

# How Do we reach rural households with the right agricultural and social protection interventions? Insights from the quantitative livelihood profiling study in Zambia

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Government making efforts to increase coverage of non-contributory social protection

Evidence on how polices/programmes can be targeted to different groups of households can help orient limited resources into impactful and cost effective interventions.

Reducing poverty and food insecurity among objectives

- •National Agriculture Policy
- •National Social Protection Policy
- •6th National Development Plan

Quantitative
livelihood profiling
is one tool that can
be used to inform
interventions
aimed at reducing
poverty and food
insecurity





#### About Livelihood Profiling

Livelihood profiling is the categorization of households based on their similarity with respect to a number of variables

It provides a more nuanced picture than simple classifications based on only **one or two dimensions** such as income or the size of land holdings.





Conclusions







Key Findings



- •The study uses the Rural Agricultural Livelihoods Survey (RALS) data of 2015 to generate livelihood clusters
- •Interviews with key stakeholders were done to ensure that study responded to information needs

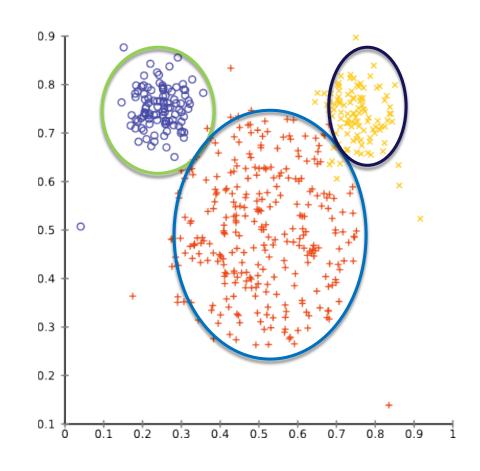
Chapoto and Zulu-Mbata 2015



#### Data and methods.....

This study uses Principal Component Analysis (PCA) and cluster analysis on:

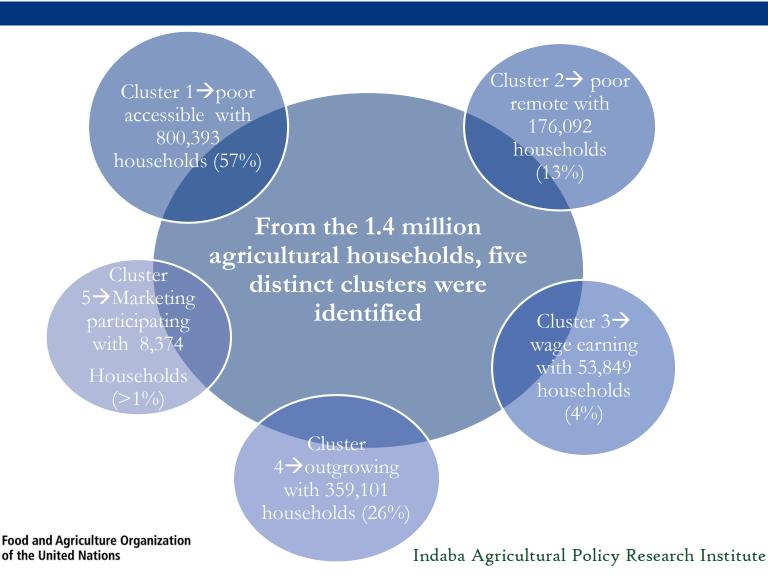
- (i) selected livelihood asset variables adopted from the Sustainable Livelihood Framework;
- (ii) variables used to target households for different agricultural and social protection programmes in rural Zambia that were collected in the RALS 2015.







#### Key Findings



#### Key cluster characteristics

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Characteristic	Household type				
	Cluster 1	Cluster 2	Cluster 3	Cluster 4	Cluster 5
Number of Households	800,393	176,092	53,849	359,101	8,374
Percent in low income group	60.7	58.5	3.1	30.1	0.5
Percent female headed	30.8	23.3	14.2	11.5	19.8
Average age of household head	47.4	43.6	44.9	46.6	53.4
Average household dependency ratio	39.4	40.4	28.8	37.5	31.9
Maximum years of education	7.3	7.1	14	8.7	10.9
Percent school age children attending school	62	59	86.5	66.6	78.7
Percent with group membership	40.5	41.8	63.4	77.1	75.8
Percent migrant households	8.7	7.8	42.6	8.6	12.6
Average cultivated land (ha)	1.5	2.3	1.9	3.3	7.9
Percent with good quality housing	10.1	6.4	89.1	30.1	73.2
Average disposable income	4,945	6,483	62,341	14,265	58,652
Percent employed in public sector	0.3	0.4	76.3	0.7	1.9
Percent obtaining credit	4.3	16.2	11	41.8	13.7
HCI	20.1	34.4	37.3	53.4	65.7
Kilometres to the nearest Boma	34.1	100.3	23.5	37.1	36.9
Hours to the nearest urban centre	14	22	11	11	9
% below the poverty line (\$1.25/day)	86.1	85.2	10.5	69.7	18.9

## Participation in the different Agricultural and social protection programmes by cluster

	Percentage by household type						
Type of participation	Poor Accessible	Poor Remote	Wage Earning	Outgrowing	Market Participating		
%Participation in SCT	2.50	0.82	0.45	0.68	0.00		
%Participation in FISP	27.78	31.33	51.89	58.72	51.33		
%Participation in FSP	0.54	0.37	0.12	0.62	1.10		

Source: CSO/MAL/IAPR I2015





#### Household source of fertilizer by cluster

	Total amount acquired by household type							
	Poor		Wage		Market			
Source	Accessible	Poor Remote	Earning	Outgrowing	Participating			
Subsidies (kg'000)	47,779	13,562	7,763	61,943	1,634			
Cash purchases (kg'000)	34,237	12,489	15,549	95,682	9,868			
Loans (kg'000)	1,048	1,611	1,268	14,005	150			
Other sources (kg'000)	1,549	347	542	2,627	73			
Total	84,613	28,009	25,122	174,258	11,725			

Source: CSO/MAL/IAPRI 2015





#### Household source of fertilizer by cluster...

	Average amount acquired by household type						
Source	Poor Accessible	Poor Remote	Wage Earning	Outgrowing	Market Participating		
Subsidies (kg)	60	77	144	172	195		
Cash purchases (kg)	43	71	289	266	1,179		
Loans (kg)	1	9	24	39	18		
Other sources (kg)	2	2	10	7	9		
Total	106	159	467	485	1,400		

Source: CSO/MAL/IAPRI 2015





#### Poverty incidence, gap and severity

	Household type					
Poverty Measure	Poor Accessible	Poor Remote	Wage Earning	Outgrowing	Market Participating	
Percentage below income poverty line*	86	85	10	70	19	
Percent poor (incidence)	_					
Poverty gap index	57	57	5	37	6	
Poverty severity index	43	43	3	23	2	

Note: \*=1; 2005 PPP exchange rate. Poverty line=\$1.25/day

Source: CSO/MAL/IAPRI 2015





### Household food security outcomes by livelihood cluster

	Average value by household type					
Food security measure	Poor Accessible	Poor Remote	Wage Earning	Outgrowing	Market Participating	
Percent food secure households	44.66	50.14	87.51	71.59	81.45	
Months without adequate food	2.14	1.93	0.44	0.86	0.58	
Household Dietary Diversity Score	5.19	4.88	8.05	6.51	7.87	







#### Conclusions

Our findings show that substantial amount of agricultural subsidies and social protection resources end up benefiting households that should not benefit from them

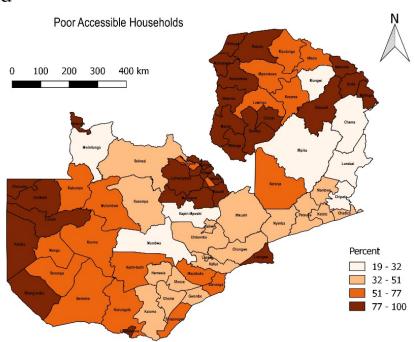
- This suggests deficiencies in current targeting mechanisms
- Our study shows that livelihood profiling could add value to the currently used targeting mechanisms as it gives a nuanced picture of household vulnerability compared to the other targeting mechanisms that are based on one or two dimensions





#### Mapping of household clusters

For example, cluster 1, the 'poor accessible households' is mostly concentrated in Luapula, Copperbelt and Western provinces and in the northeastern districts of Muchinga province.











### Thank you

For more information visit our websites at:

http://www.iapri.org.zm/

http://www.fao.org/home/en/





#### Bonus Slides

This study uses Principal Component Analysis (PCA) and cluster analysis to generate the clusters and followed the method in Ansoms and Mckay (2010) paper.

Due to multicollinearity and the advantage of using fewer variables for cluster analysis, PCA is used which generates principle component scores which are used as input variables in cluster analysis. Hierarchical cluster analysis was then used. This is used to determine the number of livelihood groups/clusters to use in Non-hierarchical cluster analysis.

Hierarchical cluster analysis has a weakness of possible misclassification of households (Jenson et al., 2006). To this effect, non-hierarchical cluster analysis was used to correct for this using k-means non-hierarchical cluster analysis.



#### Potential uses of livelihood profiling

