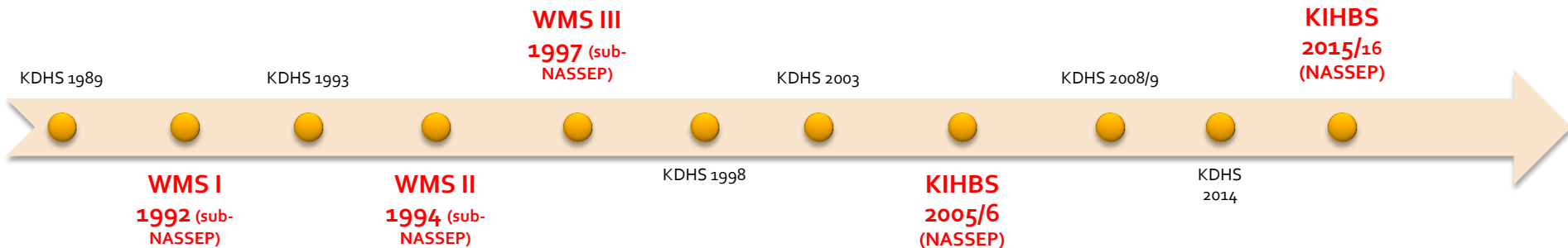


# DO THE POOR THINK THEY'RE POOR? WELFARE METRICS, MONITORING AND MAPPING IN COASTAL KENYA

Jacob Katuva: University of Oxford

**Lessons from a Decade's Research on Poverty:  
Innovation, Engagement and Impact**  
10<sup>th</sup> anniversary conference of the  
ESRC—DFID Joint Fund for Poverty Alleviation Research  
16—18 March 2016, Pretoria, South Africa

# Rationale

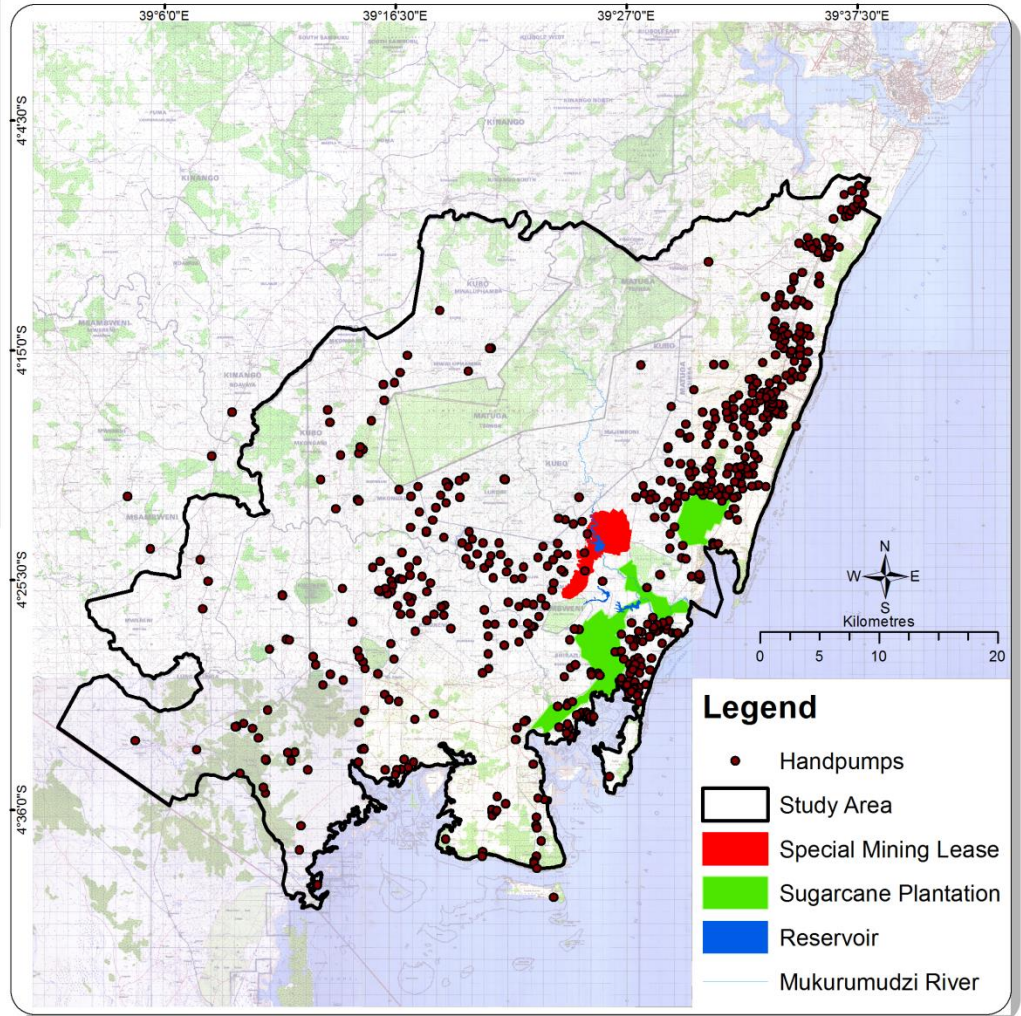
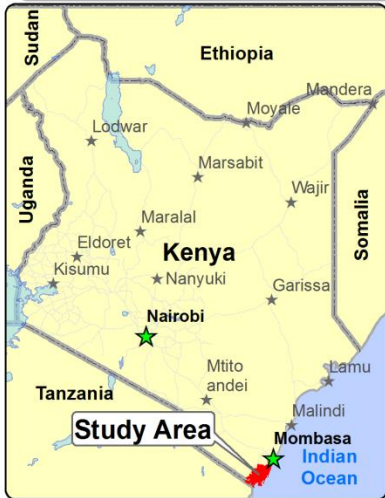


- Infrequent national surveys and a dominant income/expenditure poverty analysis has contributed to;
  - Weaknesses in tracking welfare changes
  - Uncertain links between long term (slow welfare changes) and short term (fast welfare changes) transitions
  - Difficulty in accounting for the impacts of environmental, economic, political and insecurity shocks on welfare
- Subjective welfare assessment is rarely assessed and compared in these surveys

# Research Questions

- 1) Is a 'fast' welfare index internally coherent and consistent, and comparable to a combined (slow and fast) welfare index?
- 2) How do fast and combined welfare indices coincide with subjective welfare assessments?
- 3) Do fast and combined welfare indices evaluate temporal and spatial welfare dynamics consistently?

# Study Area





# Why Kwale County?

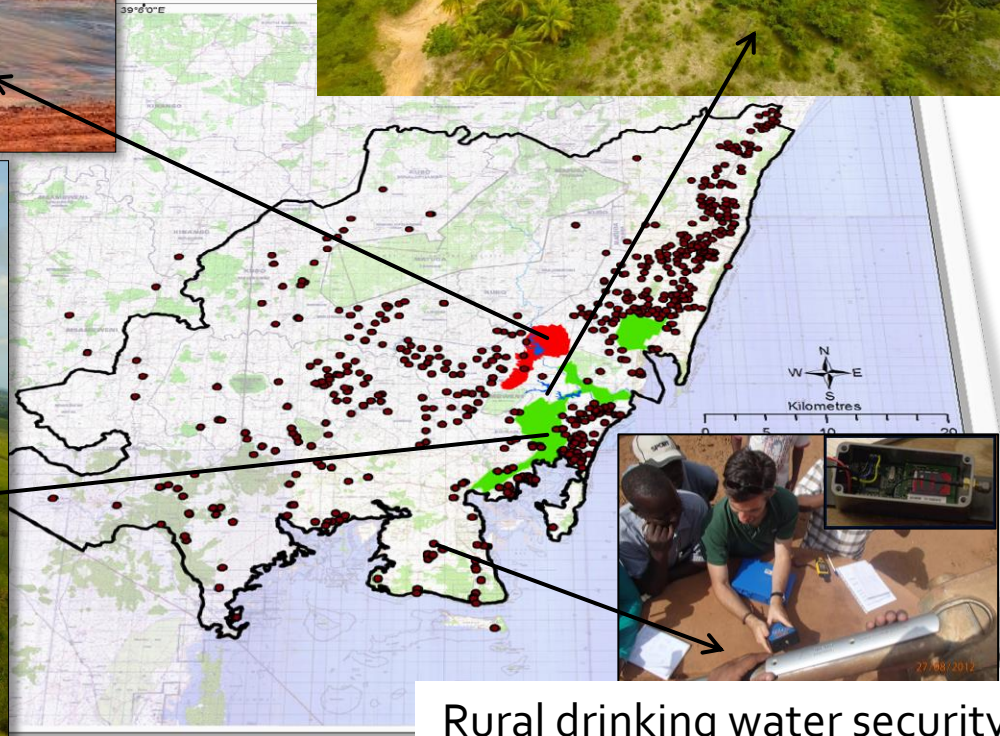
MINING OF HEAVY MINERAL SANDS



Surface Storage



SUGARCANE PLANTATIONS



Rural drinking water security





# Methodology: The survey

Training Field Teams



Communities and Water Usage



Household sampling and Survey



Household composition; Dwelling characteristics; Asset Ownership; Sanitation and health; Drinking water. [29 combined indicators vs 8 Fast Indicators]

# Provisional Results (Wealth groups by Welfare Indices)

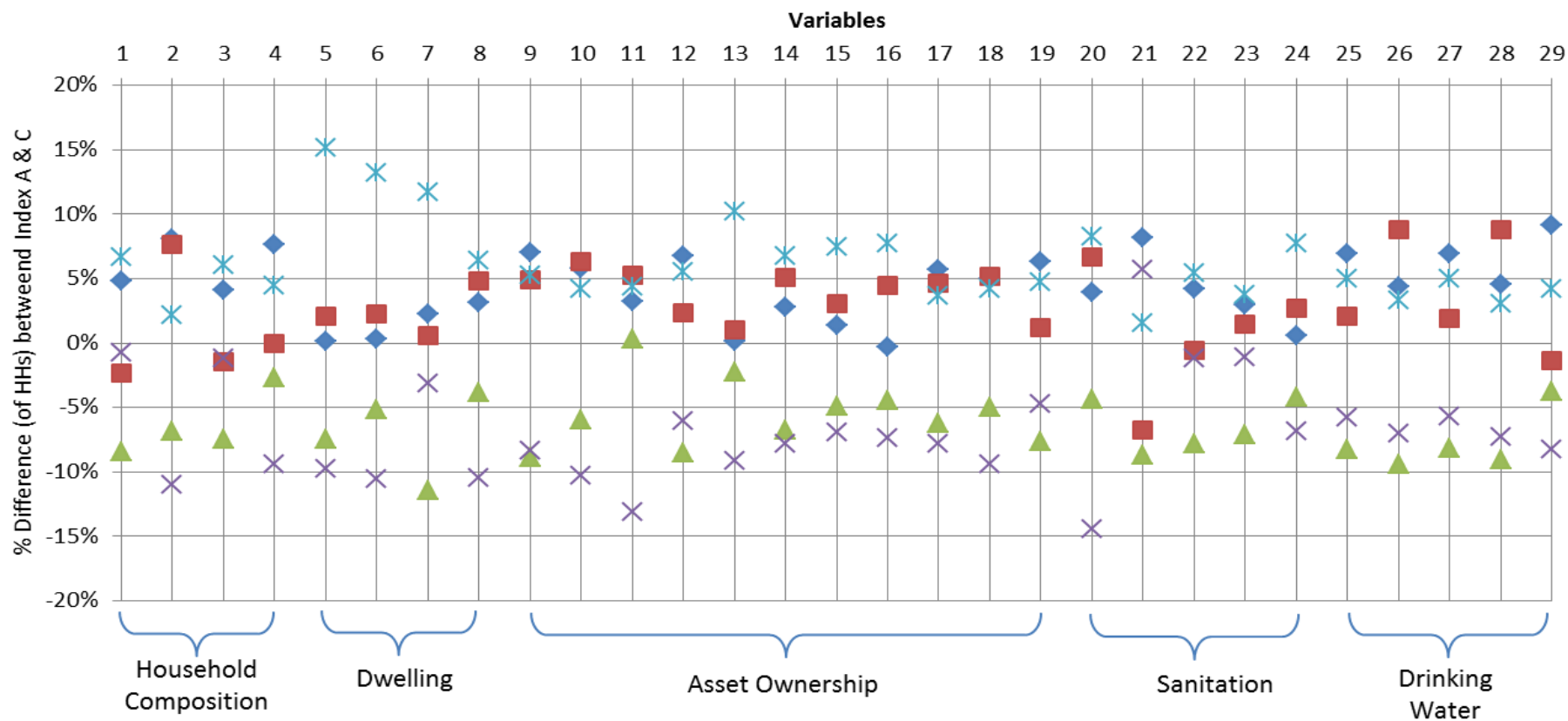
Percentage of households in different wealth groups by welfare indices, year 2014.

Wealth Group; (n = 3,229)	Combined Welfare Index A (29 variables)	Fast Welfare Index C (8 variables)
Bottom 20%	10%	16%
Second	27%	28%
Middle	30%	22%
Fourth	23%	18%
Top 20%	10%	15%
Spearman Rank Correlation Coefficient (Households)		0.908**

# Provisional Results (Consistency between Welfare Indices)

Degree of Variation Between Index A & C per Variable by Wealth Group

◆ Bottom 20%    ■ Second    ▲ Middle    ✕ Fourth    ✕ Top 20%





# Provisional Results (Stability of HHs)

		Fast Welfare Index C (8 variables)
Coastal	Households with falling welfare	21%
	<b>Status Quo</b>	<b>62%</b>
	Households with rising welfare	17%
Inland	Households with falling welfare	29%
	<b>Status Quo</b>	<b>59%</b>
	Households with rising welfare	11%
Ukunda Town	Households with falling welfare	18%
	<b>Status Quo</b>	<b>64%</b>
	Households with rising welfare	18%

Relative Welfare Changes due to Welfare Index C (by Locality) Year 2014: Reference is Combined Welfare Index A (29 variables)

# Provisional Results (Subjective Welfare vs Welfare Indices on Wealth Group)

Wealth group, n=3,229

Combined Welfare Index  
A - 29 variables

Fast Welfare Index C  
8 variables

	Bottom 20%	Second	Middle	Fourth	Top 20%	Bottom 20%	Second	Middle	Fourth	Top 20%
<b>Average</b>	2%	15%	28%	34%	21%	5%	19%	22%	25%	28%
<b>Not well off</b>	15%	36%	31%	15%	3%	24%	35%	23%	13%	6%

Overall Cross Tabulation of Subjective Welfare versus Welfare Indices

# Provisional Results (Subjective Welfare vs Welfare Indices on Welfare Change)

## Cramer's V coefficient on Positive/Negative Change

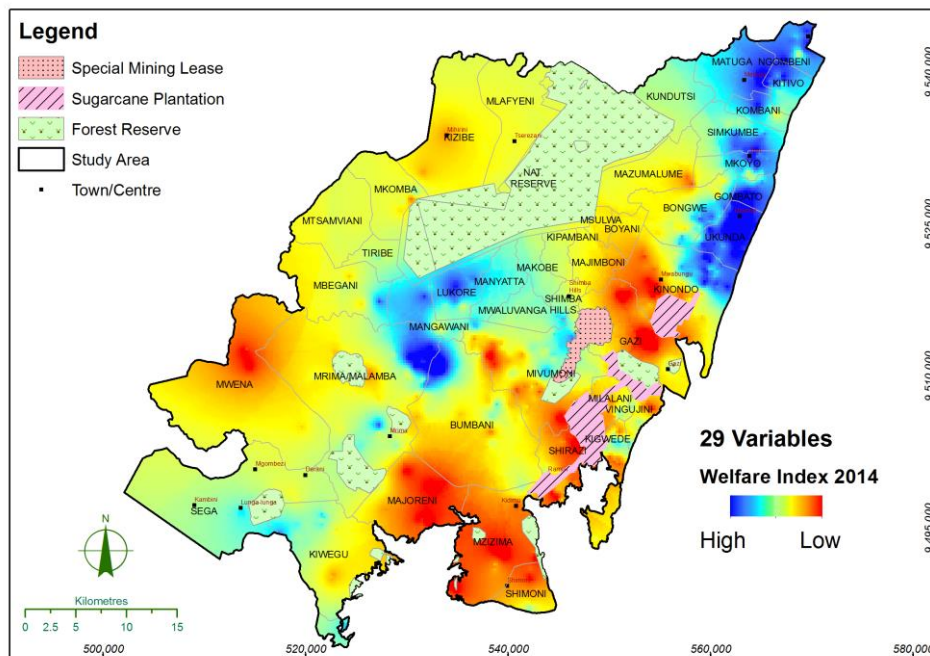
Welfare Indices	Subjective Welfare, n=1,275
Combined Welfare Index A 29 variables	0.368**
Fast Welfare Index C 8 variables	0.420**

\*\* Statistically significant at  $p < 0.001$ . Note:  $V = \phi$  for a 2 by 2 matrix

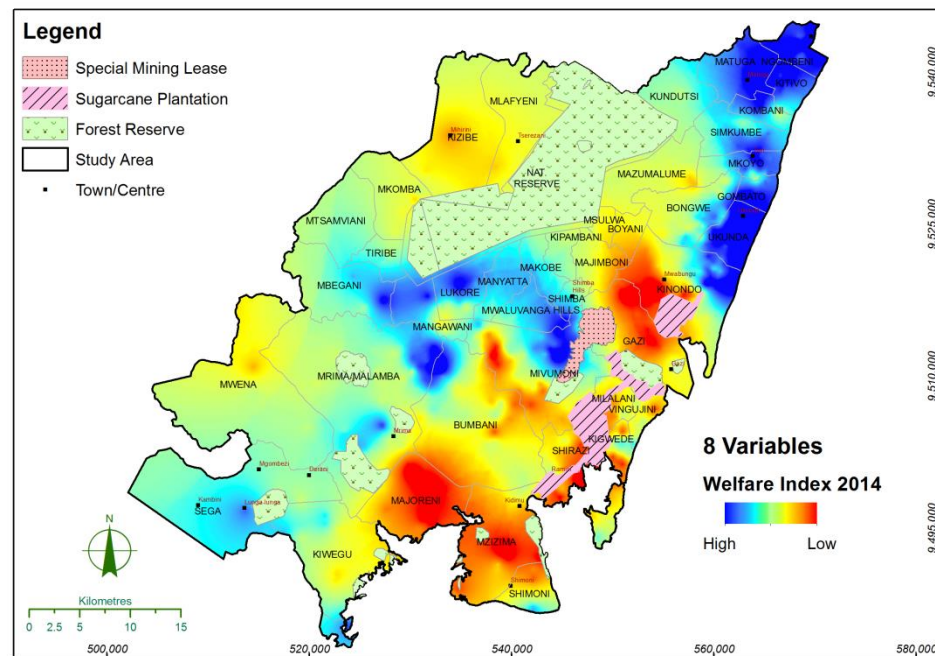


# Provisional Results (Mapping Welfare) Year 2014

## Combined Welfare Index



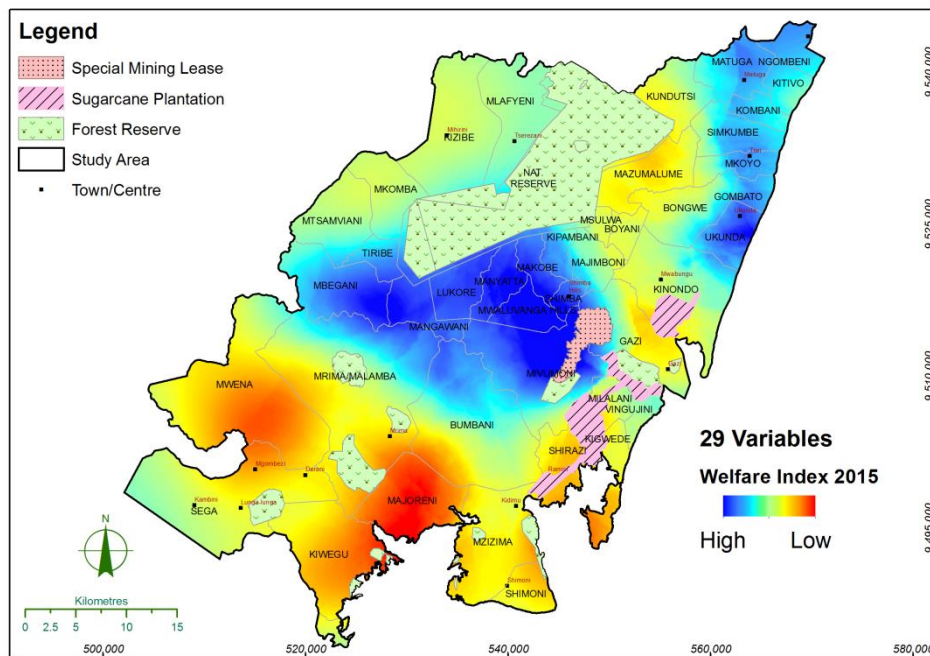
## Fast Welfare Index



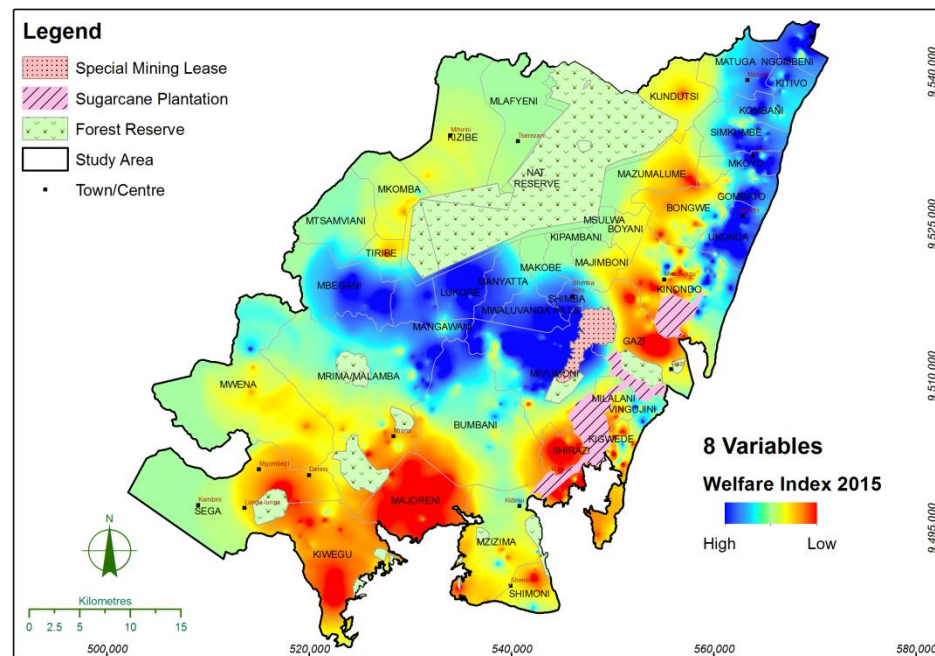
# Provisional Results (Mapping Welfare)

## Year 2015

### Combined Welfare Index



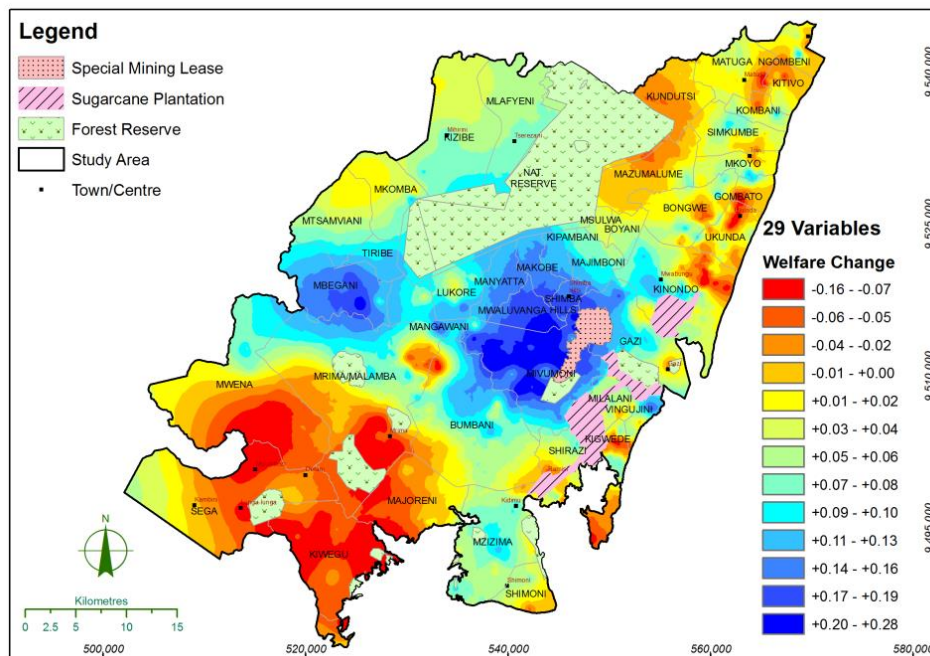
### Fast Welfare Index



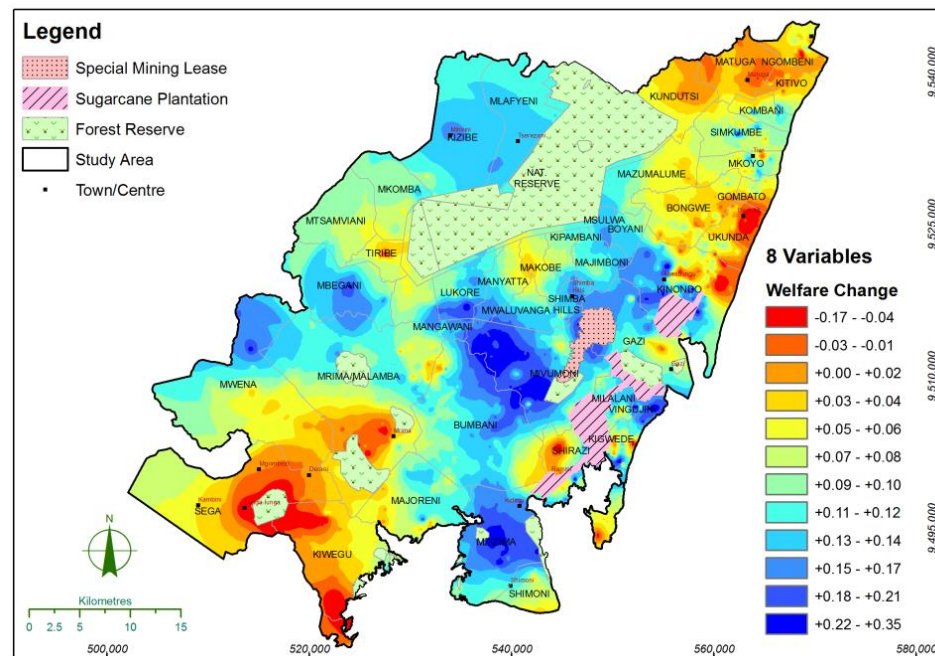
# Provisional Results

## Welfare Change over 1 Year

## Combined Welfare Index



# Fast Welfare Index





# Conclusion

- Mapping welfare transitions using different methodological approaches provides decision-makers with new and visual evidence to illustrate dynamic processes and help understand how they affect different groups over space and time.
  - For example, how do different communities benefit from proximity to a major mine, and how are benefits distributed?
  - Do terrorism events affect the poor more than the non-poor in the short and long term?
- 'Faster' time-step monitoring of welfare to support new County Governments in Kenya shape policy in political windows rather than lagged and irregular traditional survey rounds.
- Rapid and low-cost SMS surveys could be used in future to track future climate or economic shocks directly against welfare changes in terms relevant for policy action.