### Harvard and Tanzania Celebrate 25 Years of Collaboration

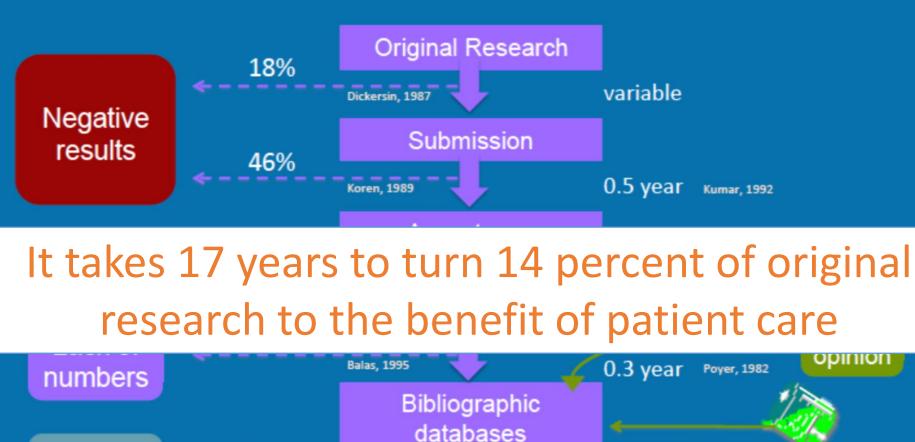
### Implementation Science and the Practice of Public Health: The Case of PEPFAR

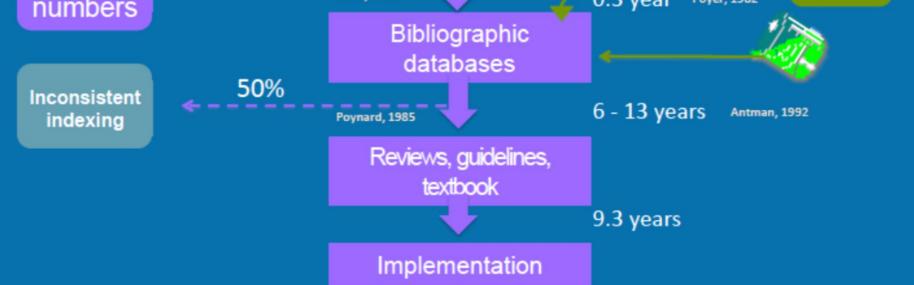
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### Introduction to Implementation Science



#### "PUBLICATION PATHWAY" Balas & Boren, 2000





# The "Know-Do" Gap: MDG 4 – Under 5 Mortality

Preventable

2/3<sup>rd</sup> of these deaths (7

million) can be prevented

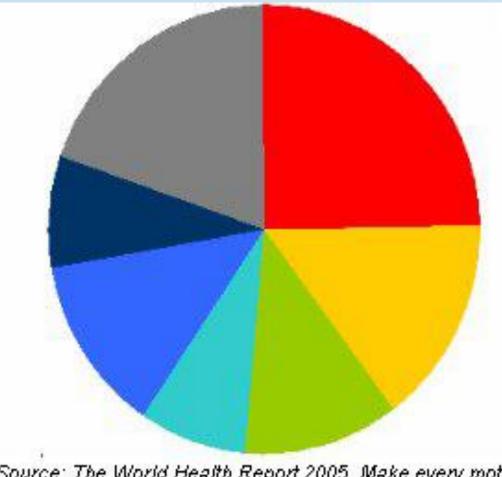
by available, effective and

cheap interventions

Under 5 mortality

11 million children under-5 years old die every year – 90% of them in the developing world

# The Know-Do Gap: MDG 5 -- Maternal mortality is nearly entirely preventable



# Causes of maternal death

- Severe bleeding (haemorrhage) 25%
- Infections 15%
- 🛚 Eclampsia 12%
- Obstructed labour 8%
- Unsafe abortion 13%
- Other direct causes 8%
- Indirect causes 20%

Source: The World Health Report 2005. Make every mother and child count. Geneva, World Health Organization, 2005. The know-do gap: Obesity and diabetes are largely preventable

- The global obesity epidemic is leading to skyrocketing rates of diabetes, CVD, cancer in the US and worldwide
  - Diabetes accounted for a full 12% of health expenditures in 2010 (Zhang P, Diabetes Res Clin Pract; 2010).
  - •60% of diabetes can be prevented by eliminating obesity (Hu et al., *N Engl J Med*, 2001)
  - •Lifestyle intervention trials have reduced diabetes incidence by 58% (US), 29% (India), 42% (China).

Implementation Science --Definitions

"A systematic, scientific approach to ask and answer questions about how to get 'what works' to people who need it with greater speed, fidelity, efficiency, quality and relevant coverage"

"The scientific study of programs and interventions which promote the systematic uptake of clinical research findings and other evidence-based approaches into routine clinical practice and public health policy, hence improving the quality (effectiveness, reliability, safety, appropriateness, equity, efficiency) of health care."

"Implementation science is about determining what works, in real-life full-scale settings."

The implementation pipeline

# efficacy $\rightarrow$ effectiveness and cost-effectiveness $\rightarrow$ implementation $\rightarrow$ dissemination

Features of Implementation and Dissemination Science Research

- 3 types of research with unique features, designs, and objectives
- Theory-driven
- Mixed methods
- Adaptation vs. Fidelity
- Contextual (?when? ?where?) vs. External validity
- Economic evaluation/cost-effectiveness
- Sustainability
- Diffusion (network science)

The Impact of PEPFAR PMTCT Funding on Infant Mortality and Antenatal Care in Kenya

A quasi-experimental evaluation

### Co-authors

Kenya Medical Research Institute/Walter Reed

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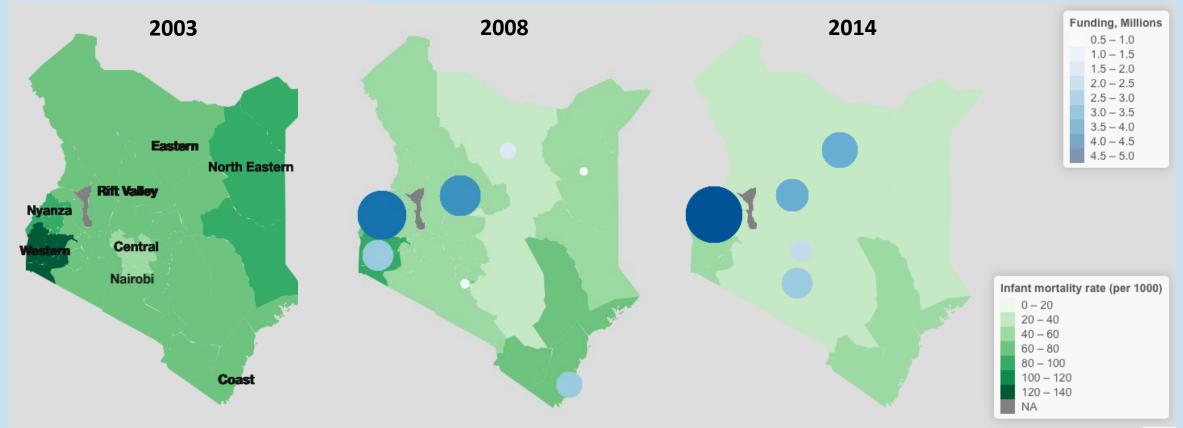
<sup>1</sup>Principle investigator; <sup>2</sup>Associate Investigator; <sup>3</sup>Co-Investigator; <sup>4</sup>Collaborator

# Background

- PEPFAR : the U.S. President's Emergency Plan for AIDS Relief
- PMTCT: Prevention of mother-tochild transmission of HIV
- Kenya has been a PEPFAR-focus country since 2004.
- PEPFAR has invested over \$240 million dollars in PMTCT programs in Kenya since 2004



### PEFPAR funding has coincided with a drastic decrease in infant mortality



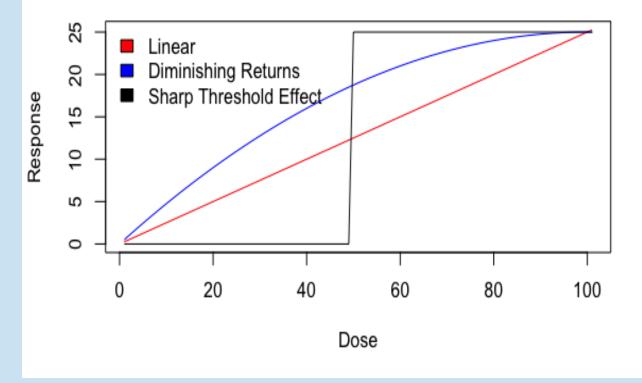
Note: 10-year mortality rates come KDHS reports.

Leafled

### Study Design: A dose-response model

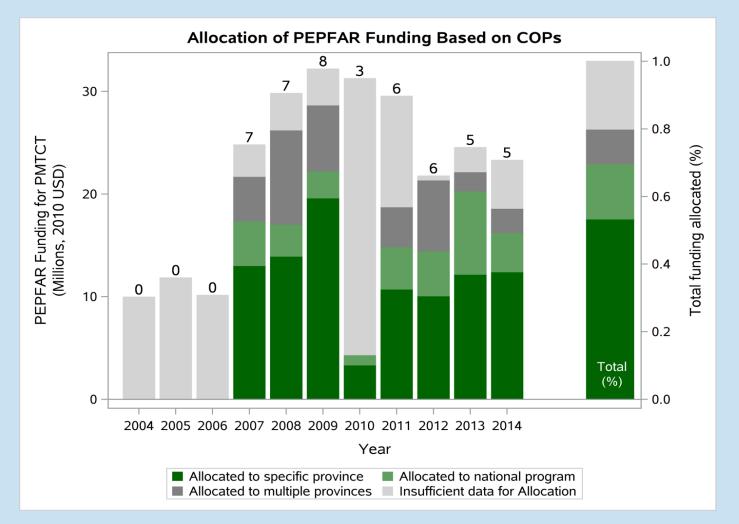
- Quasi-experimental design
- Continous generalization of a difference-in-difference study
- Based on the principle that if a causal relationship exists, a higher "dose" of the exposure should lead to a stronger effect
- Dose-response models do not need to be linear!

#### Illustrative Dose-response Curves



### **Exposure:** PEPFAR funding for PMTCT

- Country Operational Plans (COPs)
  - Submitted annually by all PEPFAR implementing partners
  - Describe planned expenditures
  - Disaggregated by activity, including PMTCT
  - PMTCT funding allocated to provinces based on narrative details in the COPs
- Used to calculate annual per capita funding and cumulative per capita funding



### Outcomes

- Data on individual outcomes came from two nationally-representative, cross-sectional surveys:
  - Kenyan Demographic and Health Survey (KDHS)
  - Kenyan AIDS Indicator Survey (KAIS)
- Use similar methodology and identical or near-identical wording for many HIV/AIDS questions
- Also gather additional demographic and socioeconomic data

NO.	QUESTIONS AND FILTERS	COOND CATEGORIES SKIP
201	Now I would like to talk about something ethe Have you ever heard of an illness called AID	
762	Is there anything a person can do to avoid ge causes AUS?	
		Г
703	What can a person do?	Appendix
	Anything else?	
	Conditional County	Individual Questionnaire
	RECORD ALL MENTIONED.	CONFORMULE KENYA ADDI NICATOR BURVEY INCOMUNA, QUESTIONARE
		DENTFICATION
		PROVINCE*
		NAGEP CLUSTER NUMBER
704	Can people reduce their chances of getting to one sex partner who has no other partners?	LARGE CITYISMALL CITYITOWAIRURAL INAITORAYODISASARSUMUH, NARURURU.CORETTHKAANYERH, SMALL TOWN-3,
		RURALHI) NAME OF HOUSEHOLD HEAD
705	Can a person get the AIDS virus from mosqu	NAME AND LINE NUMBER OF RESPONDENT
706	Can people reduce their chances of getting t	SEX OF RESPONDENT (MALE = 1 FEMALE = 2)
100	condom every time they have sex?	INTERVIEWER VISITS
707	Can a person get the AIDS virus by sharing i	DATE
	has AIDS?	MONTH
708	Can people reduce their chances of getting I	NTERVEWERS YEAR
	having sex at all?	NAMER INT. NUMBER NESALT
709	Is it possible for a healthy-looking person to it	NEXT VISIT: DATE
		TME TOTAL NUMBER OF VISITS
710	Do you know someone personally who has the or someone who died of AIDS?	LAB TECHNICIAN VISITS Lab Tech ID
	and the second	RESULT" RESULT
711	Can the virus that causes AIDS be transmitte	NEXT VISIT: DATE
712	Can the virus that causes AIDS be transmitte	
	During preptancy?	
	During delivery? By braziliseding?	HOME LANGUAGE OF RESPONDENT: 01 EMBU GARKKUYU 07 LUO 10 MUHRENDA 13 ENGLISH 10 KALKUM 05 KSI 01 MAASA 11 SOMAL 14 OTHER
	by pressive (197	03 KAMBA 06 LUHYA 09 MEHU 12 KISWAHIU
		NAME OFFICE EDITOR REVED BY
3		
Ε		"NESULT CODES: (1) COMPLETED (3) NOT AT HOME (3) POSTPONED (4) REPUSED (5) PARTLY COMPLETED (5) INCAPACITATED (7) OTHER REPORTS (7)
		(7) OTHOR (RECEPT) "HEBULTOR (2) HERVER (3) ABSENT

#### Sample pages from DHS and KAIS surveys

### Outcomes

### **HIV testing at ANC**

- Receiving HIV counselling, testing, and test results as part of antenatal care
- N=21,048 mothers who gave birth ≤5 years before the interview date

### **Neonatal mortality**

- Death within the first 30 days of life
- N=37,616 children born 1-60 months prior to the interview date

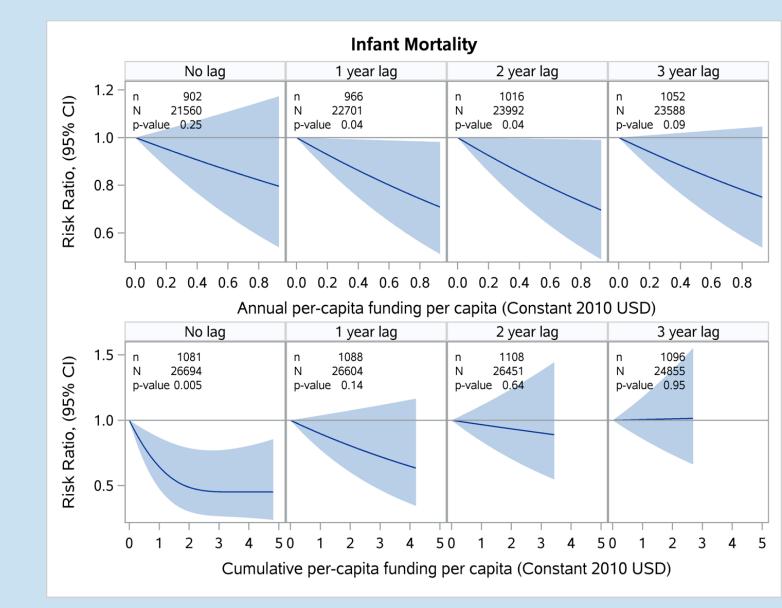
### Infant mortality

- Death within the first year of life
- N=30,424 children born 12-60 months prior to the interview date

# Infant mortality

### Annual per-capita Funding for PMTCT

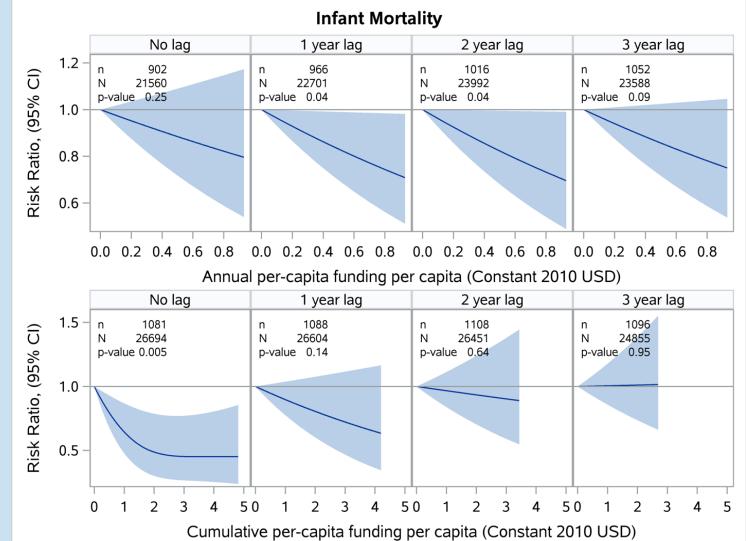
- Not associated with infant mortality in the year of funding allocation
- A \$0.33 increase was associated with:
  - a 11% (95% CI: 1-21%) reduction in infant mortality after a 1-year lag
  - a 12% (95% CI: 0-22%) reduction in infant mortality after a 2-year lag



# Infant mortality

Cumulative per-capita Funding for PMTCT

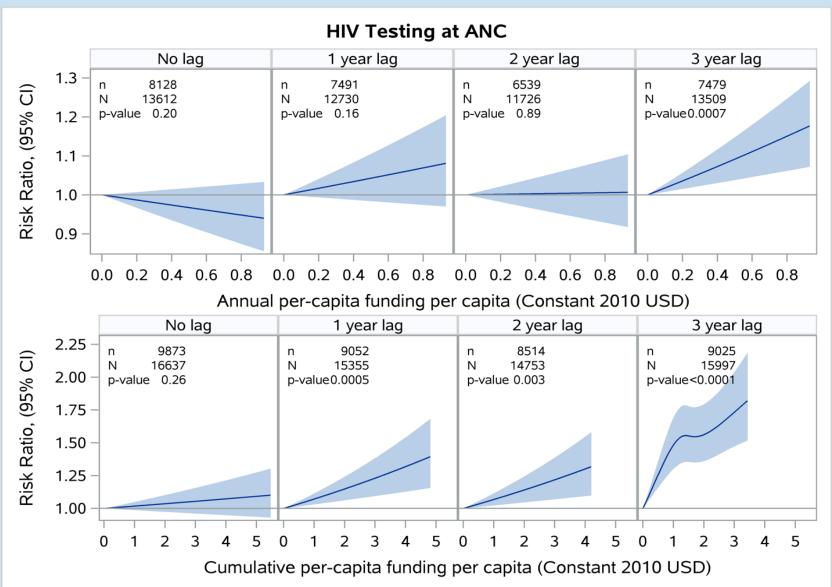
- A \$0.83 increase was associated with a 31% reduction in infant mortality (95% CI: 11-46%) in the year of funding allocation.
- Association became weaker after subsequent lags



# HIV testing at ANC

Annual per-capita Funding for PMTCT

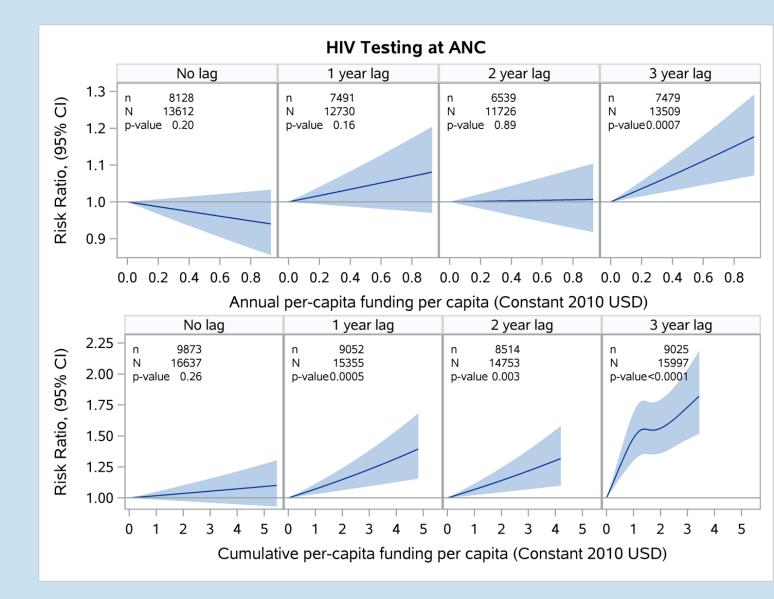
- Not associated with HIV testing at ANC in the year of funding allocation or when using a 1- to 2-year lag
- With 3-year lag, a \$0.33 increase was associated with a 6% increase (95% Cl: 2-10%) in HIV testing at ANC



# HIV testing at ANC

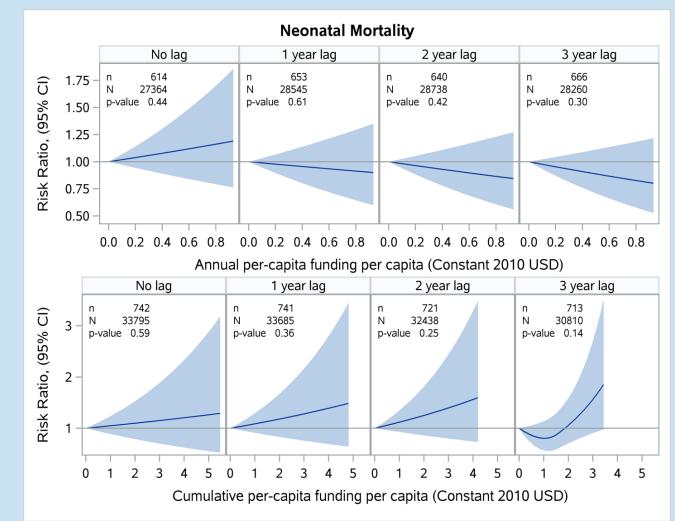
Cumulative per-capita Funding for PMTCT

- Not associated with HIV testing at ANC in the year of funding allocation
- A \$0.83 increase was linearly associated with:
  - 6% increase in HIV testing at ANC after a 1- and 2-year lag (95% CIs: 3-9% and 2-9%, respectively)
  - a 42% (95% CI: 26-59%) increase in testing after a 3year lag



# Neonatal mortality

- Not associated with annual per-capita funding
- Not associated with cumulative per-capita funding



### Delayed Effects of Annual Funding

Annual per capita funding was not associated with health outcomes in the year of allocation but **became beneficial** at later lags

Lagged effect may reflect:

- Logistical delays in transferring PEPFAR funds to implementing partners and local PMTCT programs
- **Biologic realities** of the outcomes of interest



Full impact of funding may not be observable for several years



Donors and program evaluators may wish to consider both logistic and biologic factors when defining a time horizon for program evaluations

### Summary: Main findings

PEPFAR funding for PMTCT was associated with **reduced infant mortality** and **increased HIV testing at ANC** in Kenya.

Our findings are best interpreted as the effect of living in a province that receives a given level of PEPFAR funding and would likely underestimate the benefits of interacting directly with a PEPFARfunded PMTCT program.

# Thank you!

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Isaac Tsikhutsu, MBChB, MMED<sup>1</sup> Fredrick Sawe, MBChB, MMED<sup>2</sup> Jane Mumbi, BSN<sup>2</sup> Duncan Kirui<sup>2</sup> William Sugut<sup>2</sup>

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