

**Economics 172**  
**Issues in African Economic Development**

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Lecture 6 – February 1, 2007



## Health and wealth: cause or effect?

- How can we determine whether poor health is the cause of poverty (as Bloom and Sachs assert) or vice versa?
- This is a difficult problem
- More generally how to interpret  $\text{Corr}(A, B) > 0$  ?
  1. "A causes B":  $A \rightarrow B$
  2. "B causes A":  $B \rightarrow A$
  3.  $A \rightarrow B$  and  $B \rightarrow A$  simultaneously
  4. Some other factor C causes both:  $C \rightarrow A$  and  $C \rightarrow B$
  5. The association is purely coincidental (but regression confidence intervals help address this)

## Another approach: analysis with “micro-data”

- Both Bloom and Sachs (1998) and AJR (2001) focus on broad country-level historical trends
- But establishing causality and theoretical channels is exceedingly difficult in that setting
- Another approach uses data at the level of individuals, communities, or firms to test theories about the link between health and wealth
- Problem Set #1 features some analysis of this kind

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Poor health

Lower income

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- Worms are among the world's most prevalent diseases:

<u>Parasite</u>	<u># infections globally</u>
Hookworm	1.3 billion
Roundworm	1.3 billion
Whipworm	900 million
Schistosomiasis	200 million

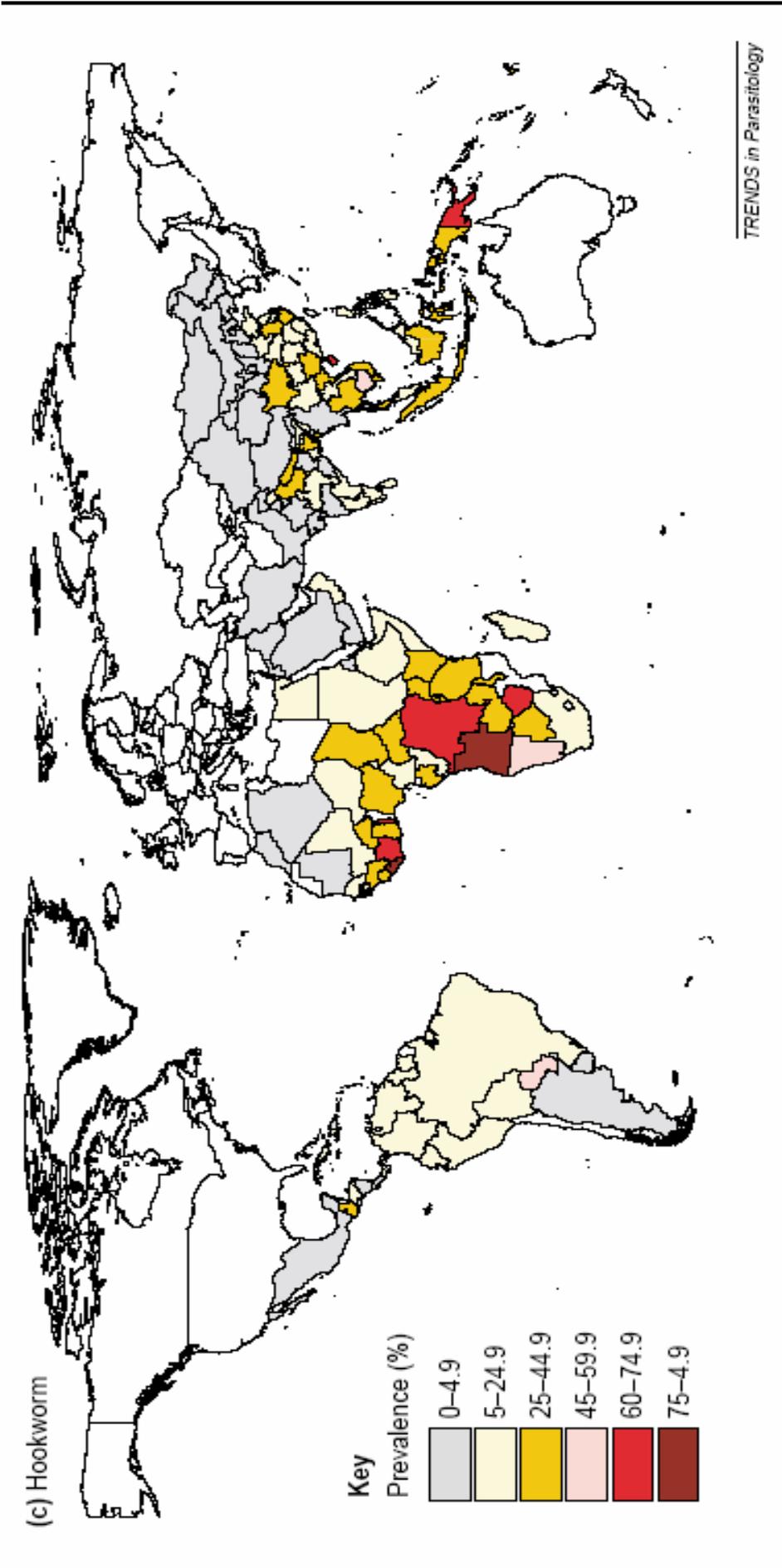


Figure 1. The global distribution of (a) *Ascaris lumbricoides*, (b) *Trichuris trichiura* and (c) hookworm. White areas represent countries not included in the present analysis. Data obtained from <http://www.fic.nih.gov/dcpp/dcp2.html>

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- Health and nutritional consequences – anemia, weakness, listlessness, stunting, wasting, stomach pain  
– especially for heavy infections

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- Treatment is cheap (<US\$1 per year)
  - Drugs: albendazole, praziquantel

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$e$ : white noise “error/disturbance” term,  $E(e) = 0$

$i$ : denotes person “ $i$ ” in the population,  $i$  from 1, ..., N

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Expectations (“average”) of the outcome for different cases:

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**True effect**

“Omitted variable/selection bias” term

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*“Experimental” variation in variables of interest*
- If treatment with deworming drugs is assigned “randomly”, and reduces worm infection, then the reduction in worms should be uncorrelated with  $X$

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- Primary School Deworming Project (PSDP) in Kenya

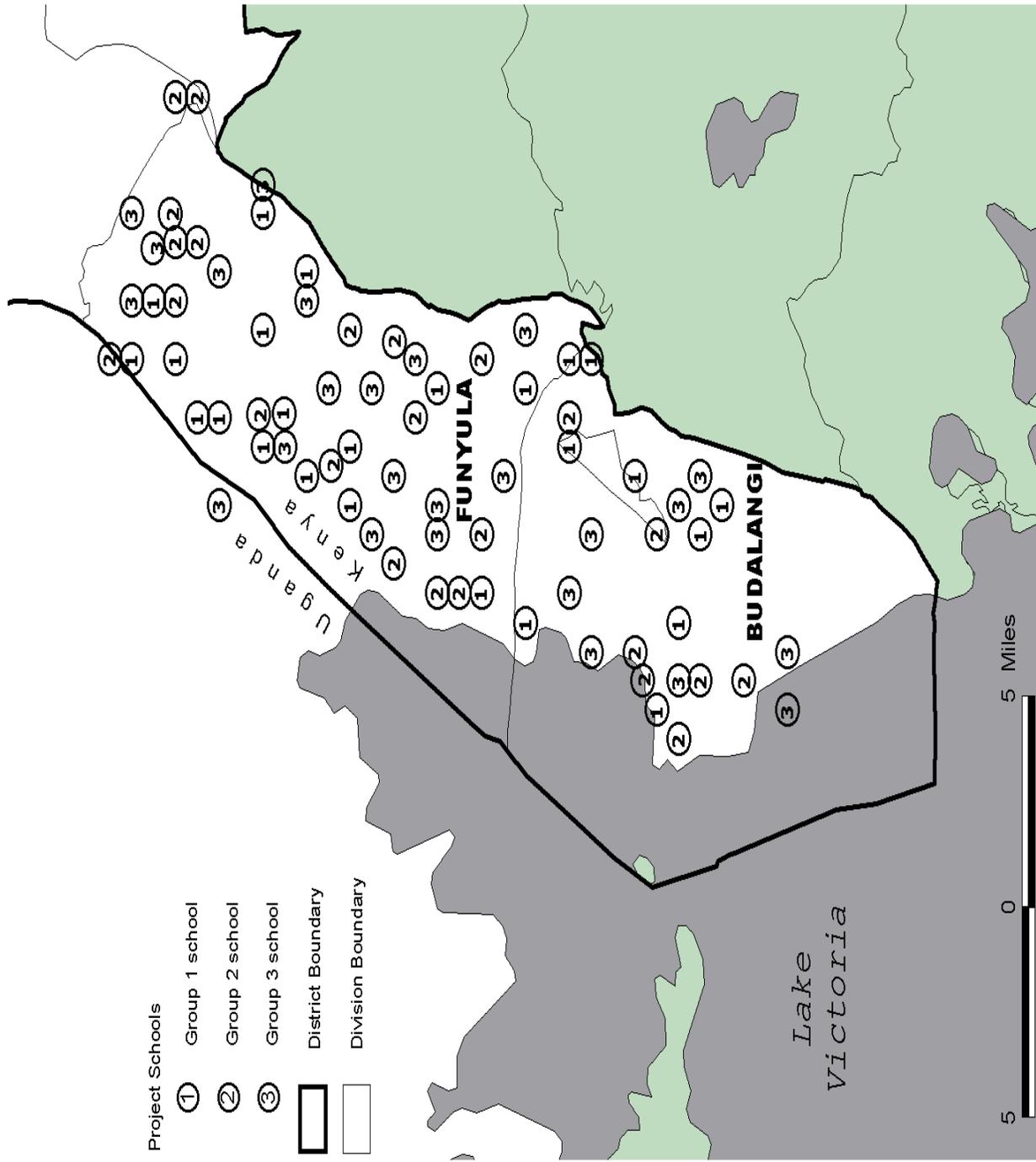
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	<u>1998</u>	<u>1999</u>
Group 1 (25 schools)	Treatment	Treatment
Group 2 (25 schools)	Comparison	Treatment
Group 3 (25 schools)	Comparison	Comparison





# PSDP Baseline differences (1998)

Average	Average	Average
<u>Group 1</u>	<u>Group 2</u>	<u>Group 3</u>
(N=25)	(N=25)	(N=25)

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	<u>Average</u> <u>Group 1</u> (N=25)	<u>Average</u> <u>Group 2</u> (N=25)	<u>Average</u> <u>Group 3</u> (N=25)
Latrine at home	0.82	0.81	0.82
Livestock at home	0.66	0.67	0.66
Child sick often (self-reported)	0.10	0.10	0.08
1996 exam score (normalized mean 0, s.d. 1)	-0.10	0.09	0.01

# Program impacts

- Three types of analysis:
  - (1) Direct treatment effects: simple difference between treatment and comparison schools
  - (2) Within-school externality impacts
  - (3) Cross-school externality impacts

# Health, nutrition impacts (1999)

	<u>Group 1</u>	<u>Group 2</u>	<u>G1-G2</u>
Rate of moderate-heavy infection	0.27	0.52	-0.25*
Sick in past week	0.41	0.45	-0.04*
Height for age (Z-score)	-1.13	-1.22	0.09*
Hemoglobin (Hb)	124.8	123.2	1.6

# Within-school infection externalities (1999)

<u>Group 1</u>	<u>Group 1</u>	<u>Group 2</u>
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<u>(Treated)</u>	<u>(Untreated)</u>	
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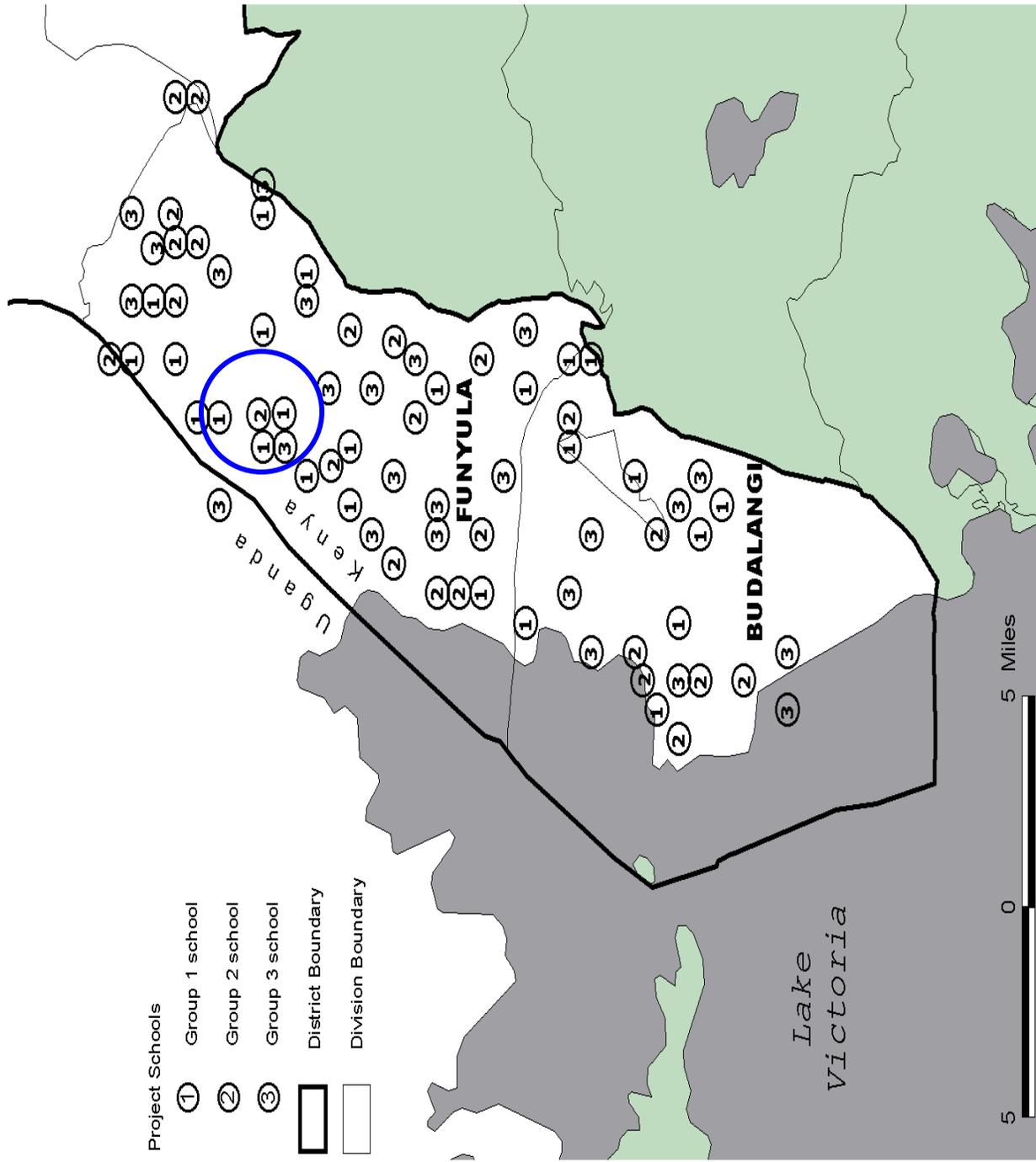
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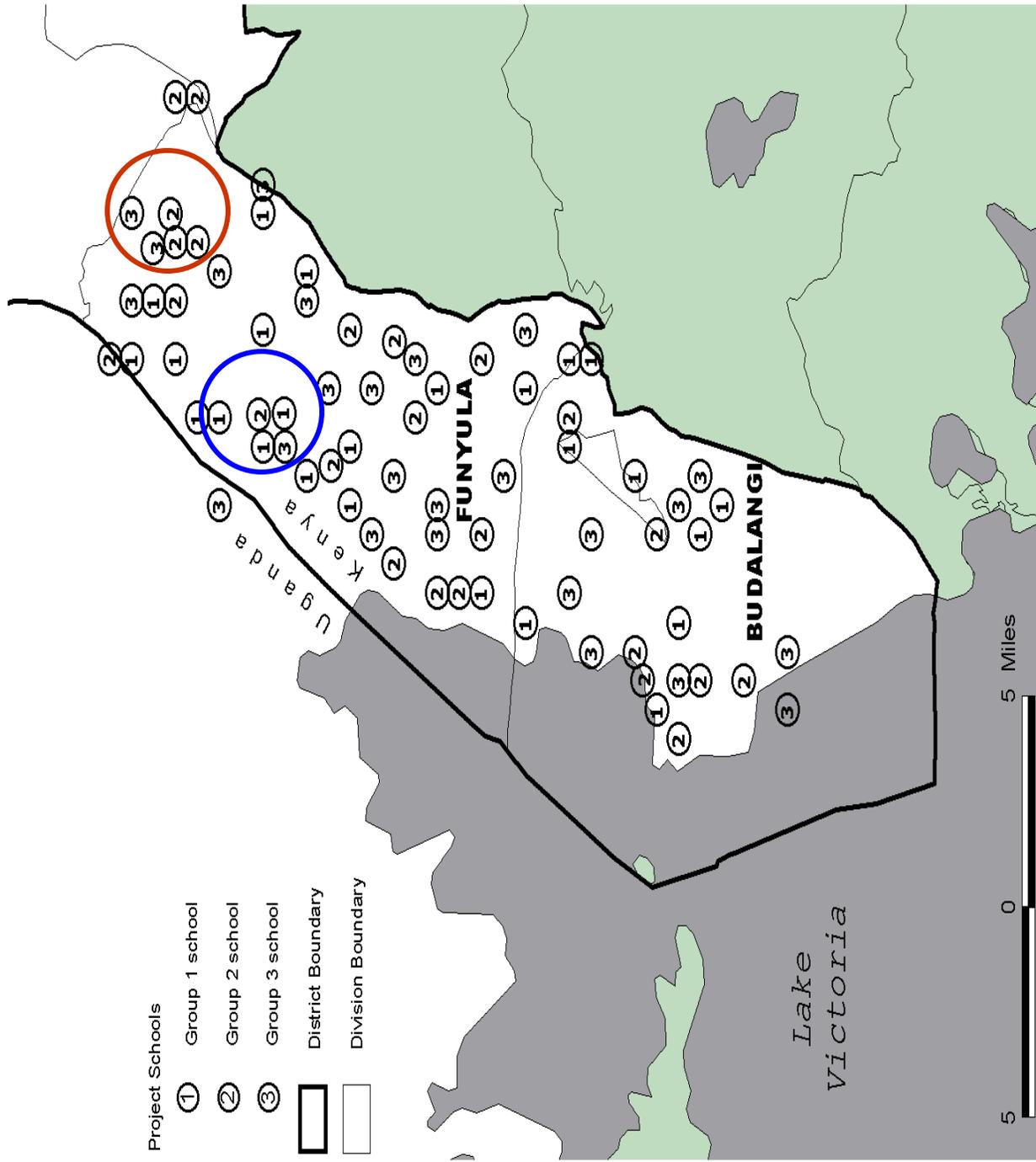
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## Cross-school infection externalities (1999)

- Large reductions in moderate-heavy infection levels within 6 km of treatment schools in 1999





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- Large reductions in moderate-heavy infection levels within 6 km of treatment schools in 1999
- An average reduction in moderate-heavy infections of 23 percentage points in the study area can be attributed to cross-school externalities

# Educational impacts – school participation

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Group 1 (T) Group 2 (C) Group 3 (C)

Younger girls, and all boys

0.84                      0.73                      0.77

Older girls ( $\geq 13$  years)

0.86                      0.80                      0.81

Pre-school, Grades 1-2

0.80                      0.69                      0.70

Grades 6-8

0.93                      0.86                      0.89

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  - Congestion effects in the classroom
  - Time lags

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- Deworming benefits are at least three times (3x) as large as treatment costs (using the Tanzania costs)

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  - (2) Socio-cultural explanations / resistance to new technologies (evidence from anthropology)

# The Impact of Higher Drug Costs

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- In 2001, parents in 25 randomly chosen Group 1 and Group 2 schools paid US\$0.10-0.30 per child
- 2001 deworming take-up:  
Free-treatment schools: 75%  
Cost-sharing schools: 18%



- For next time: Read Miguel (2005)

# Whiteboard #1

# Whiteboard #2

# Whiteboard #3

# Whiteboard #4

# Whiteboard #5

