# Assessment of cropping systems and net economic returns in three agroecosystems of Southern Myanmar

Phyu Thaw Tun<sup>1</sup>, Zikrullah Safi<sup>1</sup>, Andreas Buerkert<sup>1</sup>

<sup>1</sup> Organic Plant Production and Agroecosystems Research in the Tropics and Subtropics, University of Kassel, Germany

### Introduction Materials and methods Semi-structured interviews with 301 households (15% of $\triangleright$ Loss of biodiverse natural forests in Kyunsu Township of the households in the Plantation Zone, 35% in the Southern Myanmar, as a consequence of agricultural Lowland Zone, and 50% in the Sea Zone) land expansion Hierarchical clustering of principal components (HCPC, $\geq$ Decline of ecosystem services

## Objective

- To assess cropping systems, their land productivity and production constraints in Kyunsu Township
- To gather information to identify potential interventions that can address production constraints, thereby increase the net economic returns from current croplands
- Housson et al., 2010) to classify cropping households with homogeneous characteristics in each zone
- Principal component analysis → retained principal components (PCs) which showed eigenvalues > 1 and explained variances >  $60\% \rightarrow$  hierarchical clustering
- $\triangleright$ Kruskal-Wallis test followed by Wilcoxon rank sum test to evaluate significant differences and calculate pairwise comparisons of variables between types of cropping systems in each zone

## **Results**

#### Hierarchical clustering on principal components Principal component analysis 2 retained PCs in the plantation zone 3 retained PCs in the lowland zone 4 retained PCs in the sea zone Hierarchical clustering Type 1: Input intensive, low productive system Type 1: Input limited, low productive system Type 1: Input limited, low productive system Type 2: Input limited, low productive system Type 2: Input intensive, high productive system Type 2: Input limited, high productive system Type 3: Input intensive, high productive system Type 3: Input limited, high productive system Type 3: Input intensive, low productive system Figure 1. Types of cropping systems resulting from hierarchical clustering on principal components (HCPC) in the three agroecological zones of Kyunsu Township. freferti Plantation Zone Lowland Zone Sea Zone variance) S (ariance) variance) explained \ (19.8% of explained b 0 explained ncrop Туре Туре Туре proplan • 1 • 2 • 3 • ▲ 2 ■ 3 (23.4% of e. N (22.6% of e on PC2 ( 03 02 022

-3 0 3 PC1 (44.4% of explained variance) Figure 2. Distributions of the corresponding types of cropping systems in the three agroecological zones of Kyunsu Township by correlation of the two most important principal components (PC1 and PC2) with key variables: culland=area of cultivated land, ncrop=number of cultivated crops, proplan=proportion of cultivated land area with plantation crops, prorice=proportion of cultivated land area with rice, freferti=frequency of fertilizer application, ininput=intermediate input, gva=gross value added, and proincrop=proportion of income from cropping activity.

## **Performance of cropping systems** > Constraints encountered

PC1 (38.9% of explained variance)



Figure 3. Land productivity of the types of cropping systems within the three agroecological zones of Kyunsu Township.



Figure 4. Proportion of major production constraints of the types of cropping systems within the three agroecological zones of Kyunsu Township.

## **Conclusion and recommendation**

-3 0 3 PC1 (26.9% of explained variance)

- Diversified cropping systems combined with appropriate input utilization exhibit the potential to enhance land productivity in terms of gross value added.
- Farmers' apparent production constraints in each cropping system should be considered in improving land productivity on current cultivated lands. This may help to prevent further encroachment of natural forests in search for enhanced crop production.

DAAD

Reference Husson, F., Josse, J., & Pages, J. (2010). Principal component methods-hierarchical clustering-partitional clustering: why would we need to choose for visualizing data. Applied Mathematics Department 17.