

DISH-TREND: INTERVENTION MODELING SIMULATOR THAT ACCOUNTS FOR TREND INFLUENCES

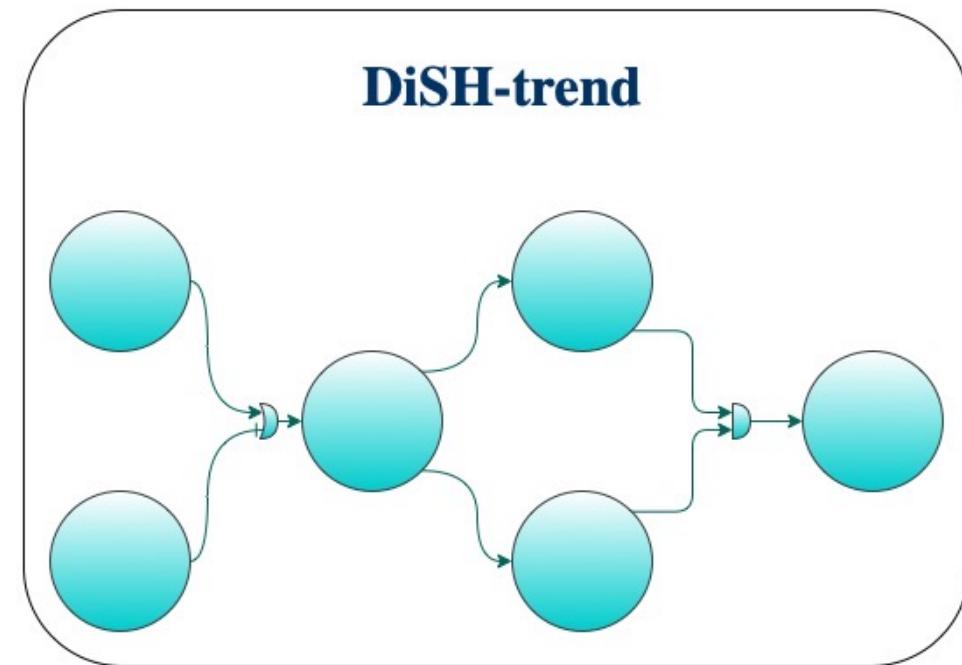
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STEFAN ANDJELKOVIC & NATASA MISKOV-ZIVANOV



INTRODUCTION

- Dynamic network models – crucial in understanding complex system dynamics
- Nodes = components, edges = interactions
- Applications: biology, psychology, sociology, economics, politics, and engineering
- Interventions = constraints on values, or structural modifications



SIMILAR MODELING APPROACHES

- Boolean networks (Albert et al 2008)
- Discrete networks (Sayed et al 2017)
- Bayesian networks (Needham 2007, Wilkinson et al 2007)
- ODE-based models (Materi and Wishart 2007)
- Rule-based models (Faeder et al 2005)
- Petri nets (Chaouiya et al 2007)

DISH (DISCRETE STOCHASTIC HETEROGENEOUS) SIMULATOR

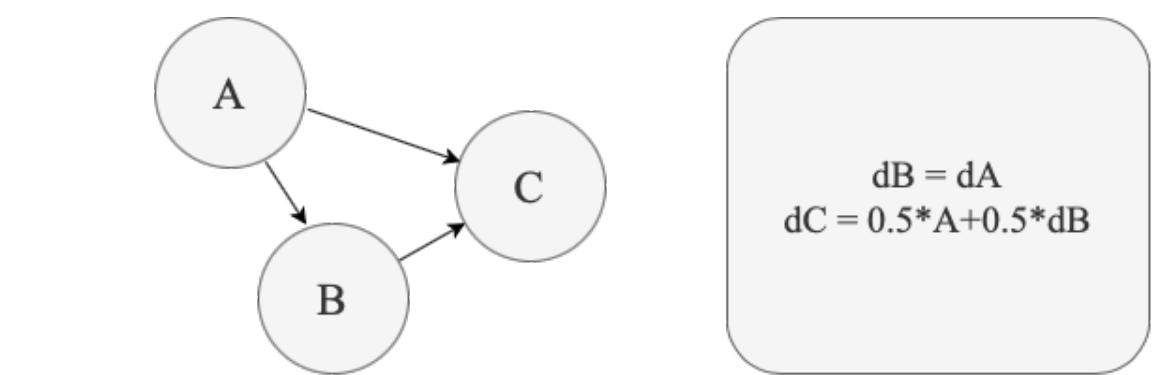
- Discrete levels (extension from Boolean networks)
- Update functions (Boolean)
- BioRECIPES model representation: element-focused
- Simulation schemes: simultaneous vs sequential
- Value toggling (enforced value constraints)

FROM DISH TO DISH-TREND

- Algebraic update function: weighted sums of regulation functions
 - $B(\mathbf{A}, \mathbf{I}) = \sum_{i=1}^k \left(w_i^{(p)} \prod_{q \in \mathbb{I}_{p,i}} A_{i,q} \right) - \sum_{j=1}^l \left(w_j^{(n)} \prod_{r \in \mathbb{I}_{n,j}} I_{j,r} \right)$ (WEIGHTED SUM-OF-PRODUCTS)
 - $F(\mathbf{A}, \mathbf{I}) = \frac{\lceil L \cdot |B(\mathbf{A}, \mathbf{I})| \rceil}{L} \cdot \text{sign}(B(\mathbf{A}, \mathbf{I}))$
- Adding trend-based regulation
 - $B_{trend}(\mathbf{A}, \mathbf{I}) = \sum_{i=1}^k \left(w_i^{(p)} \prod_{q \in \mathbb{I}_{p,i}} \Delta A_{i,q} \right) - \sum_{j=1}^l \left(w_j^{(n)} \prod_{r \in \mathbb{I}_{n,j}} \Delta I_{j,r} \right)$
- Hybrid regulation – the sum of level-based and trend-based regulation
- Beyond linear models – any polynomial of $A_{i,q}$, $\Delta A_{i,q}$, $I_{j,r}$, and $\Delta I_{j,r}$

SMALL EXAMPLE

- 3-element network
- A – controlled input
- $t=1$: B remains unchanged
- $t=4$: B increased by $A(t=3) - A(t=0)$



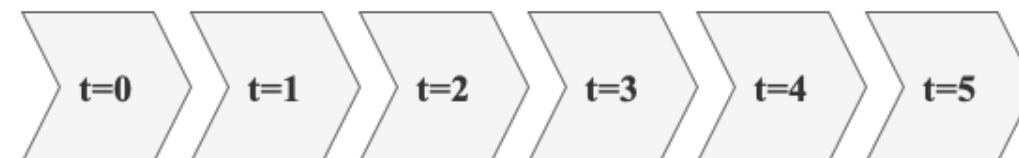
Toggle sequence of A:
 $v=0.2, (v=0.6, t=2), (v=0.4, t=3)$

Update sequence:
B, C, C, B, C

A:	0.2	0.2	0.6	0.4	0.4	0.4
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B:	0.1	0.1	0.1	0.1	0.3	0.3
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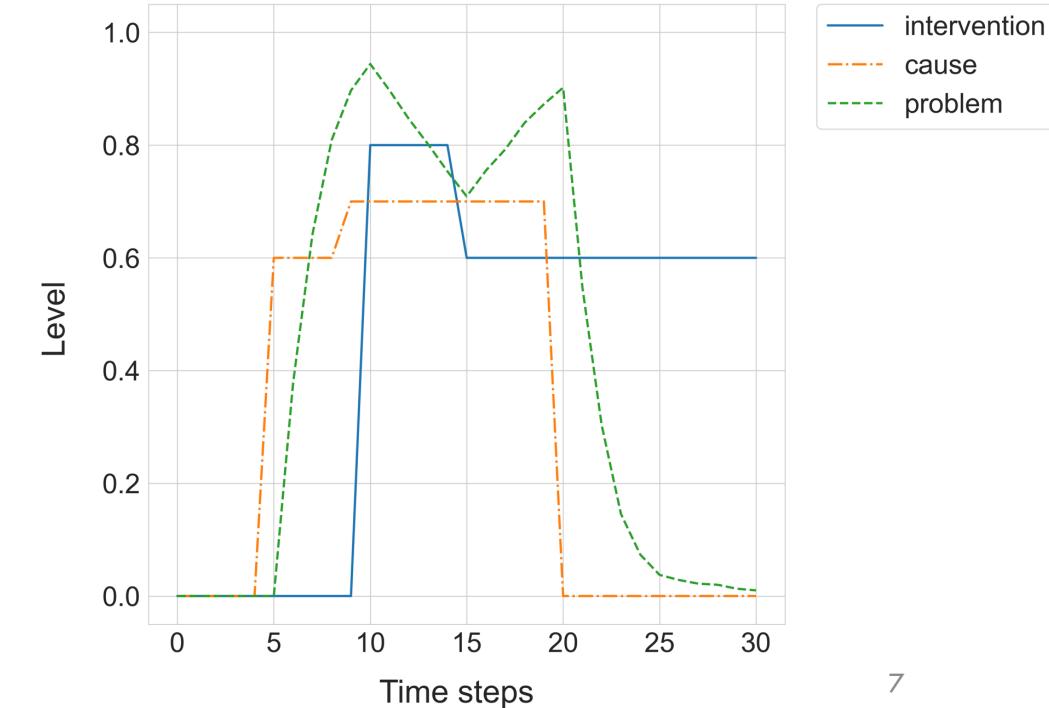
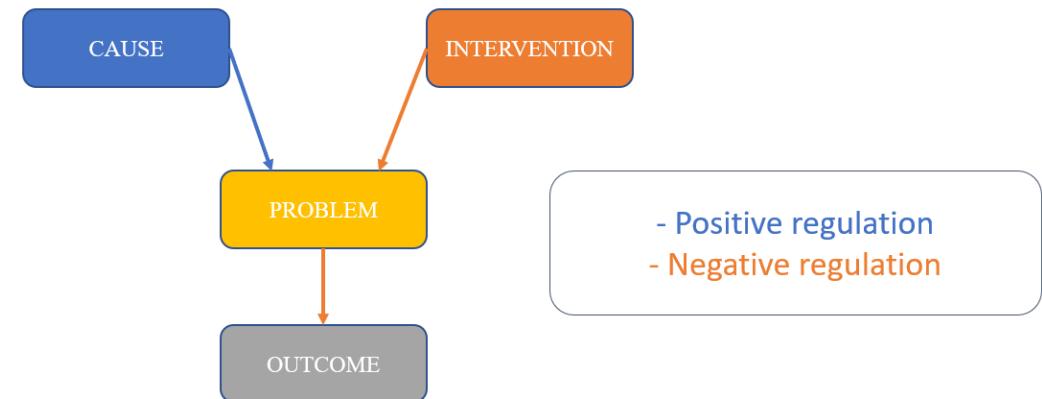
C:	0	0	0.1	0.4	0.4	0.7
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INTERVENTION TOY MODEL

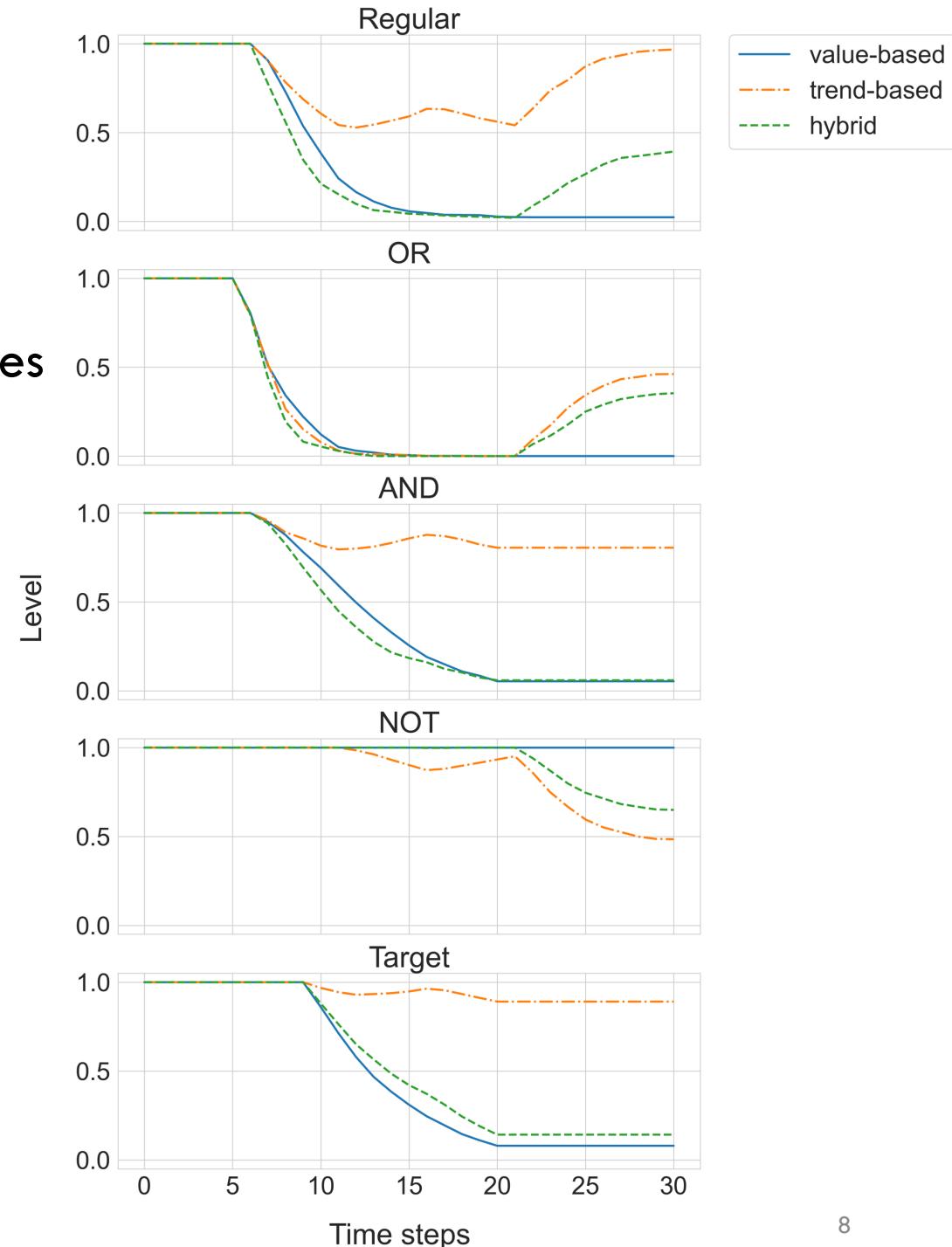
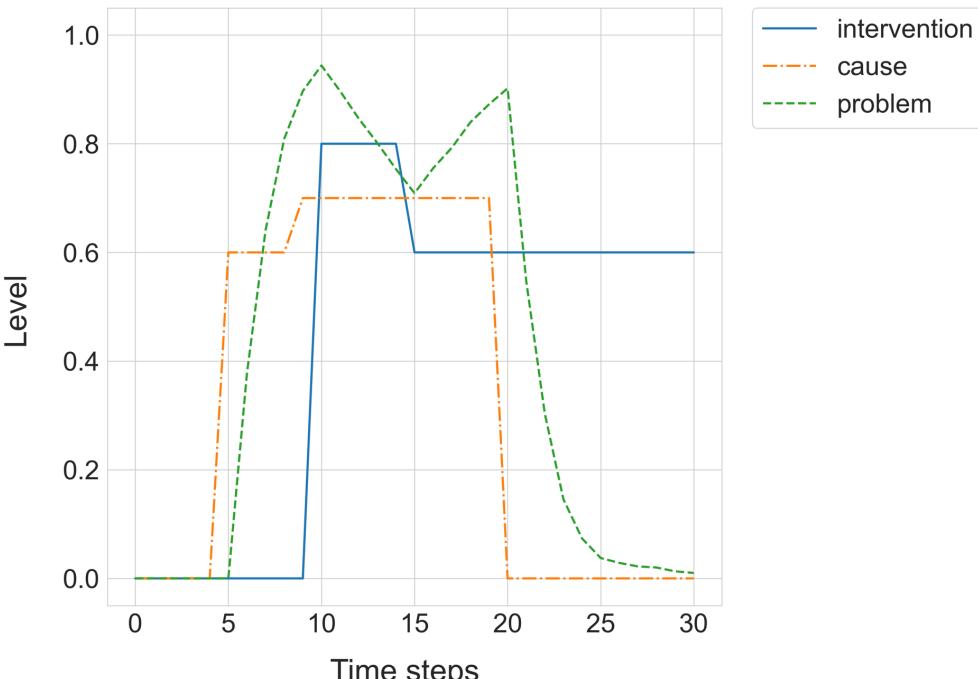
- 4-element network
- Linear model
- 5 tested scenarios:

Scenario	Negative regulation
Regular	$d(\text{OUTCOME}) = -\text{PROBLEM}$
OR	$d(\text{OUTCOME}) = -(\text{PROBLEM} + \text{CAUSE})$
AND	$d(\text{OUTCOME}) = -(\text{PROBLEM} * \text{CAUSE})$
NOT	$d(\text{OUTCOME}) = \text{PROBLEM}$
Target	$d(\text{OUTCOME}) = -(\text{PROBLEM} * \text{CAUSE})$ only when $\text{CAUSE}=0.7$



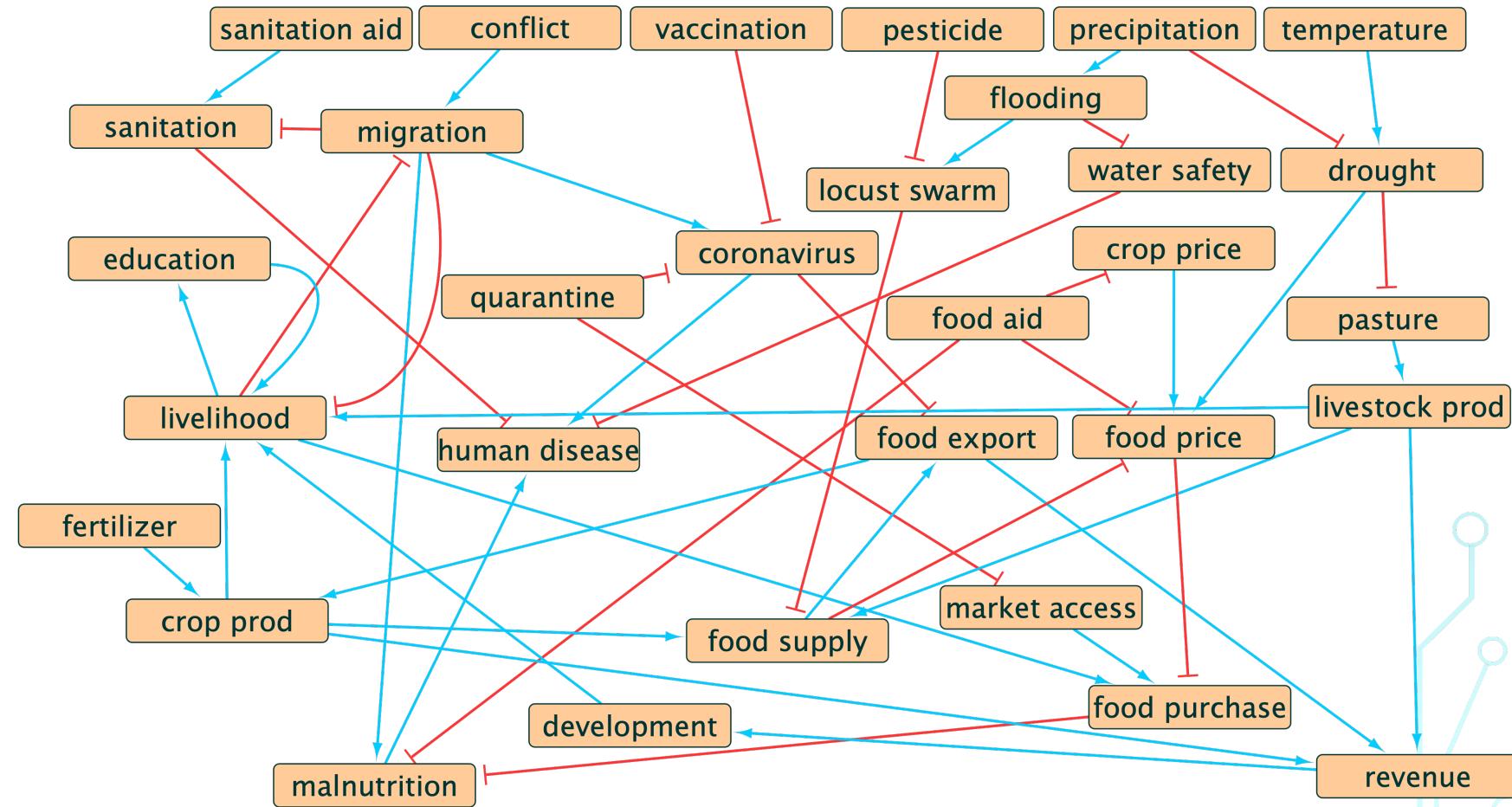
RESULTS

- Toy model, 5 scenarios, 3 regulation types
- Execution time: 0.37-0.54 s



LARGER MODEL – FOOD INSECURITY IN ETHIOPIA

- 31 nodes, 49 edges
- 10 inputs:
 - Temperature
 - Precipitation
 - Pesticide
 - Medical treatment
 - Fertilizer aid
 - Sanitation aid
 - Food Aid
 - Conflict
 - Quarantine
 - Vaccination

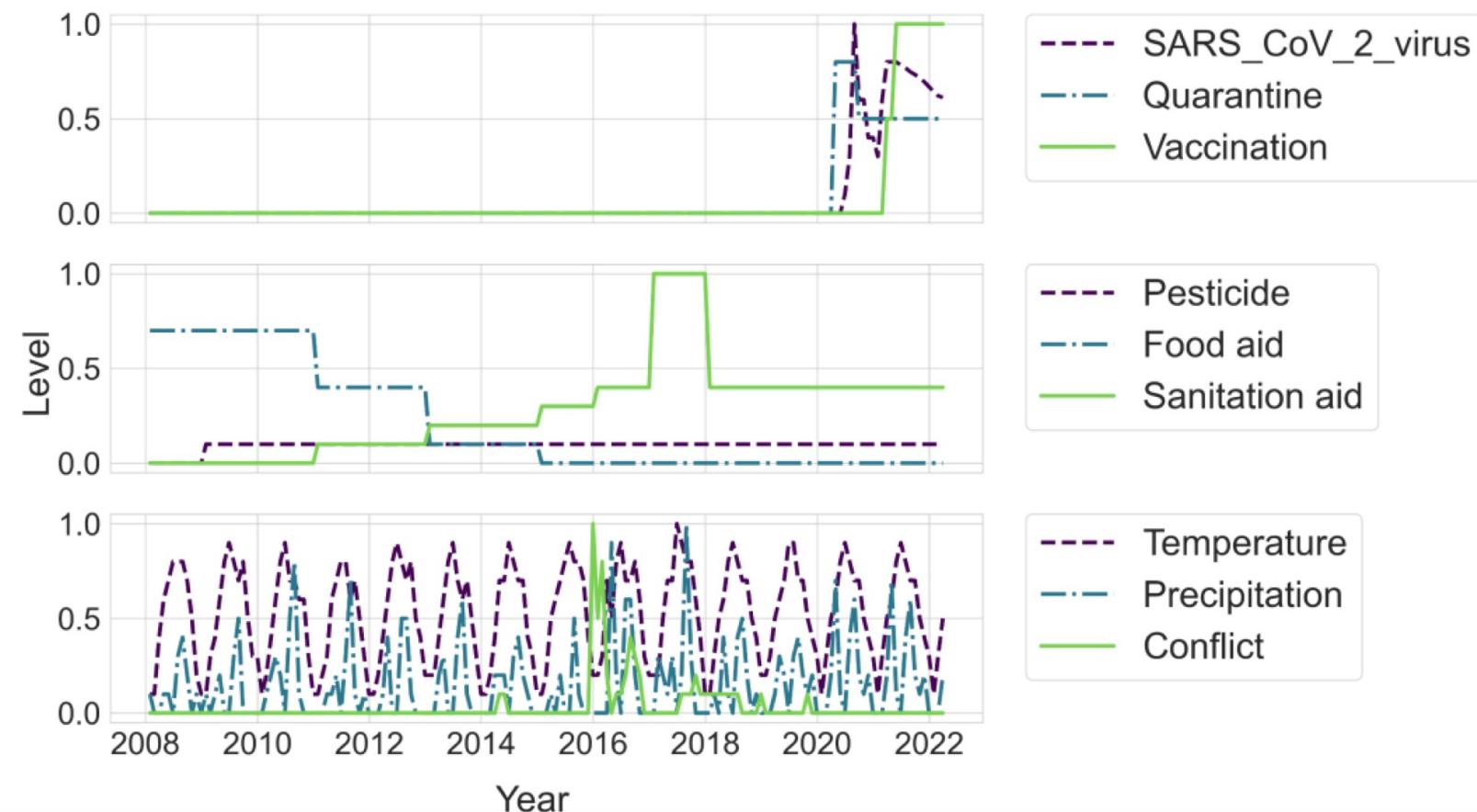


LARGE MODEL – DATA SOURCES

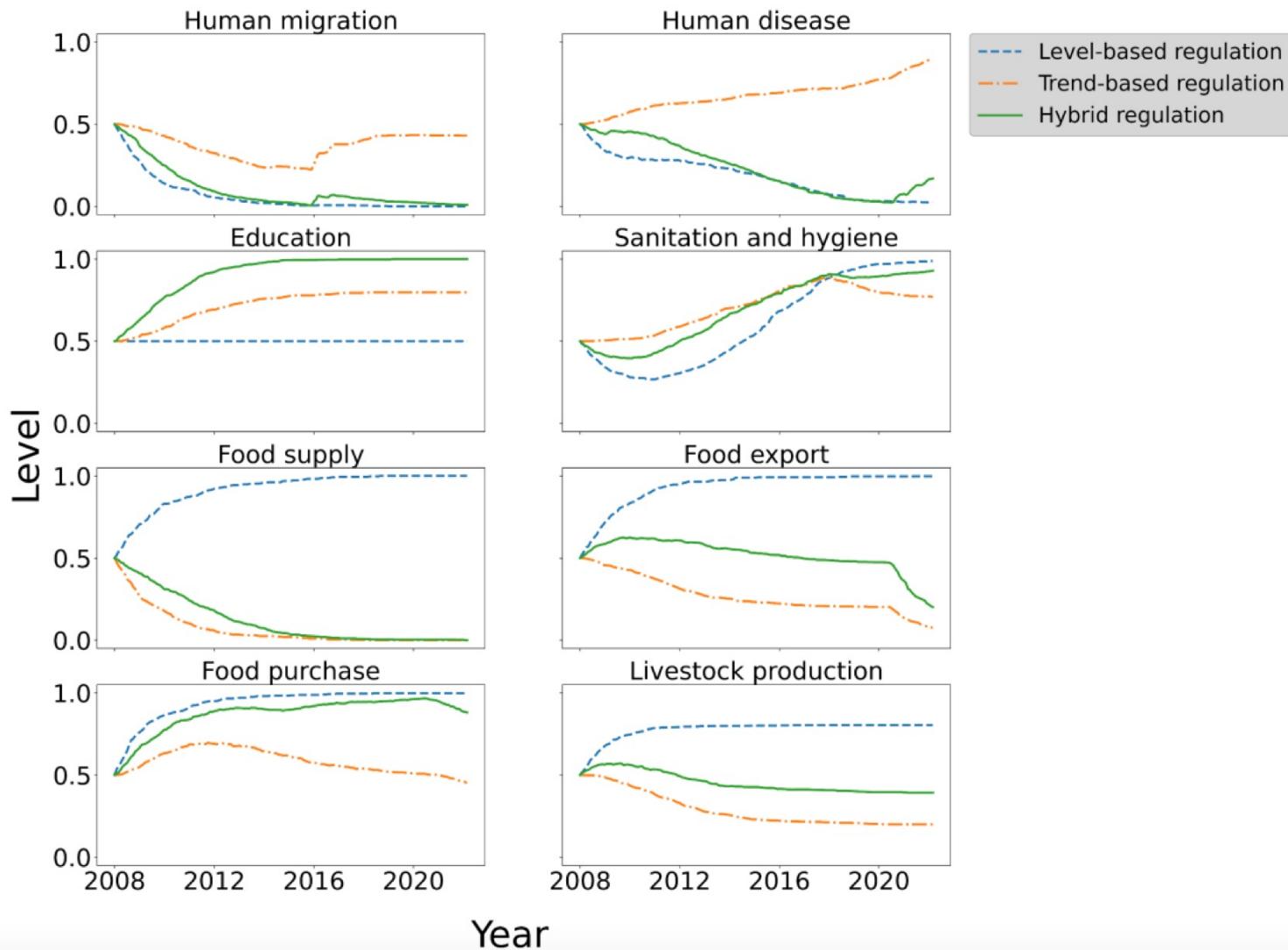
Variable	Source	Reference	Resolution (spatial and temporal)
Conflict	ACLED	Raleigh et al 2010	Ethiopia/daily
Food aid	FAOSTAT	FAOSTAT	Ethiopia/yearly
Pesticide	FAOSTAT	FAOSTAT	Ethiopia/yearly
Precipitation	CHIRPS	Funk et al 2005	Oromia/daily
Quarantine	EPHI	EPHI	Ethiopia/monthly
Sanitation and hygiene aid	USAID	USAID Database	Ethiopia/yearly
SARS-CoV-2 virus	OWID	OWID; Roser et al 2020	Ethiopia/daily
Temperature	ERA5	ERA5; Hersbach et al 2020	Oromia/daily
Vaccination	WHO	WHO	Ethiopia/daily

RESULTS – LARGE MODEL

- Execution time:
 - Level-based: 68.7 s
 - Trend-based: 71.2 s
 - Hybrid: 74.7 s



RESULTS – LARGE MODEL



CONCLUSIONS

- Trend-based regulation resolves some of the level-based counter-intuitive behaviors
- Hybrid regulation gives more modeling flexibility
- DiSH-trend simulator can be used to model multi-domain complex systems
- Major advantages:
 - Highly tunable
 - Robust to missing data
 - Easy to automate and integrate with other software

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MeLoDy lab

Mechanisms and Logic of
Dynamics



Natasa

Miskov-Zivanov, PhD



Kara Bocan,
PhD



Cheryl
Telmer, PhD



Yasmine
Ahmed



Casey
Hansen



Niteesh
Sundaram



Evan
Becker



Khaled
Sayed, PhD



Adam
Butchy



Emilee
Holtzapple



Gaoxiang
Zhou



Handa
Ding

Collaborators

- George Wittenberg, MD, PhD, Neurology, UPitt
- Peter Spirtes, PhD, Philosophy, CMU
- Brent Cochran, PhD, School of Medicine, Tufts
- Sandra Cascio, PhD, Immunology, UPitt
- Michael Lotze, MD, UPCI, UPitt
- Christof Kaltenmeier, MD, UPCI, UPitt
- Ed Hovy, PhD, LTI, CMU
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<https://www.nmzlab.pitt.edu>



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