

\*\*\*PRELIMINARY: DO NOT CITE\*\*\*

# “The Risk of Asking”

or, more formally,  
Being Surveyed  
Can Change Later Behavior  
and Related Parameter Estimates

Presented by Clair Null  
WSH Evaluation Convening  
UC Berkeley  
March 31, 2011

# Data Collection is an Intervention

- Social interactions are an important determinant of many behaviors
  - “interpersonal communication” is key
- Surveys are a social interaction between respondent & enumerator (whether we want them to be or not)
- Does the data collection process change respondents’ later behavior?
  - Possibly even affecting the outcome of interest
- Even if these changes are an identical level shift for treatment and comparison groups, they can still affect conclusions regarding the outcome of interest <sup>2</sup>

# Measuring Diarrhea

- Common approach: self-reported
  - Subjective “three or more loose or watery stools”
  - (Bi)weekly visits for many weeks to monitor
    - Motivated in part by concern about recall bias
    - Increases compliance with intervention for efficacy studies
- Potential biases in treatment effect estimates
  - Social desirability bias
  - Courtesy bias
  - Respondent fatigue
  - Question-behavior / mere measurement / self-prophecy
  - Survey effect (surveys → behavior change)
  - Hawthorne effect (observation → behavior change)

# Our Research Contribution

- Think of data collection frequency as a “treatment” that can be randomized
  - Compare frequent and infrequent monitoring
    - Only other paper to do this is McCarney et al., 2007 BMC Med. Res. Method. (Ginko biloba & dementia)
- Quantify the bias in measured diarrhea prevalence in biweekly self-reported data
  - Survey/Hawthorne effect is large
  - Changes conclusions about health effects of a water quality improvement project

# Evolution of Our Research:

## Estimate Health Effects of Spring Protection

### **First Approach (Economist-style)**

- randomize 180 communities into phases of intervention
- random sample of 8 households in each community
- 4 survey visits  
(approximately 6-12 months)

### **Second Approach (Epidemiologist-style)**

- randomize last phase of communities (n=76) into 2 final waves of intervention
- random sample of 2 households in each community drawn from existing sample
- 19 survey visits  
(every 2 weeks)

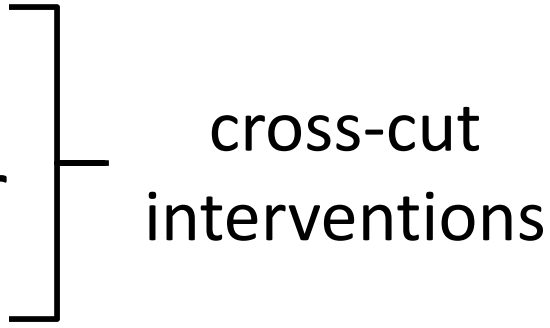
# Data Collection Frequency: Another “Treatment”?

- Randomly selected households for biweekly monitoring from among those who had already been surveyed 4 times (BWM households)
- Diarrhea rates started to fall drastically among this group compared to previous 6-12 month surveys
- Was this because of the surveying?
- Selected another random subset of original households for comparison (extension households); continued to survey them every 6 months

# Summary of “Treatments”

- Frequency of data collection  
(BWM versus extension)

## Rural Water Project (RWP):

- Spring protection
  - Distribution of dilute chlorine for point-of-use water treatment
- 
- cross-cut interventions

# Summary of RWP “Treatments”

<p>Spring Protection Phase 1 (treatment)</p>	<p>Spring Protection Phase 2 (control)</p>
--	--



# Summary of RWP “Treatments”

Chlorine Distribution  
(treatment: Households given a 6-month supply of chlorine  
and an improved water storage container)

Chlorine Distribution  
(control)

# Summary of RWP “Treatments”

Protected Spring + Chlorine	Unprotected Spring + Chlorine
Protected Spring, No Chlorine	Unprotected Spring, No Chlorine

# Summary of RWP “Treatments”

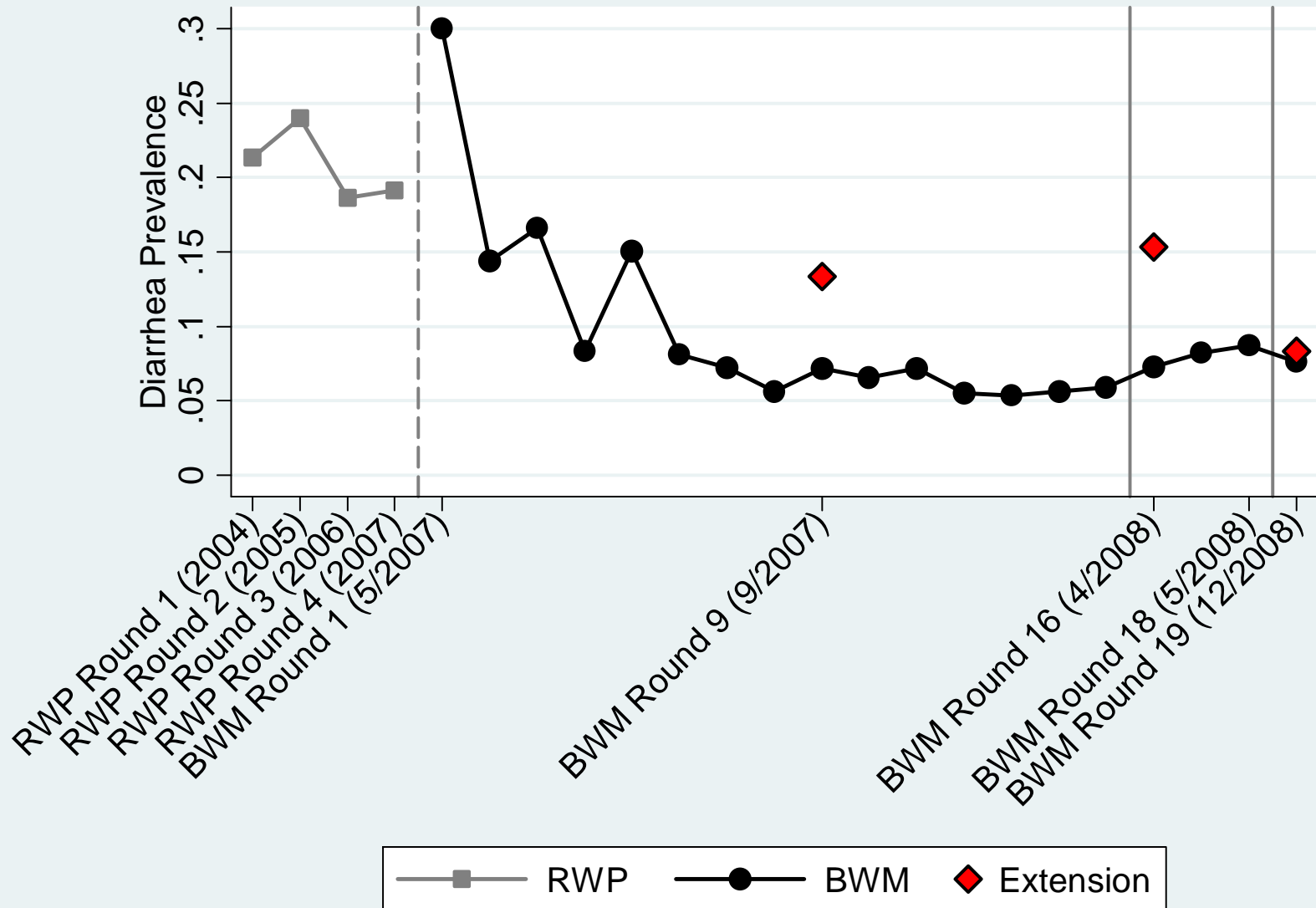
Protected Spring + Chlorine	Unprotected Spring + Chlorine
Protected Spring, No Chlorine	Unprotected Spring, No Chlorine

**BWM,  
extension**

# BWM & Extension: Samples and Timeline

- Both sub-samples *randomly chosen* from larger sample using same methodology
  - BWM n = 170, extension n = 160
- BWM surveys began in May 2007
  - Springs protected between rounds 5-8
- Extension households surveyed 3 times
  - BWM rounds 9 (9/2007), 16 (4/2008), 19 (12/2008)
- Interruptions
  - Kenyan political crisis (12/2007 – 3/2008)
  - Supplies delayed by customs (6/2008 – 11/2008)

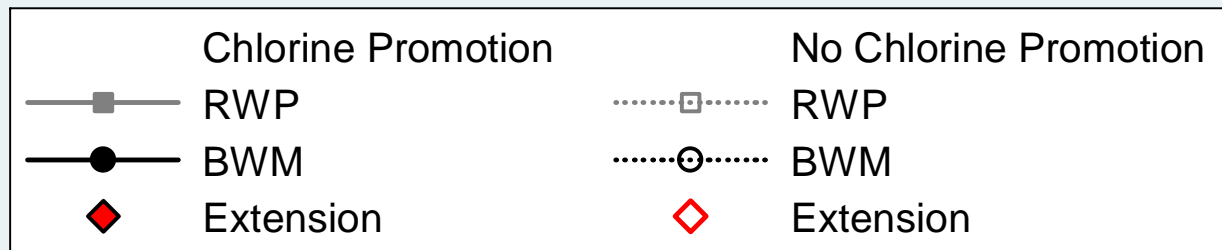
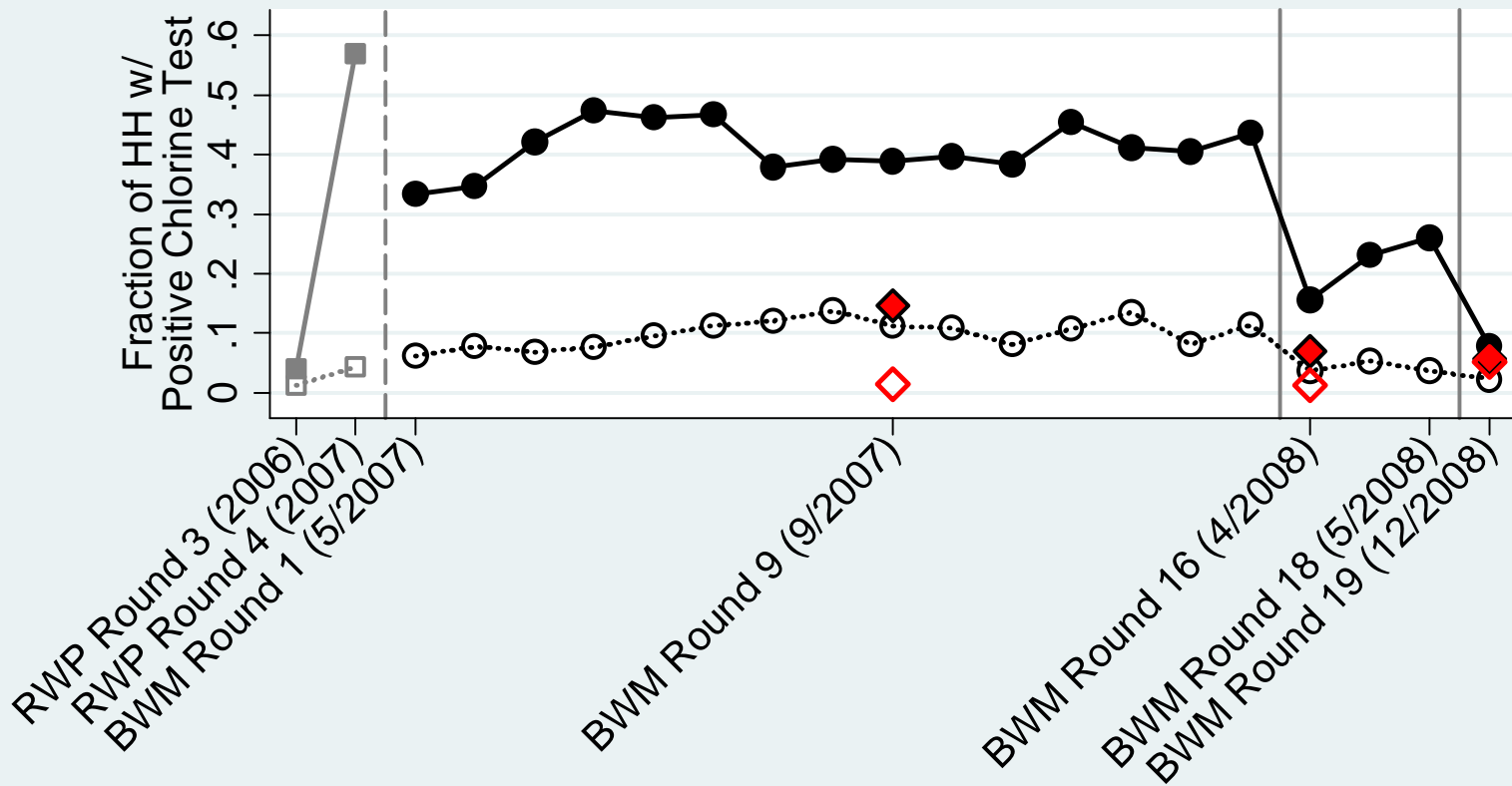
# Diarrhea Trends



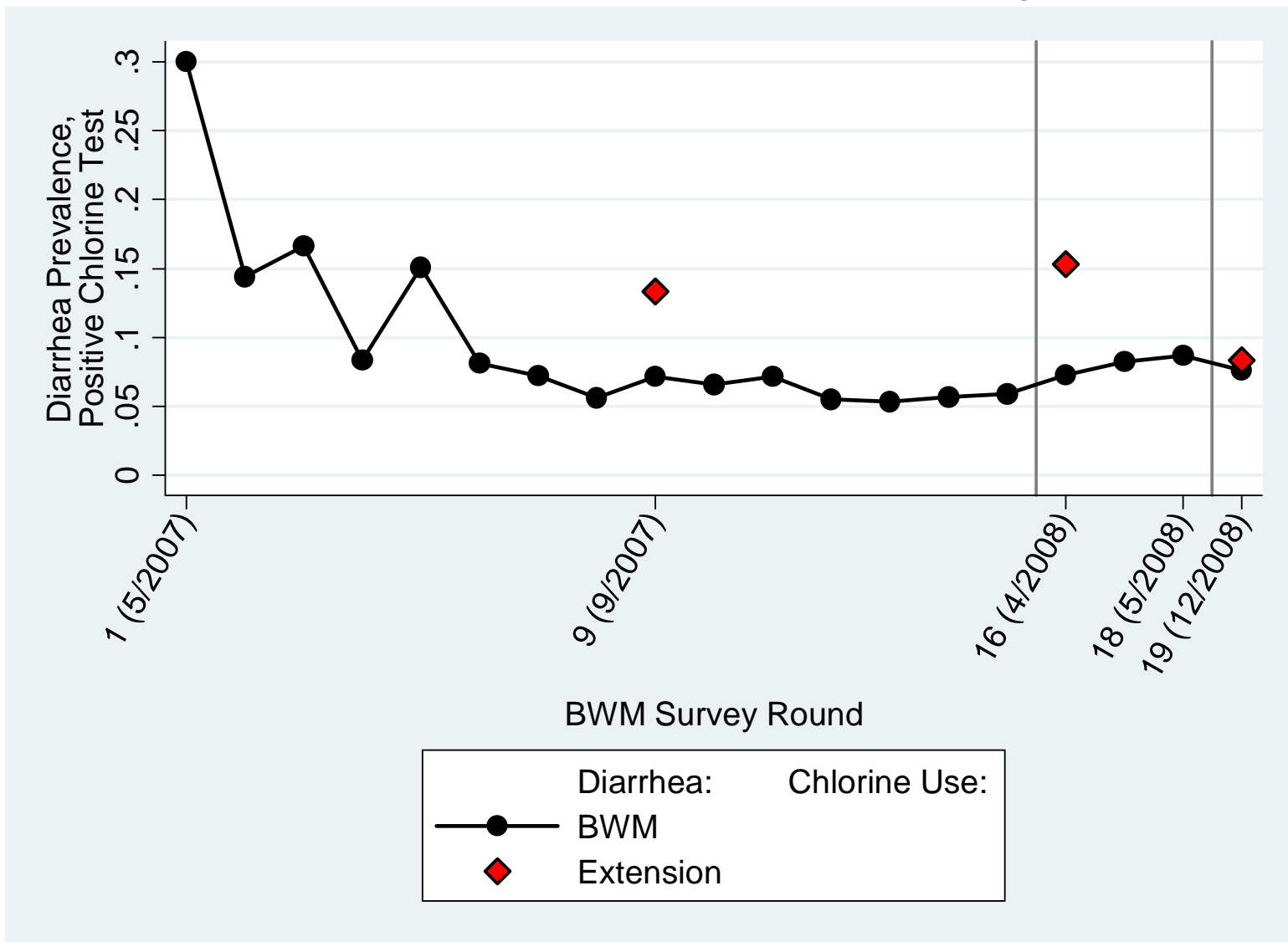
# Explaining Falling Diarrhea Prevalence

- Changes in reporting
  - Courtesy bias
    - Subjects aware that NGO that conducted BWM surveys also implemented water quality improvement interventions
  - Survey fatigue
    - Basic survey lasted around 20 minutes; reporting diarrhea symptoms triggered additional survey questions (up to 5 minutes per child) and in some cases a mini-exam (an additional 5 minutes)
- Changes in behavior
  - Prevention (water treatment)
  - Treatment (oral rehydration therapy)
  - Others?

# Chlorination Trends

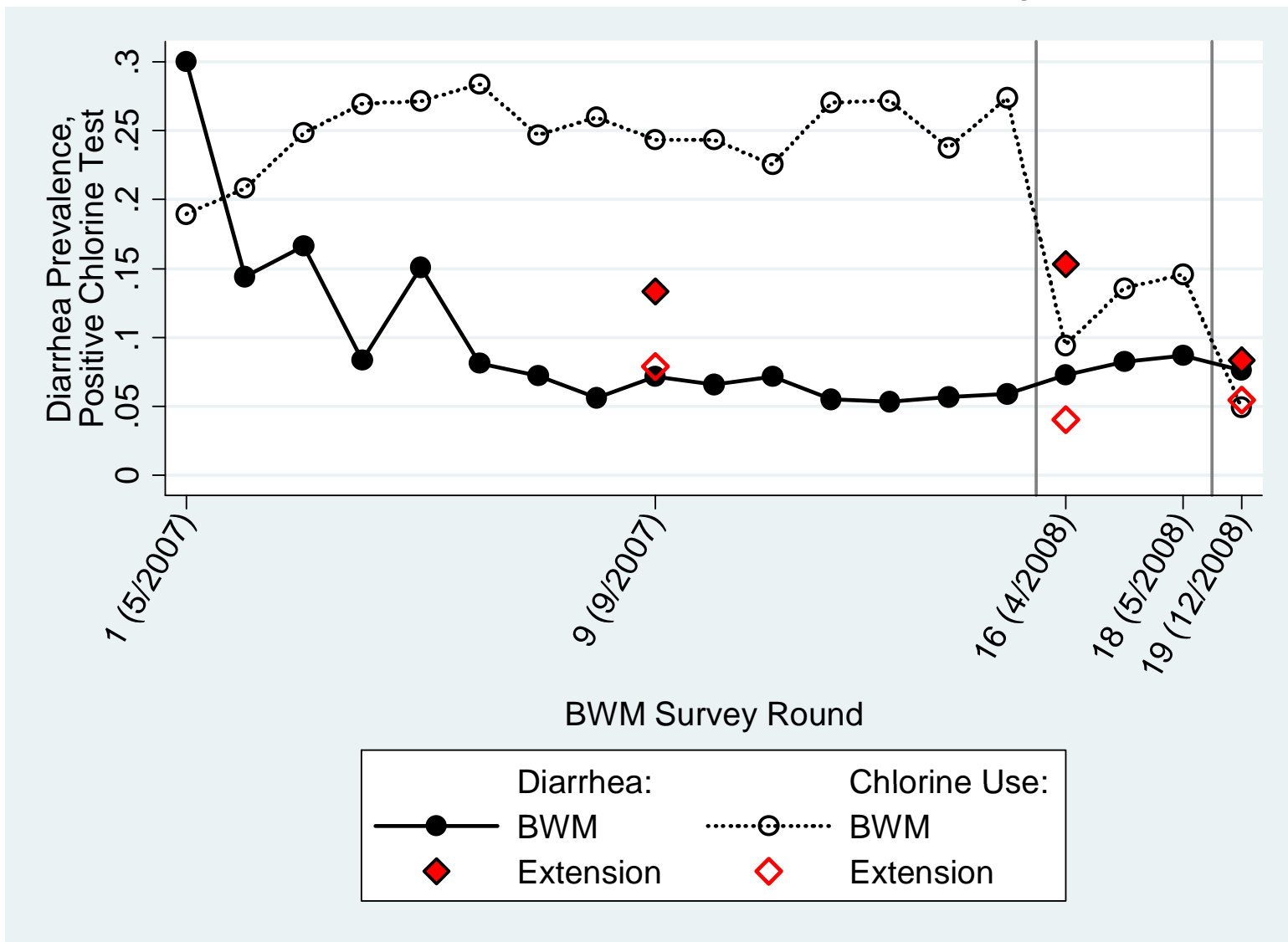


# How Much of the Fall in Diarrhea Can Increased Chlorination Explain?





# How Much of the Fall in Diarrhea Can Increased Chlorination Explain?



# How Much of the Fall in Diarrhea Can Increased Chlorination Explain?

- **If effect of chlorination on diarrhea is the same in this sample as in RWP** (35% reduction, comparable to other health studies of chlorination), accounting for different usage rates, then difference in chlorination between BWM and extension households in round 9 would lead to 12% lower diarrhea among BWM relative to extension
- Since BWM had 50% less diarrhea than extension in round 9, **chlorination alone could explain roughly a quarter of the difference between BWM & extension households**

# Comparing Results: Health Effects of Spring Protection by Survey Frequency

**Table 1. Impact of more-frequent surveys on chlorine use, diarrhea, and spring protection treatment effect stability in Kenya**

Parameter	Child diarrhea (self-report) in the past week—survey round 9 (1)	Child diarrhea (self-report) in the past week—panel data (2)	Chlorine in home drinking water container—survey round 9 (3)	Chlorine in home drinking water container—survey round 16 (4)	Child diarrhea (self-report) in the past week—panel data (5)
Surveyed more frequently	-0.068* (0.022)	-0.149* (0.026)	0.150* (0.050)	0.053 <sup>†</sup> (0.028)	-0.182* (0.033)
Protected spring					-0.104* (0.034)
Surveyed more frequently × Protected spring					0.135* (0.048)
Mean of dependent variable	0.081	0.140	0.164	0.068	0.140
No. of observations	713	4,296	293	309	4,296

Ordinary Least Squares estimates with Huber-White SEs in parentheses, clustered at the spring level. Specifications reported in columns 2 and 5 include child fixed effects, because the data used there are an unbalanced panel.

\*Significant at 1%.

<sup>†</sup>Significant at 10%.

Different conclusions about reduction in diarrhea due to spring protection, depending on frequency of data collection

# 4 Other Studies

- Marketing firm conducted a baseline among a random subsample of the target population
- Microfinance agency offered everyone a product
  - 2 health insurance products, 2 microloans
- Take-up of the products was collected from administrative data
  - Higher for rates for health insurance if surveyed at baseline; no effect for microlending
  - No effect on price elasticity of demand
- Even a baseline alone affected behavior
  - Draws attention to choices you might not otherwise consider

# Conclusions

- High frequency data collection appears to have led to unreliable measures of child diarrhea prevalence
- We found objective evidence that at least one confounding behavior (chlorination) changed as a result of frequent data collection (social interactions)
  - Even much less intensive baseline surveys changed outcomes
- This in turn partially explains why estimates of the health effects of spring protection differ depending on data collection frequency
  - Frequent surveys → chlorination, which could be a substitute for spring protection (both lead to cleaner water)

# Hawthorne Effects Change Unobservables

- Chlorine use is just one of many behaviors that could have affected the parameter of interest
- But what about all sorts of other behaviors that we didn't / couldn't objectively observe?
- Self-reported outcome variables are problematic, but objective outcomes won't eliminate Hawthorne bias
- Blinding subjects and/or enumerators to the treatment also isn't enough to avoid the problem

# Recommendations

- Studies that use frequent data collection should report time trends in the outcomes of interest
- Future studies should randomly assign data collection frequencies so that it is possible to control for survey/Hawthorne effects
  - Large infrequent samples may preferred to high frequency data collection
  - Even baseline surveys might have consequences

# Acknowledgements

Coauthors: Alix Peterson Zwane, Jonathan Zinman, Eric Van Dusen, William Pariente, Edward Miguel, Michael Kremer, Dean Karlan, Richard Hornbeck, Xavier Gine, Esther Duflo, Florencia Devoto, Bruno Crepon, Abhijit Banerjee

Citation: PNAS (2011) volume 108, number 5, pp. 1821-1825  
<http://www.pnas.org/content/108/5/1821>

Funding for frequent surveying from The Bill and Melinda Gates Foundation  
Funding for the RWP from the Hewlett Foundation, USDA/Foreign Agricultural Service, International Child Support (ICS), Swedish International Development Agency, Finnish Fund for Local Cooperation in Kenya, google.org, Harvard University Center for International Development Sustainability Science Program, and The Bill and Melinda Gates Foundation



*Thank You!*