



## Evaluating the performance of different commercial and pre-commercial maize varieties under low nitrogen conditions using affordable phenotyping tools

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# Introduction



- **Maize**
  - Important in Africa (FAO, 2017)
- **Low Nitrogen**
  - Low N and low money in Africa (Cairns et al., 2013)
- **Breeding Strategy**
  - Breeding genetic gains specific to low N





We evaluated the selection of maize varieties using a set of remote sensing indices derived from RGB images acquired from a UAV (Unmanned Aerial Vehicle) and at the ground level compared with the performance of the field-based NDVI and SPAD sensors, and then we tested their capacity for yield estimation both alone and in combination with standard agronomical variables, such as ASI (Anthesis Silking Data), AD (Anthesis Data), and Plant Height (PH).



# Materials and Methods

## Case Study

Zimbabwe

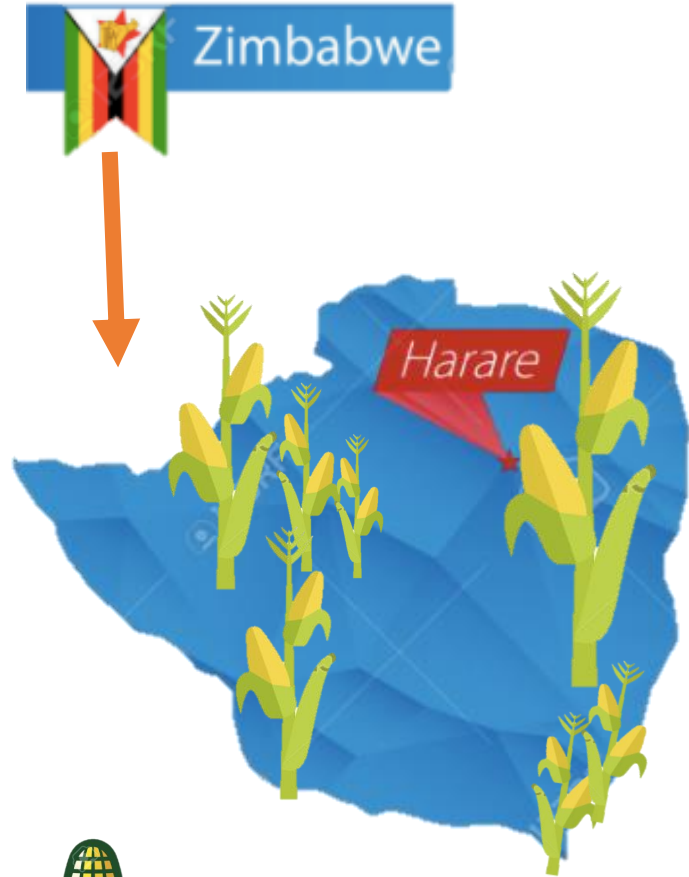


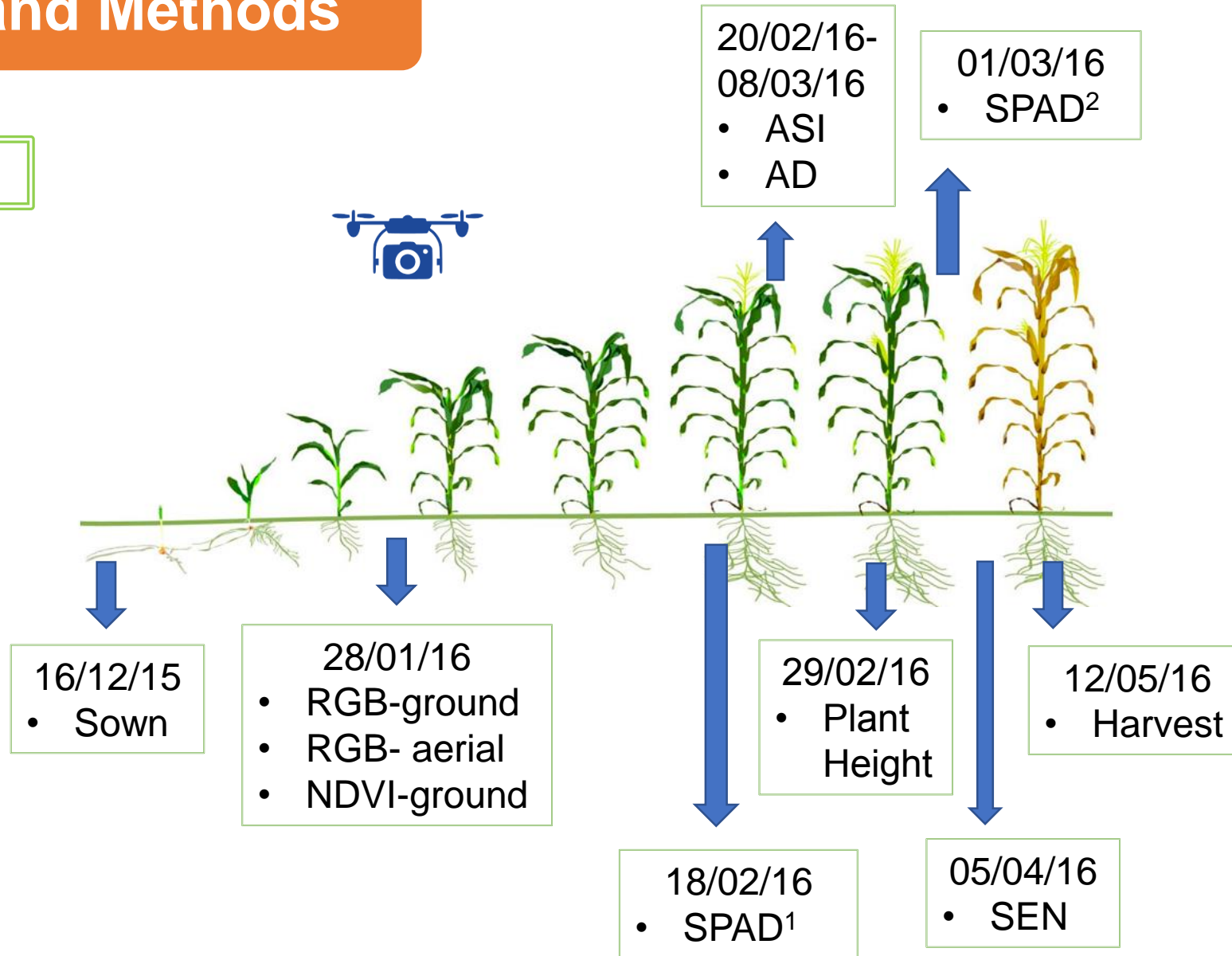
Figure 1. RGB aerial orthomosaic of the plot images under managed low nitrogen.

- December 2015- May 2016
- 49 pre-commercial varieties of Centro Internacional de Mejoramiento de Maiz Y Trigo (CIMMYT).
- 15 commercial varieties of private company.
- In Low managed nitrogen
- 192 plots (5.25m<sup>2</sup>) with 3 replica per varieties

# Materials and Methods



## Plot Sampling



# Materials and Methods



Remote Sensing

RGB images at ground level

RGB images at aerial level



Taken with an Olympus OM-D, holding the camera about 80 cm.

Taken with an UAV Mikroopters OktoXL, flying under remote control at about 50 m. The digital camera used for aerial imaging was a Lumix GX7, Panasonic.



# Materials and Methods



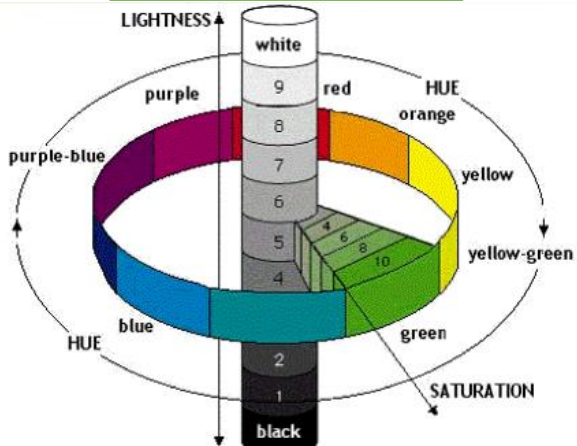
Image processing

FIJI

Maize Scanner

Canopy Macros

HIS color space

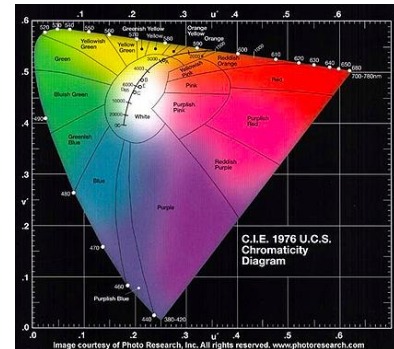
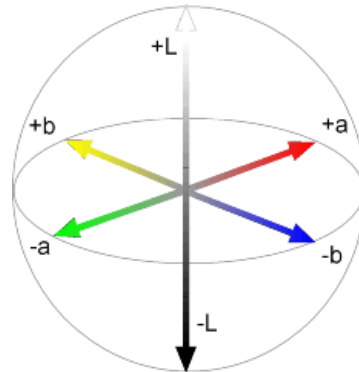


Breedpix

(Casadesús et al., 2007)

Cie-Luv

Cie-Lab



- $u^*$
- $v^*$

- $a^*$
- $b^*$

$$\text{Triangular Greenness Index (TGI)} = -0.5 [190(R670 - R550) - 120(R670 - R480)]$$

(Hunt et al., 2012)

$$\text{Normalized Green - Red Difference Index (NGRDI)} = \frac{(\text{Green DN} - \text{Red DN})}{(\text{Green DN} + \text{Red DN})}$$

(Hunt et al., 2005)

- Green Area (GA) (pixels with  $60^\circ < \text{Hue} < 180^\circ$ )
- Greener Green Area (GGA) (pixels with  $80^\circ < \text{Hue} < 180^\circ$ )
- Crop senescence index (CSI) 
$$100 \times \frac{(GA - GGA)}{GA}$$

(Zaman-Allah et al., 2015)





# Materials and Methods



Field Sensors

Relative Chlorophyll Content measured with Minolta SPAD-502 chlorophyll meter

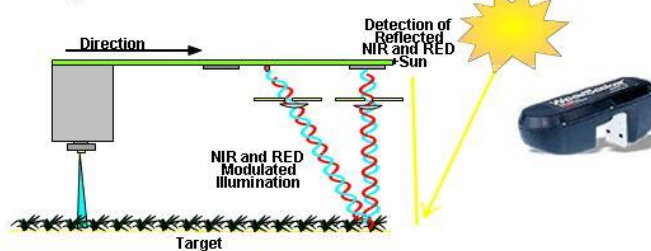
Normalized Difference Vegetation Index (NDVI) measured with GreenSeeker



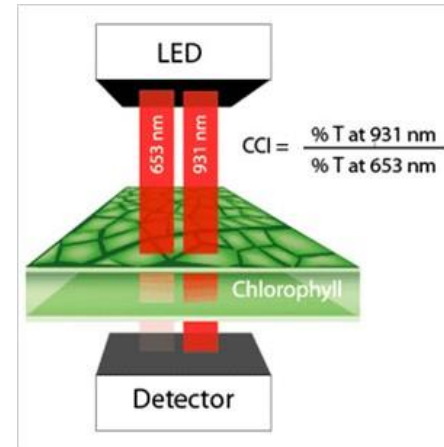
24 - 48" (60 - 120 cm)



GreenSeeker™ Sensor  
Light Detection and Filtering



$$NDVI = (R840 - R670) / (R840 + R670) \text{ (Rouse et al., 1973)}$$





## The effect of optimal condition and low managed nitrogen on grain yield.

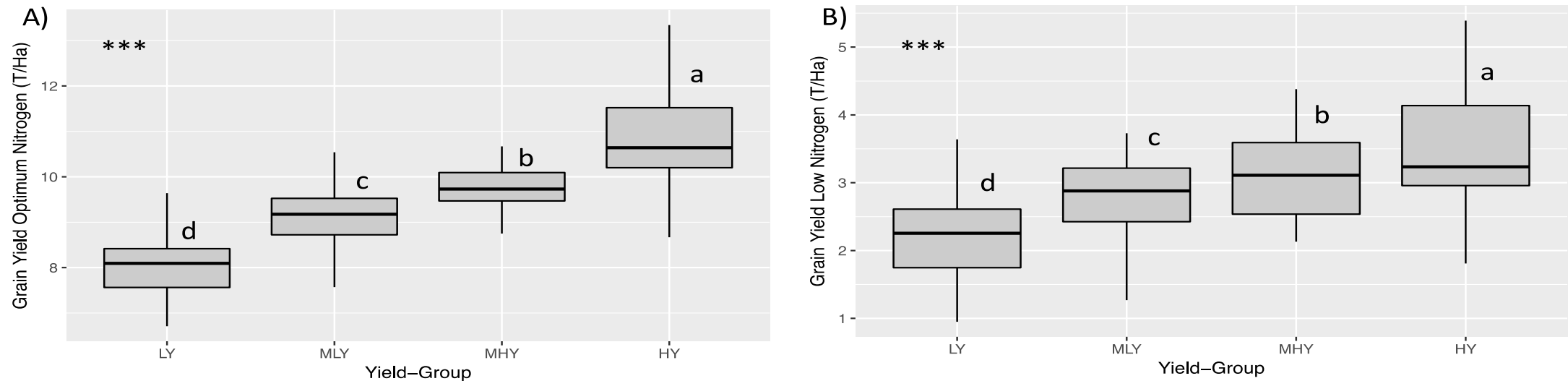


Figure 2. LY (Low Yield), MLY (Medium Yow yield), MHY (Medium High Yield) and HY (High Yield) maize variety in two different conditions: (A) Optimum Nitrogen (OP) and (B) Low Nitrogen (LOW). Each value is the mean  $\pm$  SD for each genotype (n= 48 per quartile with 16 different variety). Bars with the different letters are significantly at  $P < 0.001$ .



# Results and Discussion



## Performance of remote sensing indices and field sensors assessing grain yield

Table. 3 Grain yield correlations with all proximal remote sensing variables from the RGB images taken from the UAV aerial platform, RGB images from the ground, and SPAD and NDVI field sensors. These indices are defined in the Introduction and Materials and Methods. Levels of significance: \*,  $P < 0.05$ ; \*\*\*,  $P < 0.001$ .

RGB indices/ aerial	GY						Additional Field Sensors	R	P
	R	P	RGB indices/ ground	R	P				
GGA	0.1978	***	GGA	0.2339	***	SPAD <sup>1</sup> (18/02/16)	0.2936	***	
GA	0.1659	***	GA	0.2175	***	SPAD <sup>2</sup> (01/03/16)	0.2564	***	
Hue	0.1449	***	Hue	0.2351	***	NDVI	0.1404	***	
Intensity	0.0932	***	Intensity	0.0090					
Saturation	0.1819	***	Saturation	0.0515	*				
Lightness	0.0848	***	Lightness	0.0208	*				
a*	0.1275	***	a*	0.1467	***				
b*	0.1573	***	b*	0.0080					
u*	0.1470	***	u*	0.2021	***				
v*	0.0884	***	v*	0.0002					
CSI	0.1830	***	CSI	0.1031	***				
TGI	0.0527	*	TGI	0.0019					
NGRDI	0.1645	***	NGRDI	0.0007					





## Multivariate Yield Estimations

Measurement Combinations	R <sup>2</sup>	P
RGB ground + Field sensors	0.403	***
RGB aerial + Field sensors	0.384	***
Agronomic + RGB ground	0.559	***
Agronomic + RGB aerial	0.560	***

Table 5. Multilinear regression (stepwise) of Grain Yield (GY) as the dependent variable the different categories of remote sensing traits RGB ground and aerial level (these indices are defined in the Introduction), agronomic data like ASI (Anthesis Silking Interval), AD (Anthesis Data), MOI (Moisture), SEN (Canopy Senescence) and PH (Plant Height) NDVI (Normalized Different Vegetation Index) and SPAD (relative chlorophyll content). R<sup>2</sup>, determination coefficient; Level of significance: \*\*\*; P<0.001.



# Conclusions



- Maize hybrid technology may show promise for improving much-needed GY in low N environments and the current range of variability in performance suggests the possibility of potential for further improvements.
- For HTPP, RGB sensors can be considered as functional technology from the ground or a UAV, but also, similar to SPAD, NDVI or any other agronomic or general plant physiological measurement
- Measurements must be carefully planned for an adequate growth stage in order to optimize their benefits to plant breeding. Possible gains with new technologies with regards to equipment and time costs, especially in larger breeding platforms.
- We need to take advantage of known effects of low N on physiological processes to focus our efforts to bring HTPP to low N breeding.



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THANK YOU



# References



- Casadesús, J., Kaya, Y., Bort, J., Nachit, M.M., Araus, J.L., Amor, S., Ferrazzano, G., Maalouf, F., Maccaferri, M., Martos, V., Ouabbou, H., Villegas, D., (2007). Using vegetation indices derived from conventional digital cameras as selection criteria for wheat breeding in water-limited environments. *Ann. Appl. Biol.* 150, 227–236. doi:10.1111/j.1744-7348.2007.00116.x
- Hunt, E. R., Cavigelli, M., Daughtry, C. S. T., McMurtrey, J. E., & Walthall, C. L. (2005). Evaluation of digital photography from model aircraft for remote sensing of crop biomass and nitrogen status. *Precision Agriculture*, 6(4), 359–378. <https://doi.org/10.1007/s11119-005-2324-5>
- Hunt, E. R., Doraiswamy, P. C., McMurtrey, J. E., Daughtry, C. S. T., Perry, E. M., & Akhmedov, B. (2012). A visible band index for remote sensing leaf chlorophyll content at the Canopy scale. *International Journal of Applied Earth Observation and Geoinformation*, 21(1), 103–112. <https://doi.org/10.1016/j.jag.2012.07.020>
- Zaman-Allah, M., Vergara, O., Araus, J. L., Tarekegne, A., Magorokosho, C., Zarco-Tejada, P. J., Cairns, J. (2015). Unmanned aerial platform-based multi-spectral imaging for field phenotyping of maize. *Plant Methods*, 11(1), 35. <https://doi.org/10.1186/s13007-015-0078-2>
- Muruli BI, Paulsen GM. 1981. Improvement of nitrogen use efficiency and its relationship to other traits in maize. *Maydica* 26, 63–73
- Cairns, J.E., Crossa, J., Zaidi, P.H., Grudloyma, P., Sanchez, C., Araus, J.L., Thaitad, S., Makumbi, D., Magorokosho, C., Bänziger, M., Menkir, A., Hearne, S., Atlin, G.N., (2013). Identification of Drought, Heat, and Combined Drought and Heat Tolerant Donors in Maize 1335– 1346. doi:10.2135/cropsci2012.09.0545
- FAO (2017). Food and Agriculture Organization of the United Nations; Statistic Division. Available online at: <http://faostat.fao.org/>.

