Investigating Seasonality & Poverty Analysis: The 2004/5 Malawi Integrated Household Survey DRAFT

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#### Background

- Many agrarian economies have seasonal patterns of income and expenditure.
- However, most studies that model determinants of poverty do no include variables that capture seasonality.
- □ In Malawi the non-seasonally adjusted model is used to predict poverty in subsequent years.
- Critical effects of seasonality on livelihoods of poor people in Malawi should affect their seasonal consumption
  - Declining food stocks & increasing purchases
  - Increasing food prices
  - Declining asset prices
  - Reliance on ganyu labour

# **Objectives**

- To investigate
  - the importance of seasonality in poverty estimates in the 2004/4 Malawi Integrated Household Survey (IHS2)
  - the effects of ignoring seasonality in investigations & estimates of poverty correlates
  - the effects of ignoring seasonality in poverty models

## **Findings: seasonality effects**



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#### Changes in Food Prices (2001 – 2008)



#### **Modelling Poverty Determinants**

- We use the Poverty and Vulnerability Analysis model but account for seasonality in the following ways for data collected over 13 months (March 2004 – March 2005):
  - Monthly dummy variables with March 2005 as base category.
  - Farming seasons dummies with March April 2004 as a base category.

Continuous categorical variable (1 to 13) based on the general trend in poverty (???)

### Findings: seasonality & model efficiency

□ Model 1 results – Monthly seasonal dummies

Monthly graph, many months significant across residence and regions. R-sq 0.47 goes to 0.5. Little change in other parameter significance.

The incremental contribution of seasonal dummies is statistically significant in all specifications.

The seasonal effects range from -0.36 (Jan05) to -0.13 (Aug04).

Absolute values of coefficients tend to increase during December, January and February (<-.30) – lean months.

# Model 1: Monthly graph



## Findings: seasonality effects & model efficiency

#### ☐ Model 2 results – Farming Seasons

Three grouped seasons (May-Aug 04, Sep-Nov 04, Dec04-Mar05) - significant at national and regions and by residence; R-sq 0.47 goes to 0.5.

□ Little change in other parameter significance.

Sep-Nov 04 and Dec04-Mar05 statistically significant and absolute value increase during the lean season.

At national level, per capita annual expenditure fall by 25% in the lean season.

# **Poverty forecast in Malawi**

Welfare Monitoring Survey (WMS)
IHS2 model of correlation between key consumption variables & poverty headcount ratio by region
Annual dry season survey uses model to estimate changing poverty incidence

	2004	2005	2006	2007
	(IHS2)	(WMS)	(WMS)	(WMS)
Poverty incidence	52%	50%	45%	40%

Is there a seasonal bias in WMS dry season estimates?

#### Conclusions

- Seasonality has significant effects on estimates of poverty incidence & depth.
- Failure to include seasonal variables in poverty models leads to model mis-specification – the parameters are therefore biased and inconsistent.
- □ Failure to allow for seasonality in poverty models can lead to unreliable estimates of poverty.
- However, we need to determine the direction of the bias in poverty estimates based on WMS.