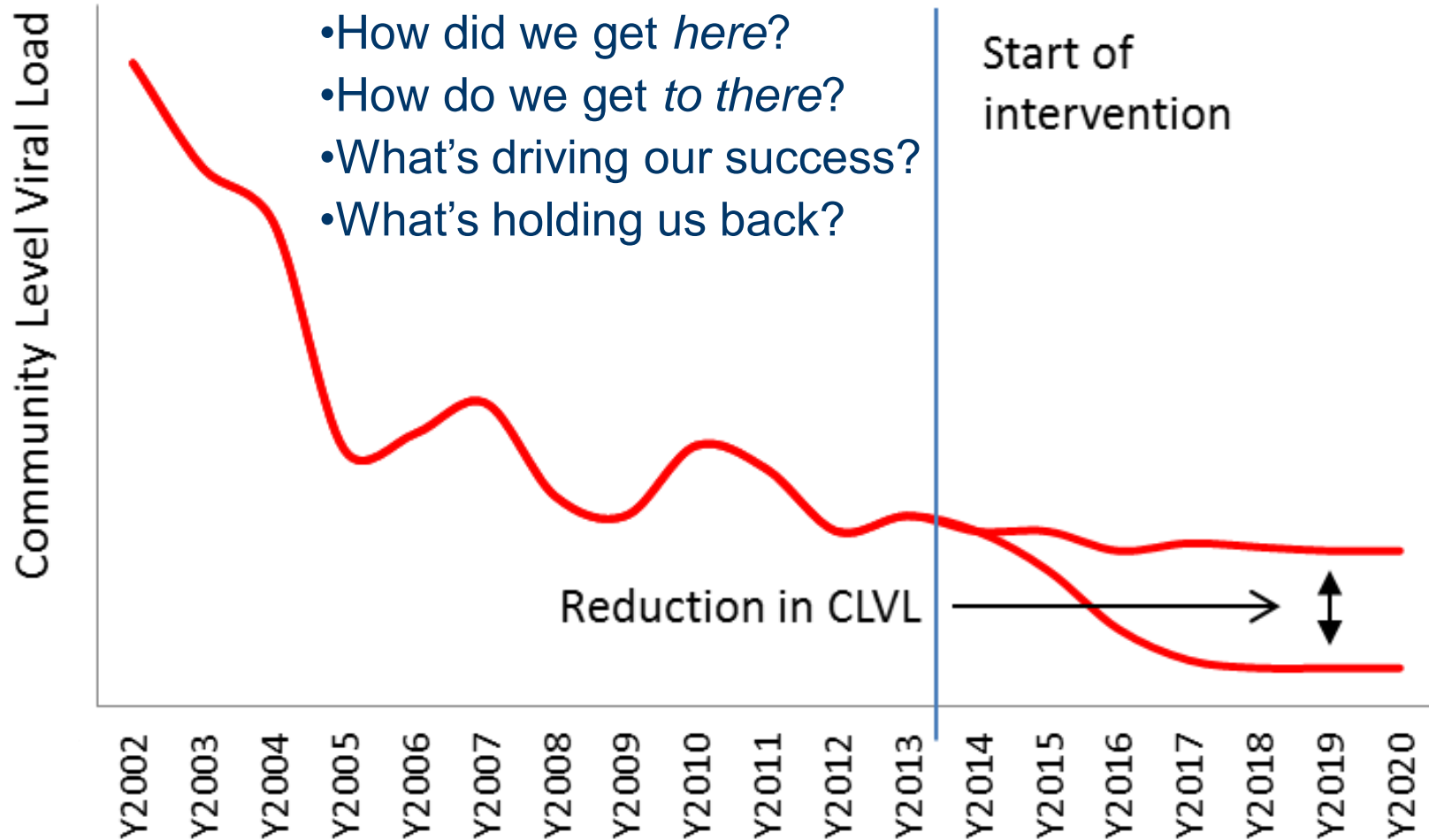
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CDC Stages of Care and the “Treatment Cascade” – US Population (July 2012)

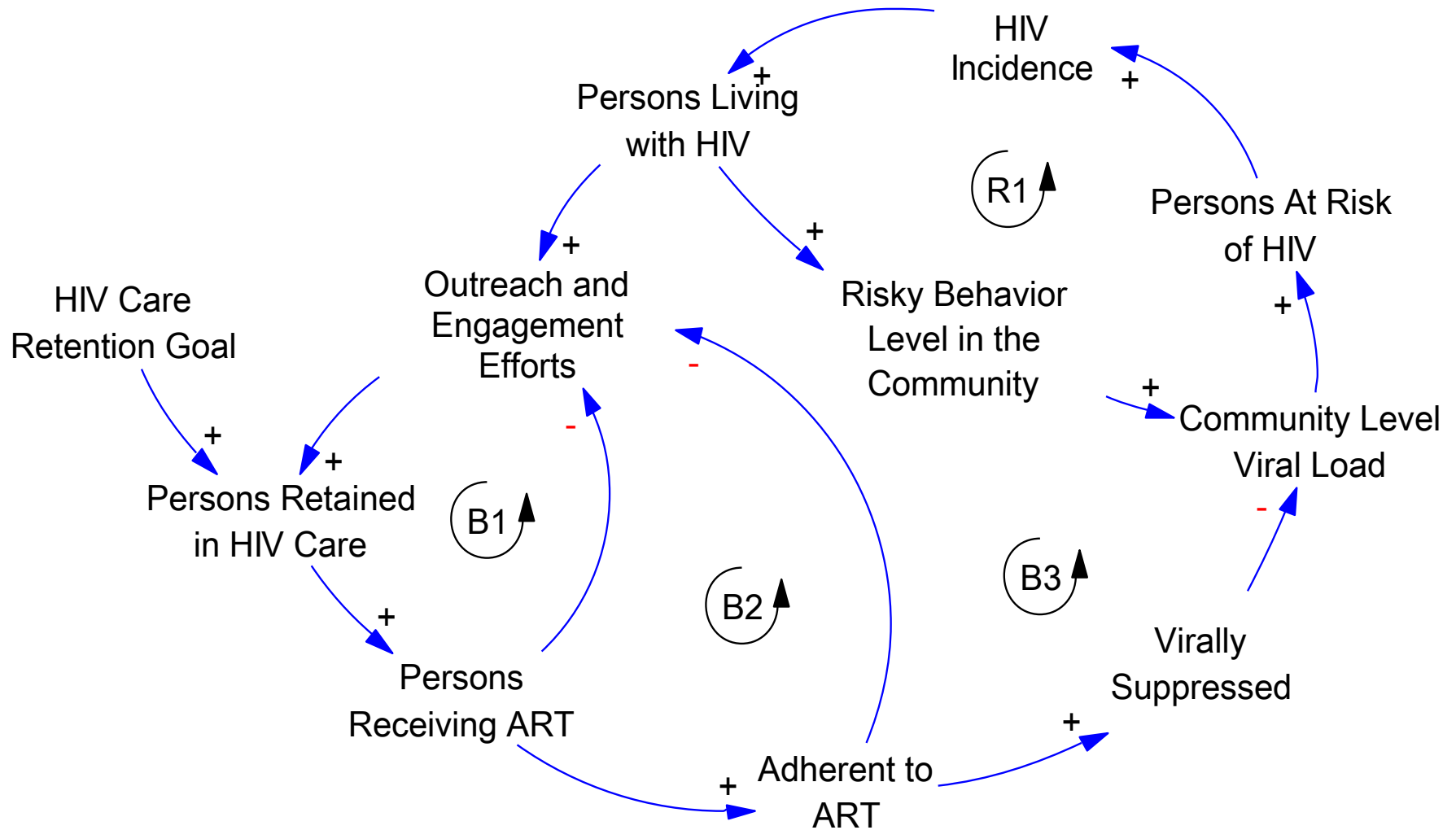
OVERALL: Of the 1.1 million Americans living with HIV, only 25 percent are virally suppressed.



A Hypothetical Reference Mode: Community Level Viral Load (CLVL) Over Time (2002 – 2020)



A Causal Loop Diagram Depicting the Dynamics of Community Viral Load





A systems thinking approach to conceptualizing
women's empowerment

EXERCISE



Need help with GitHub download or installing
Stella Architect?

BREAK & TECHNICAL ASSISTANCE CHECK



Systems Science Methodologies

- A broad class of intellectual methods for understanding part-and-whole interactions (Trochim, 2006)
 - Social Network Analysis
 - Agent-based Modeling
 - Micro-simulation Modeling
 - **System Dynamics Modeling**
- What is **System Dynamics Modeling (SDM)**?
 - A computer-aided approach to policy analysis and design characterized by information feedback (i.e., circular causality)
(Richardson, 1996)
 - [mathematically] A set of differential equations representing hypothesized time-dependent inter-dependencies among specified variables
 - [philosophically] A methodology intended to foster in-depth understanding about the ‘structure’ and ‘behavior’ of complex problems

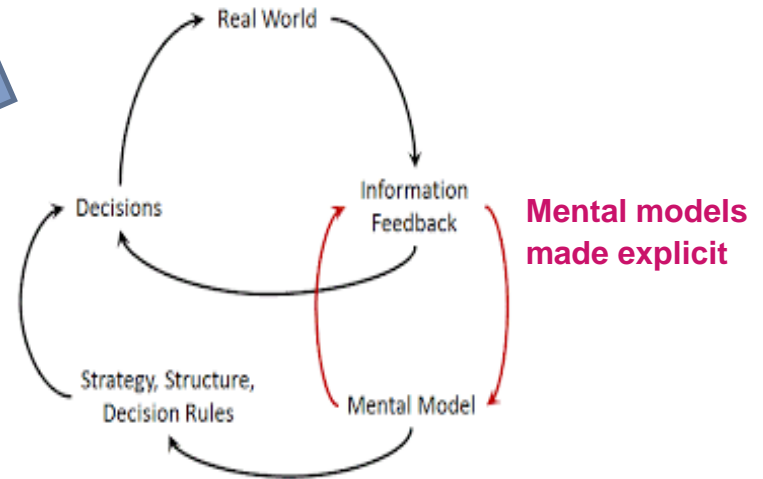


System Dynamics Modeling (SDM) is *Participatory Research*

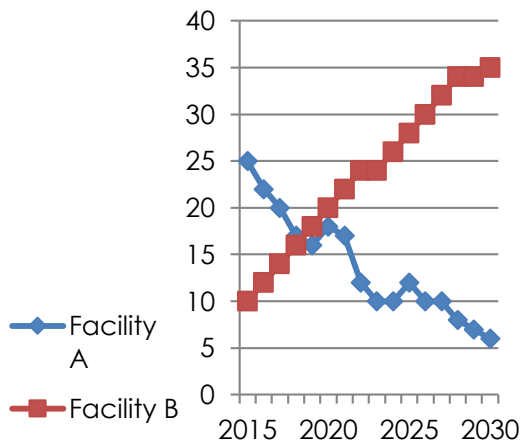
- In SDM, *group model building* engages *key stakeholders* in all phases of modeling
 - Patients and caregivers, frontline health workers, payors, and policy-makers, researchers, others
 - Introduction to group model building:
<https://en.wikibooks.org/wiki/Scriptapedia>
- SDM has the potential to:
 - Promote deep understanding about complex problems
 - Identify robust policies and implementation strategies
 - Inform novel research questions, hypotheses, and/or study designs
 - Build models that build community: Establish task forces, coalitions and other collaborative partnerships

System Dynamics Modeling is Participatory

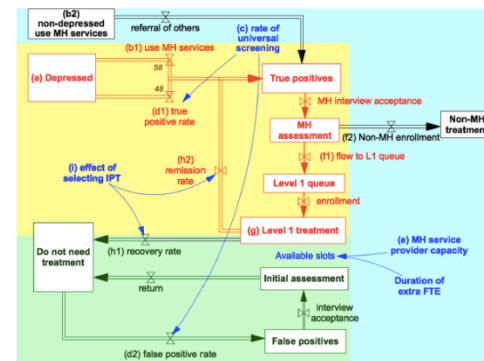
Stakeholder
engagement



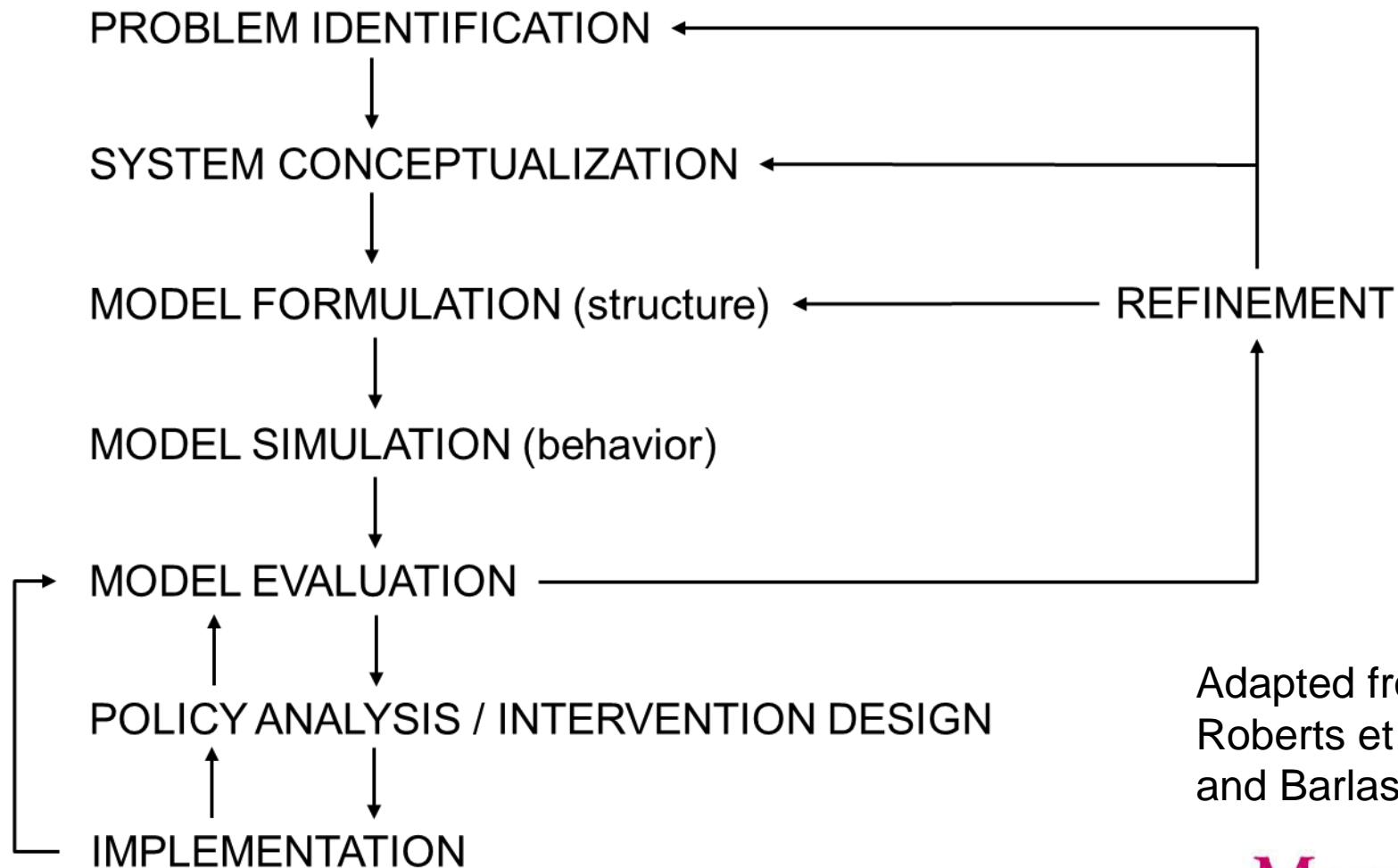
Model 'behavior' evaluated



System 'structure'
formulated and calibrated

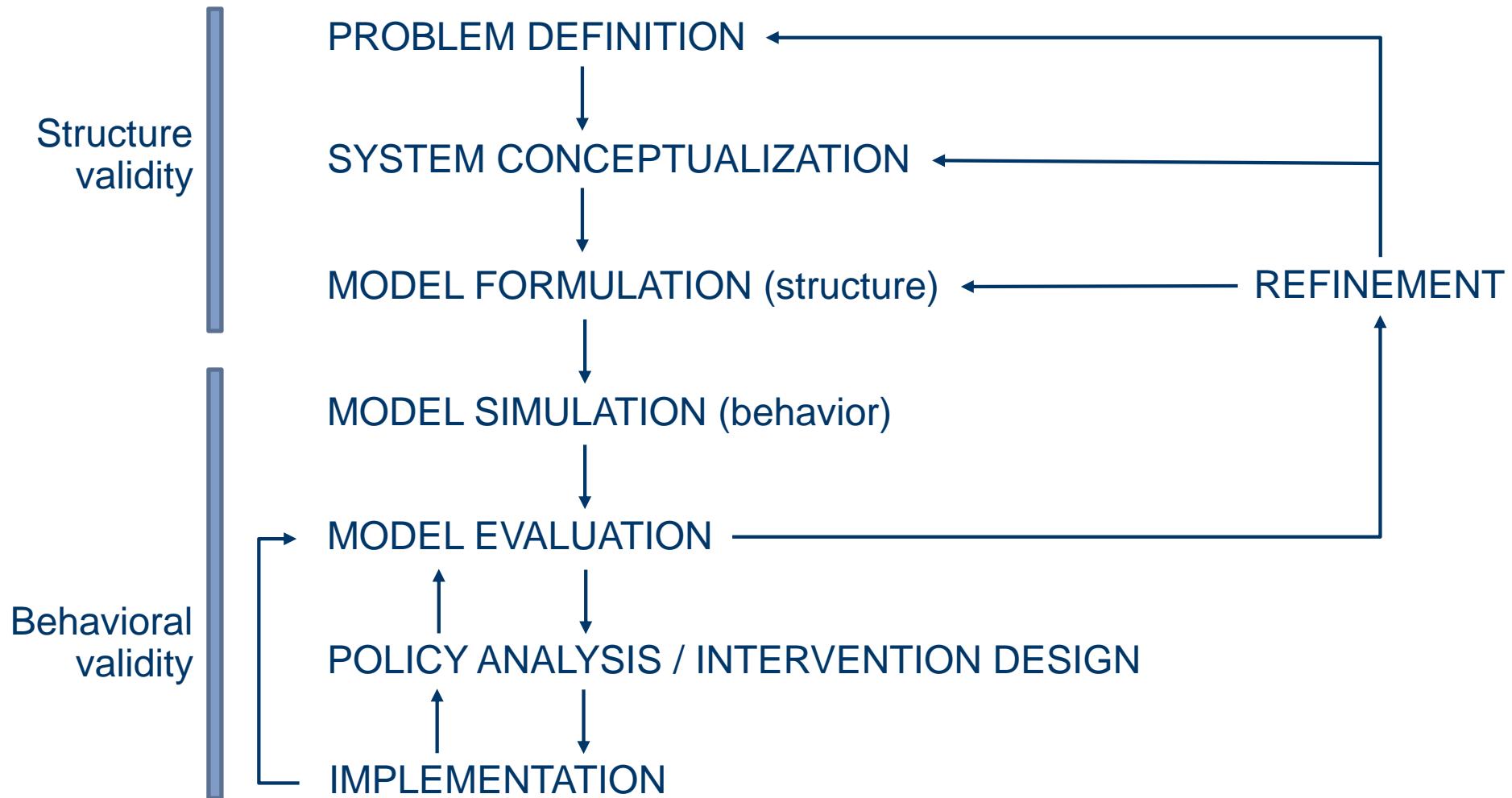


The Iterative Nature of SDM Development and Validation



Adapted from
Roberts et al. (1983)
and Barlas (1996)

The Iterative Nature of SD Model Development and Validation



Adapted from Roberts et al. (1983) and Barlas (1996)

A Topology of Validation Tests

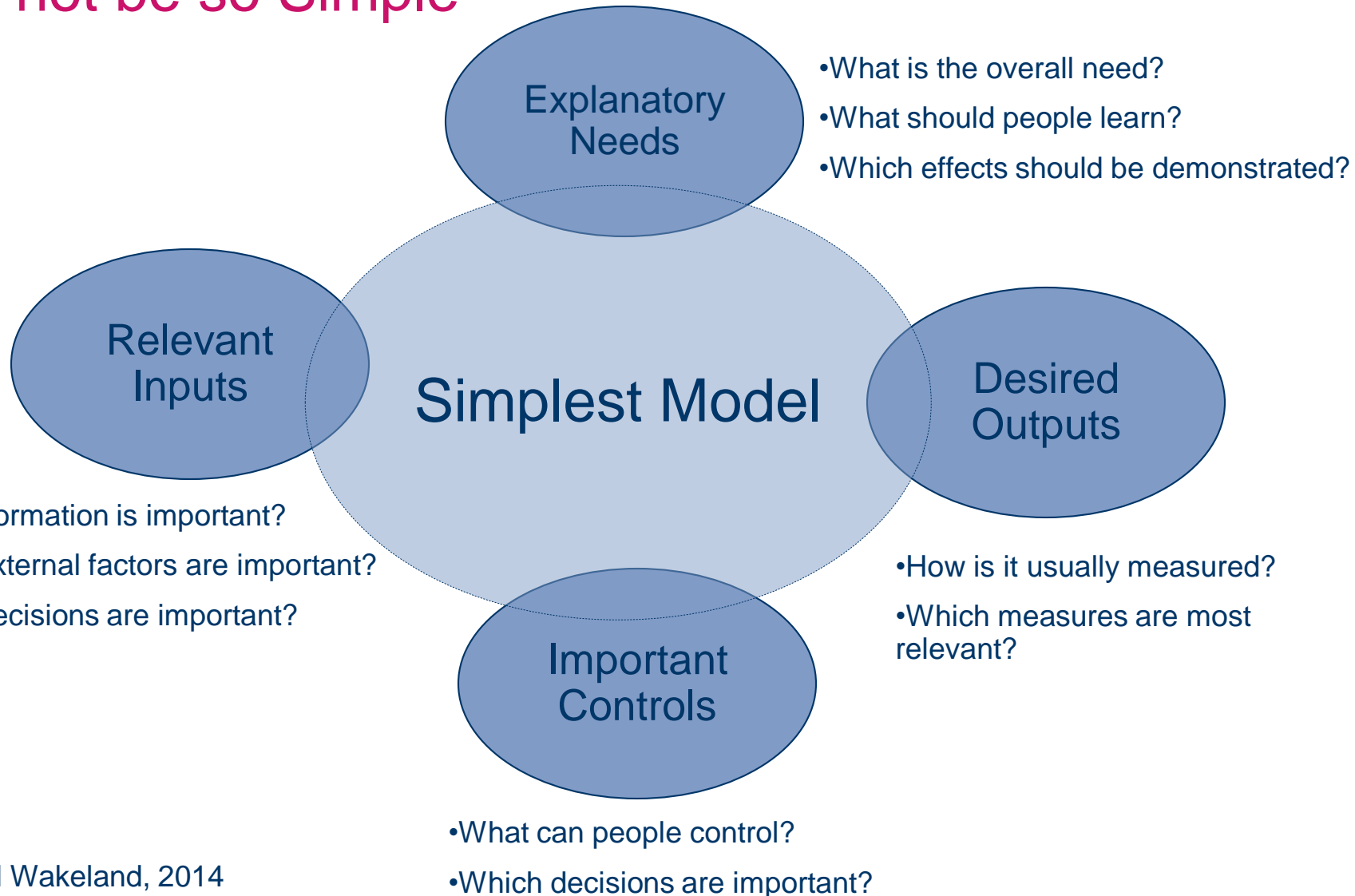
(Forrester & Senge, 1980; Barlas 1996; Martis, 2006)

Validity Type	Tests of suitability	Tests of consistency	Tests of utility
Structure	<ul style="list-style-type: none">• Structure-Verification• Dimensional-Consistency• Extreme-Conditions• Boundary-Adequacy	<ul style="list-style-type: none">• Face Validity• Parameter-Verification	<ul style="list-style-type: none">• Appropriateness for Target Audience
Behavior	<ul style="list-style-type: none">• Parameter Sensitivity• Structural Sensitivity• [Policy Sensitivity and Robustness]	<ul style="list-style-type: none">• Behavior-Reproduction• Behavior-Prediction• Behavior-Anomaly• Family Member• Surprising Behavior• Extreme-Policy• Boundary Adequacy• Behavior-Sensitivity• Statistical• [Changed Behavior Prediction]	<ul style="list-style-type: none">• Counter Intuitive Behavior• [Implementable Policy]

Modeling Challenge

The “Simplest Model” may not be so Simple

Models sometimes need to be complex, making them hard to understand and trust. To be useful, such models must be well calibrated and tested.





Balancing Breadth and Depth in System Dynamics Model Building and Calibration

1. Parsimony: Continually look to simplify the model as much as possible, given its intended purpose and targeted audience
2. Allow for redesign along the way, but know when to stop
3. Carefully document choices made throughout process to stay organized and minimize 'cycling'
4. Define *rubrics* | *tools* that facilitate model testing and foster stakeholder confidence



Types of Data for Model Formulation

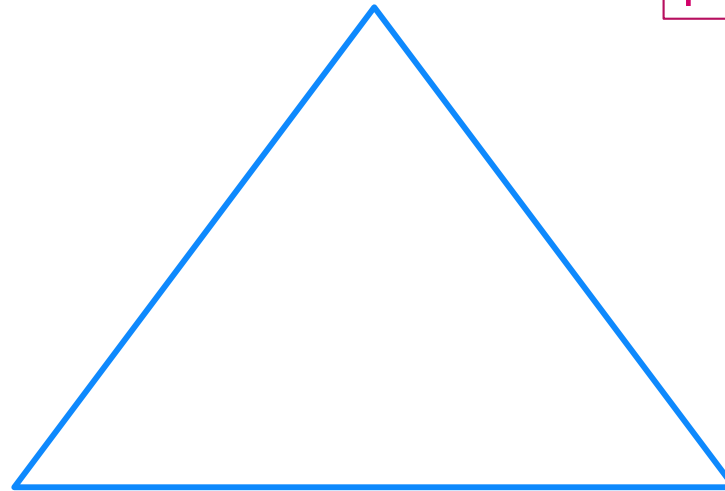
- Quantitative Data (numerical)
 - Cross sectional data: Charting the nonlinearities
 - Time Series Data: Defining the problem (i.e., defining key ‘reference modes’)
- Qualitative Data (written, mental)
 - Elicit the concept of the ‘reference mode’ from published findings or participant stakeholders (operators)
 - Methods
 - Observational approaches
 - Key informants interviews
 - Group interviews or Focus groups

Triangulating Sources of Information in System Dynamics Modeling Projects

Stakeholders' Expertise
and Lived Experience

Validity threat:
Selection and/or
participation bias

Validity threat:
Error of omission



Scientific Reviews and
Published Empirical
Evidence

Primary & Secondary
Data Analyses
(Qualitative and
Quantitative)

Validity threat:
Data quality/
fallibility




Understanding structure and behavior: First
order delays and smooths

EXERCISE



Simulating characteristics of populations: Co-flow structures

EXERCISE



Simulating patterns of disease in populations:
Aging chain structures (S-I-R)

EXERCISE



Tradeoffs in Modeling

1. Generality: Applicability of model to phenomena other than that for which it was developed
2. Realism: Degree to which the model reflects reality as viewed by experts in the field
3. Fit: Degree to which the model output matches historical data and has predictive accuracy
4. Precision: Fineness of model and level of details specified

Levins, R. (1966); Ip et al. (2013)



Three Levels of System Dynamics Modeling


Costanza and Ruth (1997)

1. Scoping and consensus-building models (limited information | high generality)
2. Research models (incorporates detailed historical or other empirical data sources)
3. Management or surveillance models (builds upon the first two stages; used to over an extended period to examine the implications of management actions)



Concluding Caveats

- EVERY model-building project is unique
 - The model and the modeling process both contribute value
 - Participatory process will help shape effective engagement and will inform modeling tradeoffs
- Need to consider
 - Time to build, test, and apply the model(s)
 - Modelers' skill/training and stakeholders' expert knowledge
 - Access to and quality of key data
 - Capacity to facilitate participatory modeling and to create practicable user interfaces for models
- Need to build effective rapport with target audience
 - Build trust; Achieve 'small wins' in model development
 - Appreciate differing philosophies of science



Thank you to all participants, to the IEA Conference Organizers, and to Dr. A. Olupelumi Adebisi!



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