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Evidence based climate risk management in health: the case of epidemic malaria

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EVIDENCE BASED CLIMATE
RISK
MANAGEMENT IN HEALTH: THE
CASE
OF EPIDEMIC MALARIA

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KEMRI
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Sensitivity of VBD to Climate Change

HIGHLY SENSITIVE

MALARIA

DENGUE

CHIKUNGUNYA

RIFT VALLEY FEVER

LOW SENSITIVITY

SCHSTOSOMIASIS

FILARIASIS

LEISHMANIA

TRYMANOSOMIASIS

YELLOW FEVER



NTD SENSITIVITY TO CLIMATE CHANGE

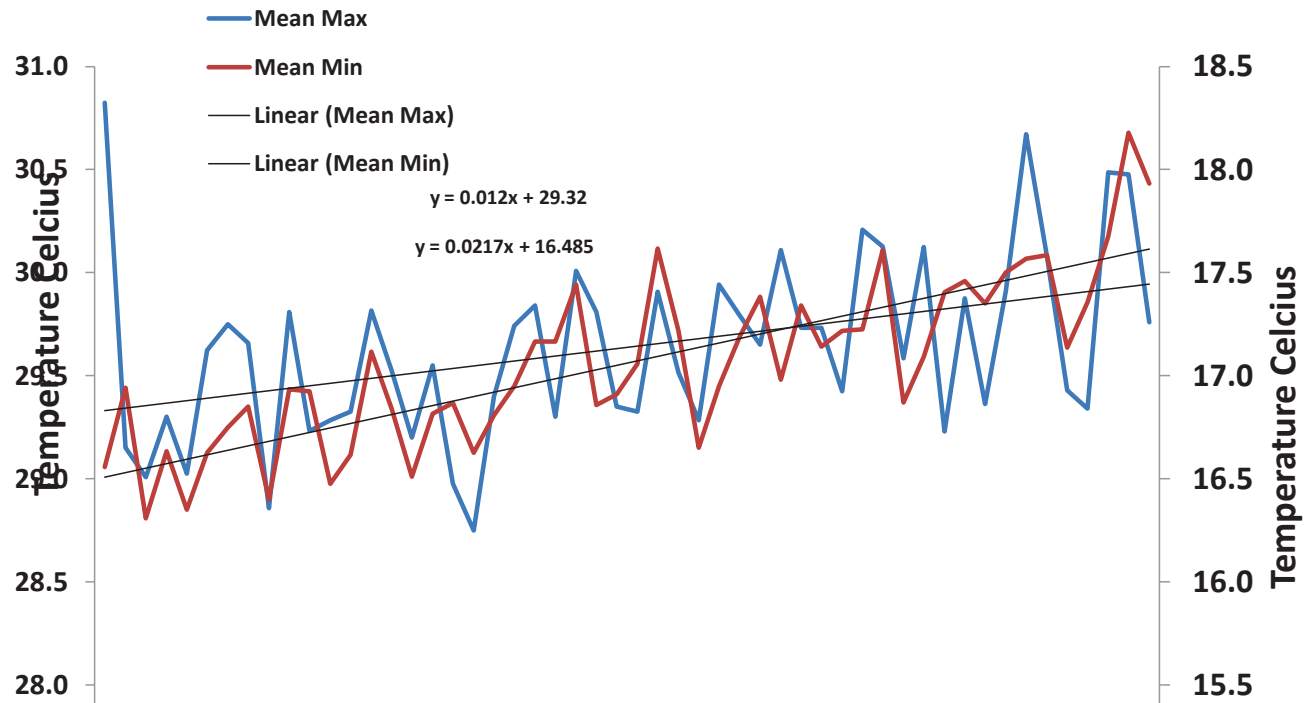
- Major NTDS in Kenya are schistosomiasis, soil helminths, filariasis, leishnaniasis hydatid disaese
- The above pathogens are compleex organisms that have low sensitivity to changes in temperature. They exisst in endemic forms
- Dengue and other viral pathogens are highly sensitive to temperature variations and exist in endemic and epidemic forms.



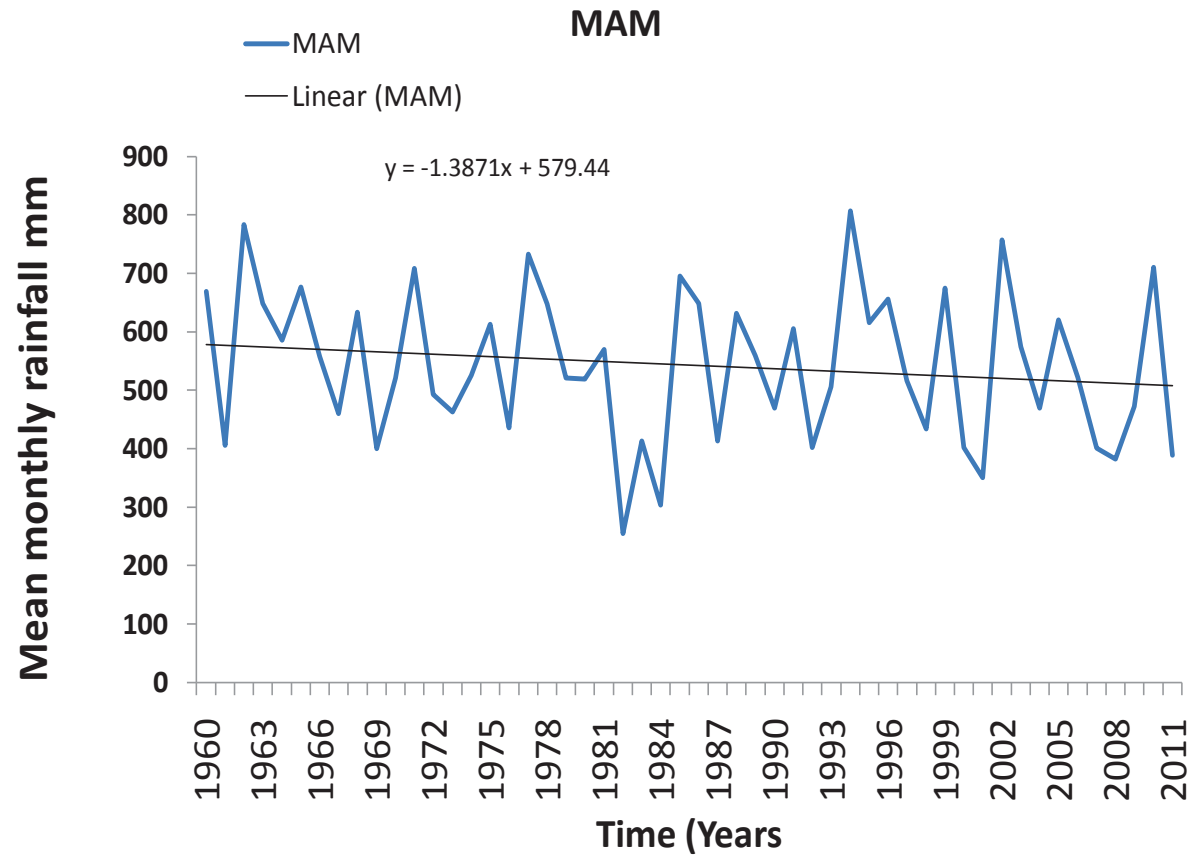
Effects of climate change and variability on VBD

- Climate change increases the suitability of new areas to disease transmission
- Climate variability drives epidemics

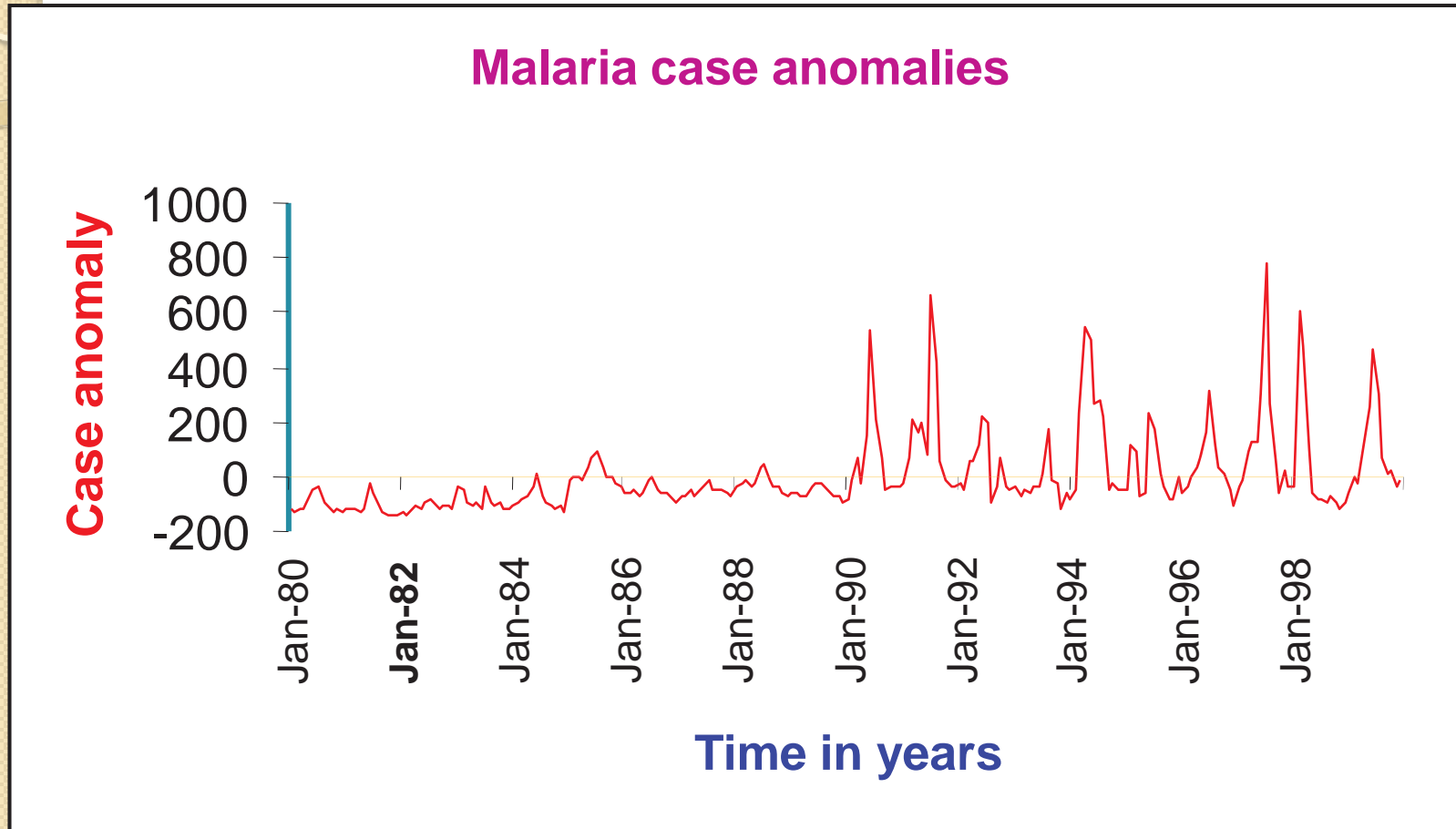
Mean annual temperature trends Kisumu



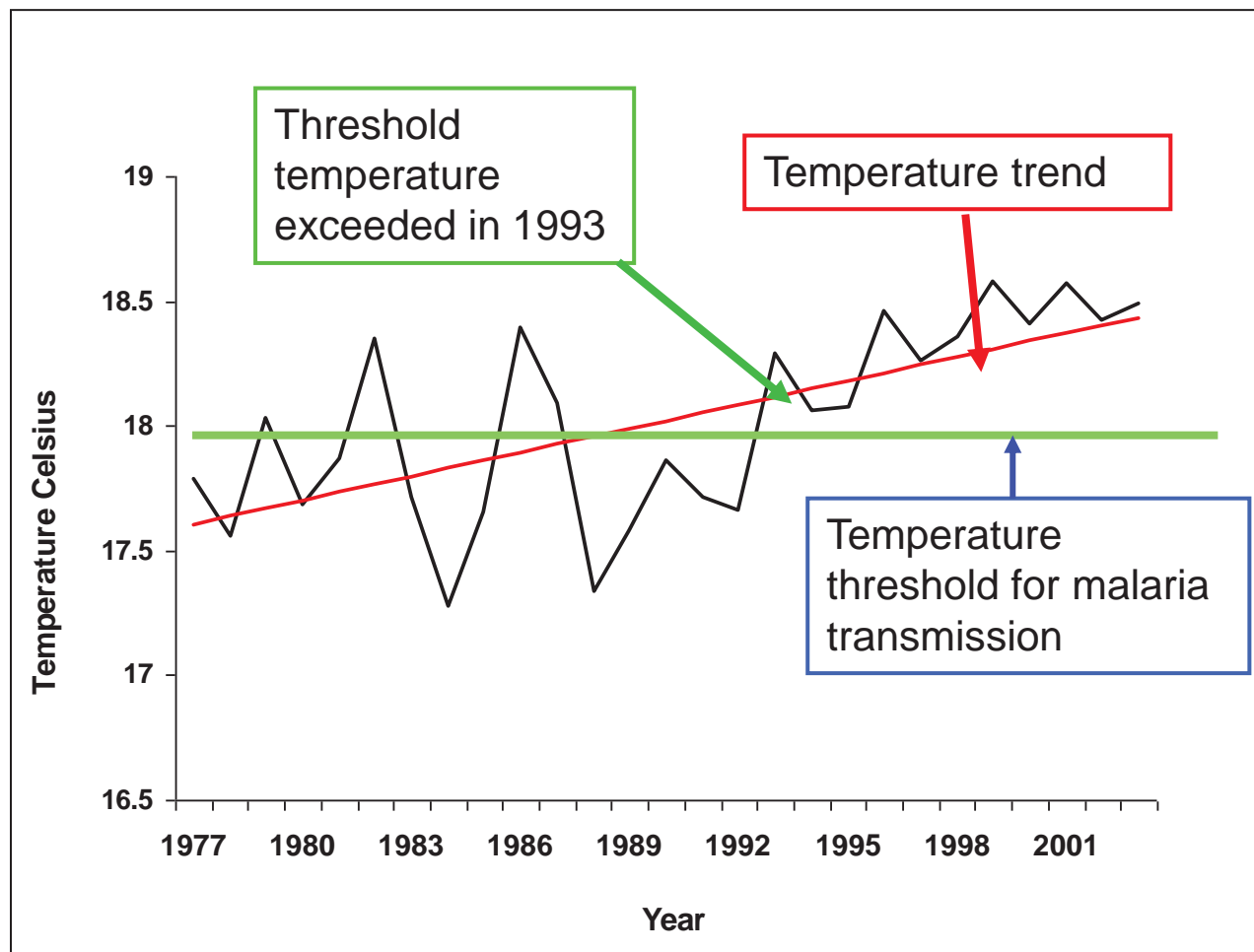
Rainfall trends March April May in Kisumu



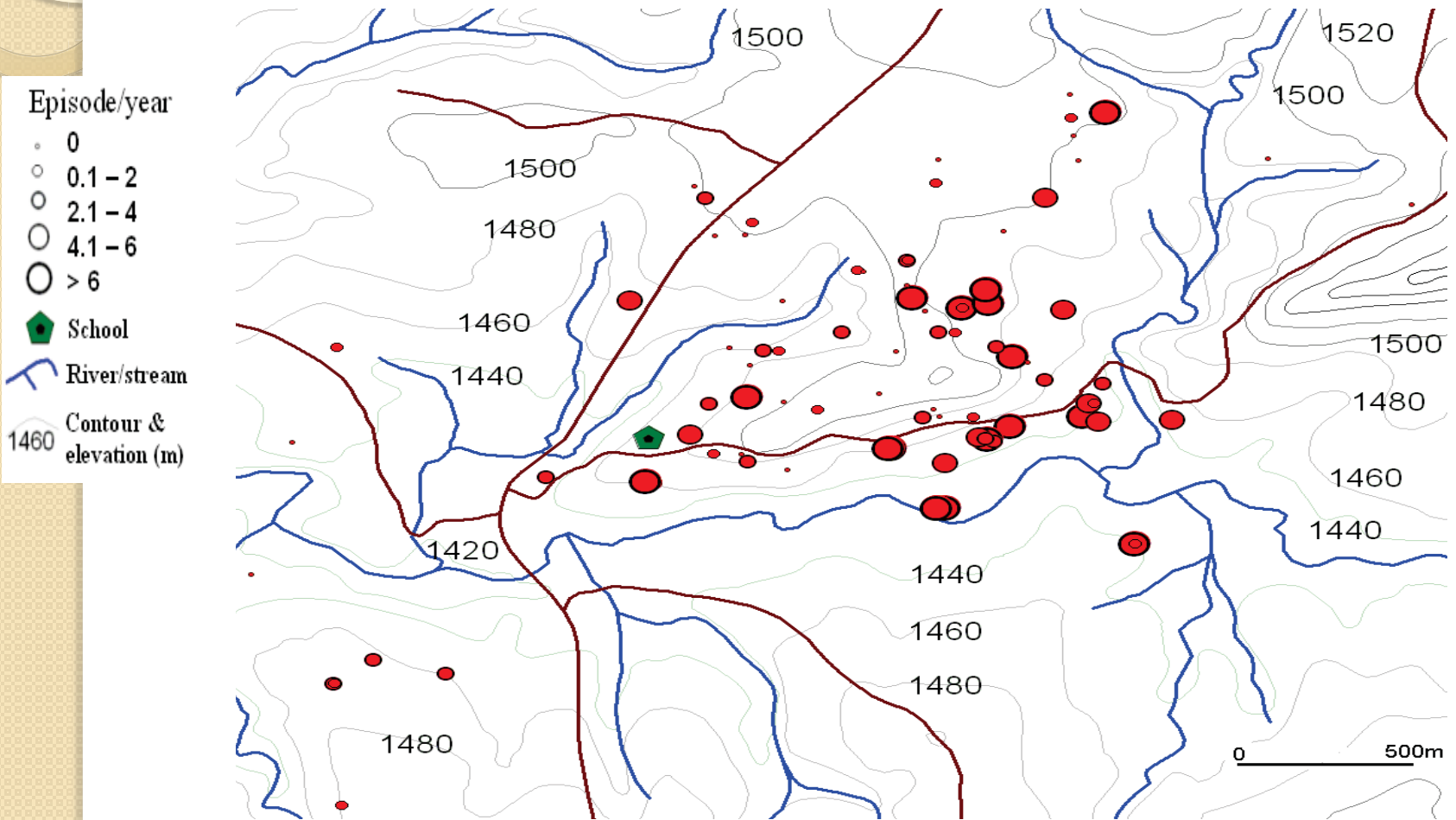
Malaria case anomalies in Nandi district western Kenya associated with El Nino weather



Malaria spreads to Central Kenya highlands because of climate change



Many episodes of malaria per year in “U” shaped valley



Many episodes of malaria per year in “U” shaped valley

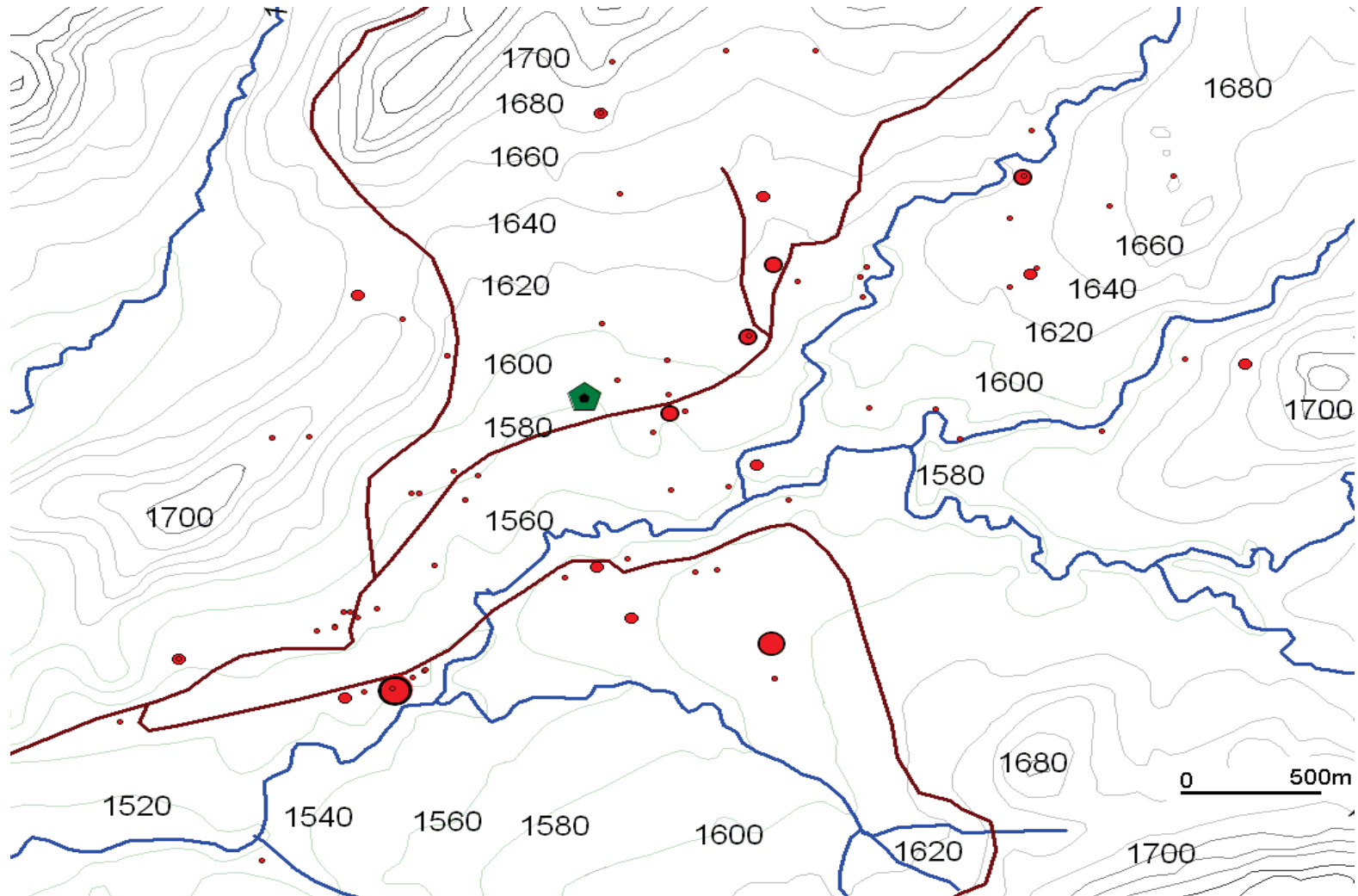
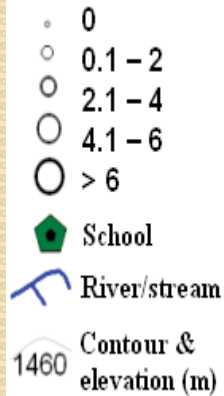


A V shaped valley with a good drainage: poor mosquito breeding

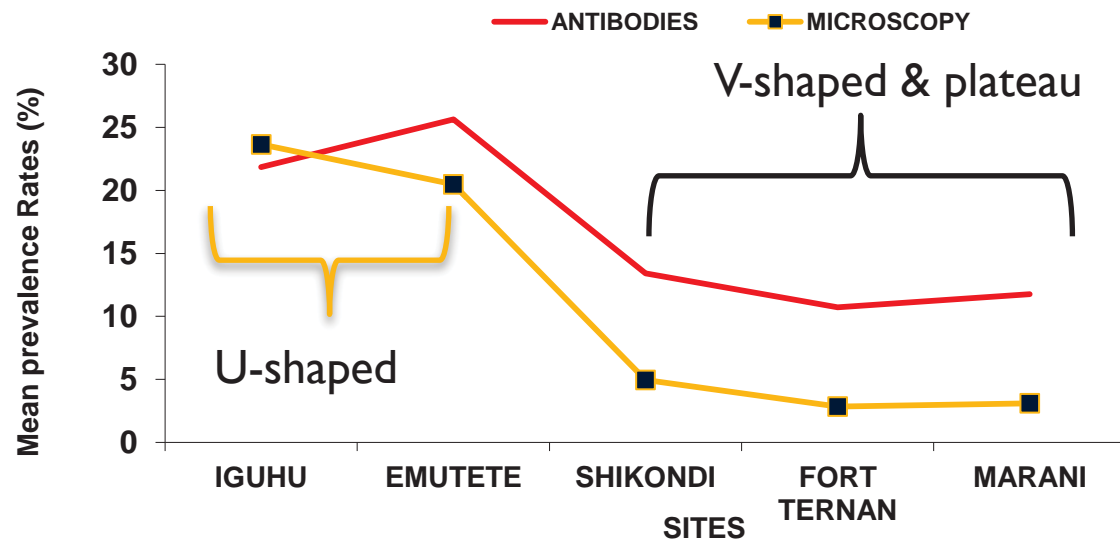
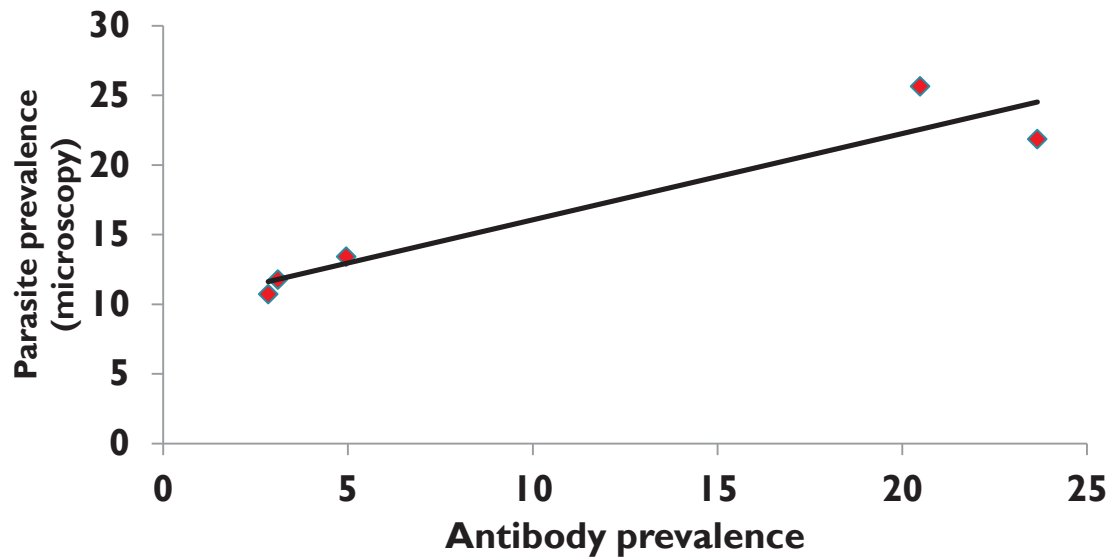


Few episodes of malaria in “V” shaped valley ecosystem

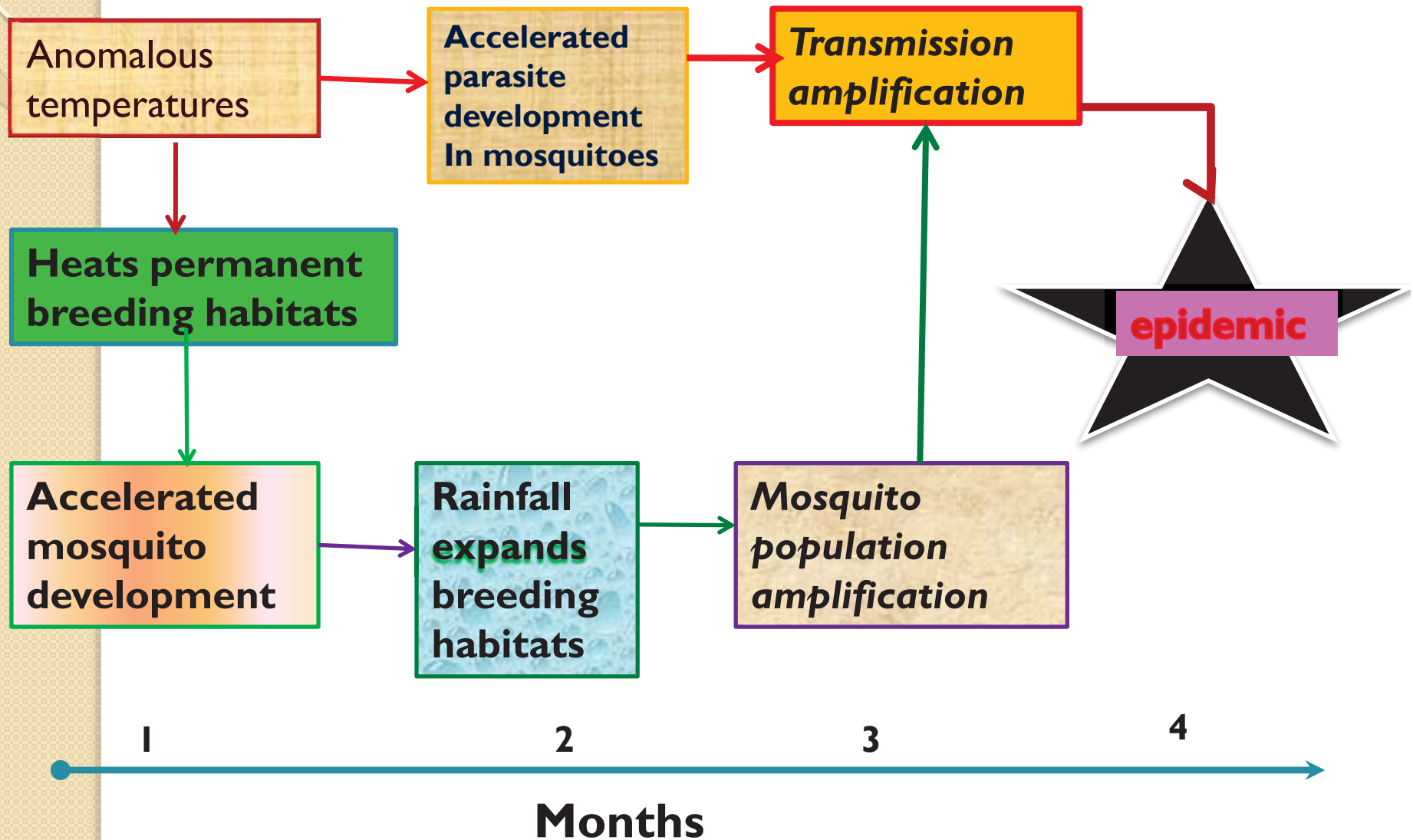
Episode/year



Low malaria prevalence low immunity: high epidemic risk



Evolution of malaria epidemics



Models using climate data

Additive model

$$ER = \frac{T^i + R^{i+2}}{T^m + R^m} \times 100$$

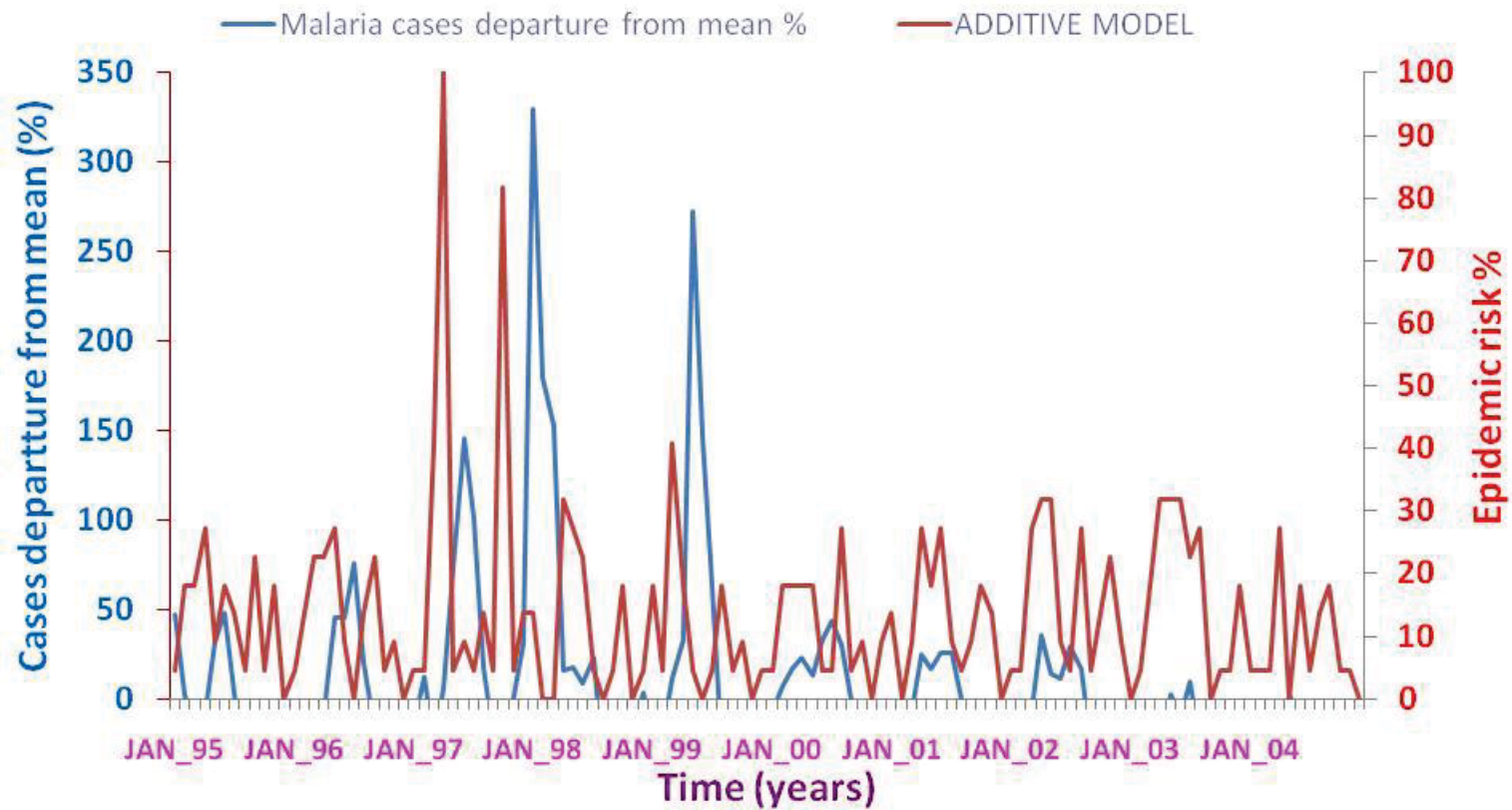
Multiplicative model

$$ER = \frac{T^i \times R^{i+2}}{T^m \times R^m} \times 100$$

18+C model

$$ER = \frac{18 - T^i \times R^i}{T^m \times R^m} \times 100$$

Kakamega: Malaria epidemic early prediction model



Evolution of an epidemic in Multiplicative model (Kericho)

Temperature
anomaly

4°C



Rainfall/
month

391 mm



**MALARIA EPIDEMIC
CASE % ABOVE MEAN**

580%

657%

795.%



Sep-97

OCT

NOV

DEC

JAN_98

FEB

MAR

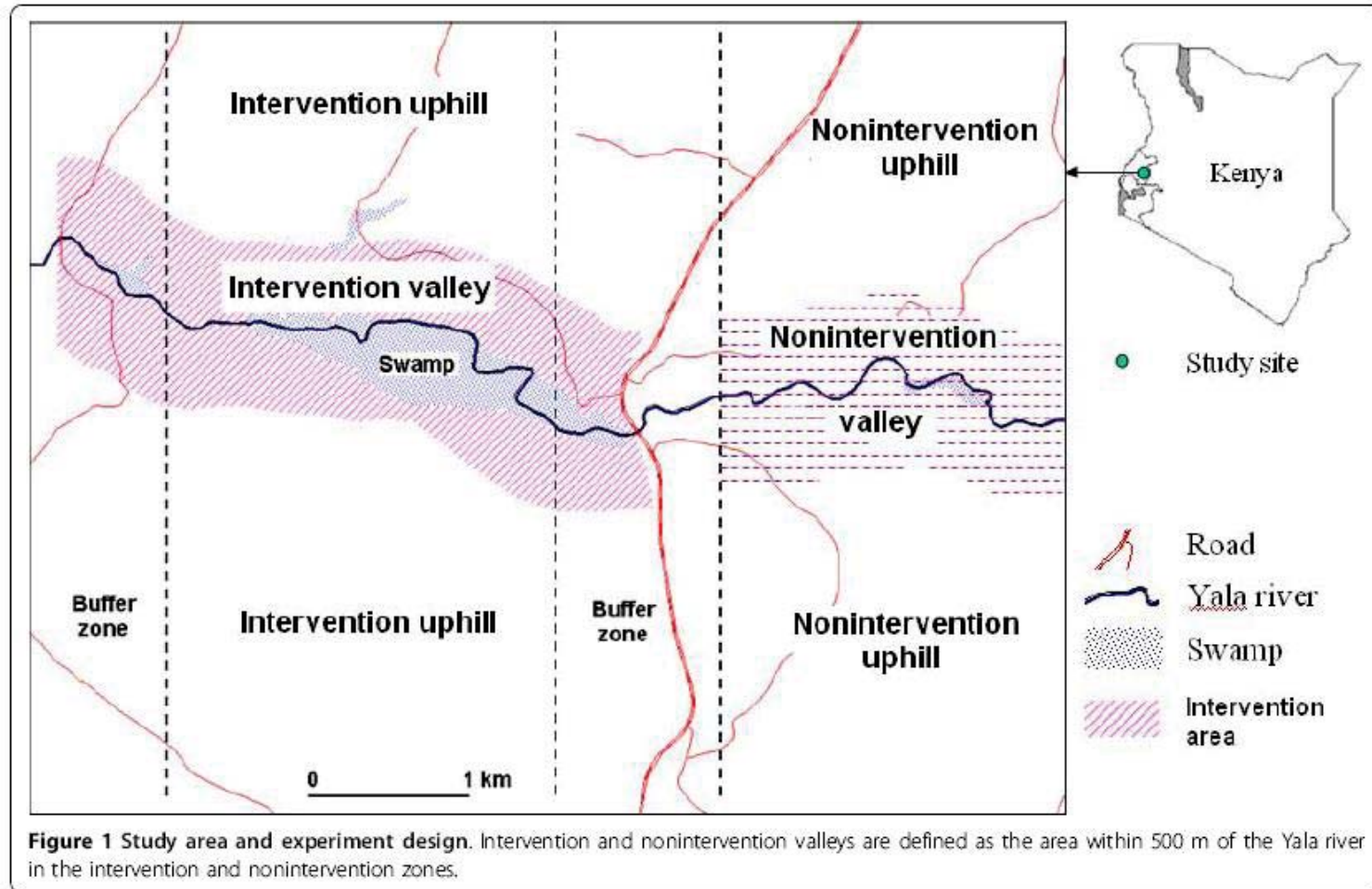
APR



Vector distribution and control

- In the highlands 98% of the malaria mosquitoes are found in houses less than 500 meters from breeding habitats at the valley bottom
- Targeting these houses for indoor resting spray is highly effective and efficient strategy for malaria epidemic control

Targeted IRS Iguhu Kakamega









Impacts of house modification

- House modification can reduce the the numbers of indoor mosquitoes by 88%
- The modification stabilizes the indoor temperatures, cooling the houses during the day and keeping them warm at night.
- The ceiling improves the house esthetics
- This is local technology that is affordable and acceptable by the local populations



Malaria epidemics

- Climate variability can increase the number of malaria cases by 100-700% and mortality by 500%
- Areas most affected by the epidemics are the highlands of Kenya, Uganda, Tanzania, Ethiopia, Rwanda, and Burundi, and Madagascar
- An early epidemic warning system was developed for early prediction and prevention of the epidemic



Adaptation to climate change

Malaria

- Develop climate based early epidemic prediction models
- Identify new areas affected by malaria transmission
- Identify epidemic hotspots
- Use universal application of insecticide treated bed nets
- Apply targeted Indoor residual Spraying (IRS)



ADAPT

THANK YOU

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