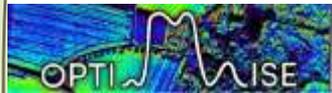




UNIVERSITAT DE  
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# UAV Phenotyping and Proximal Sensing for Maize Assessments in Breeding Programs

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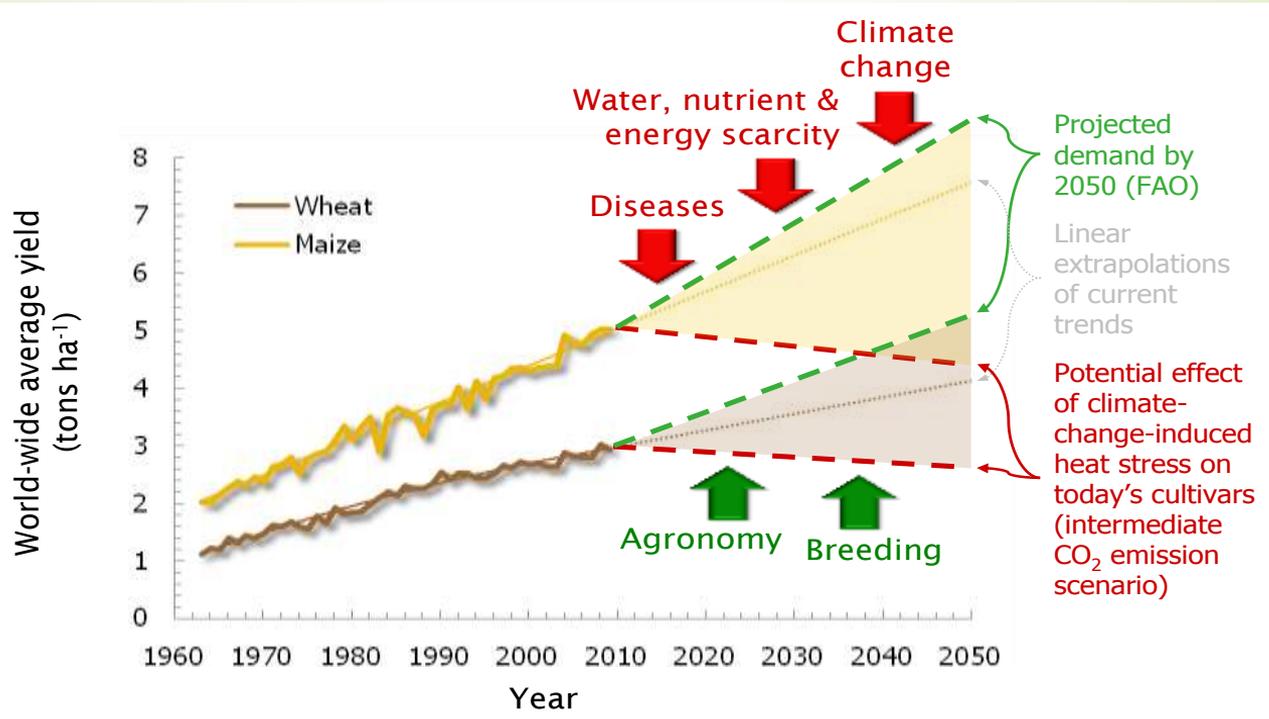
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(2) International Maize and Wheat Improvement Center, CIMMYT Southern Africa Regional Office, Harare, Zimbabwe

OPTIMISE Final Conference, 21– 23 February 2018, Sofia, Bulgaria

# Global Challenge of Phenotyping

How to enhance annual yield gains of maize from 1.6% to 2.4% on essentially the same land area, and as climates change and the costs of fertilizer, water, and labor increase?



# RGB, Red, Green and Blue broad-band visible light reflectance

Meaningful data can be had from these normal digital images, especially considering 1. excellent color calibration during production covering photosynthetically active radiation (PAR) and 2. very high spatial resolution (mm).

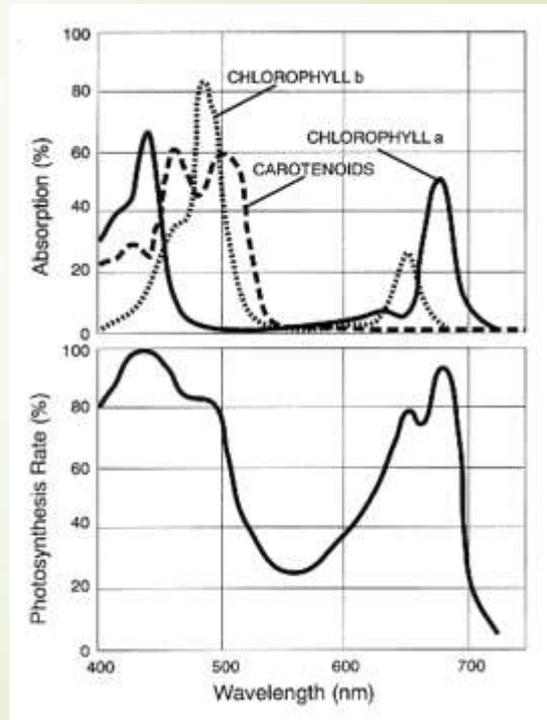


Image processing

FIJI

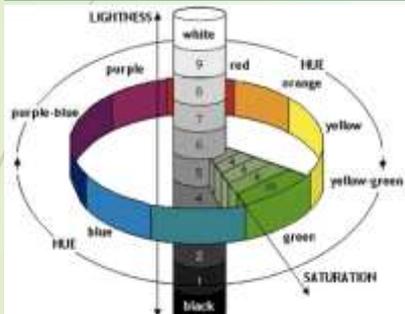
CIMMYT Maize Scanner

Breedpix

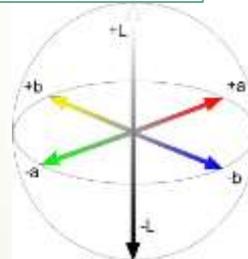
(Casadesús et al., 2007)

Canopy Macros

HIS color space

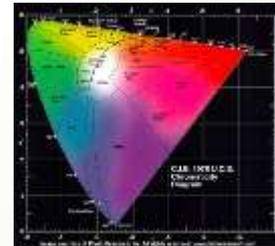


Cie-Lab



- a\*
- b\*

Cie-Luv



- U\*
- V\*

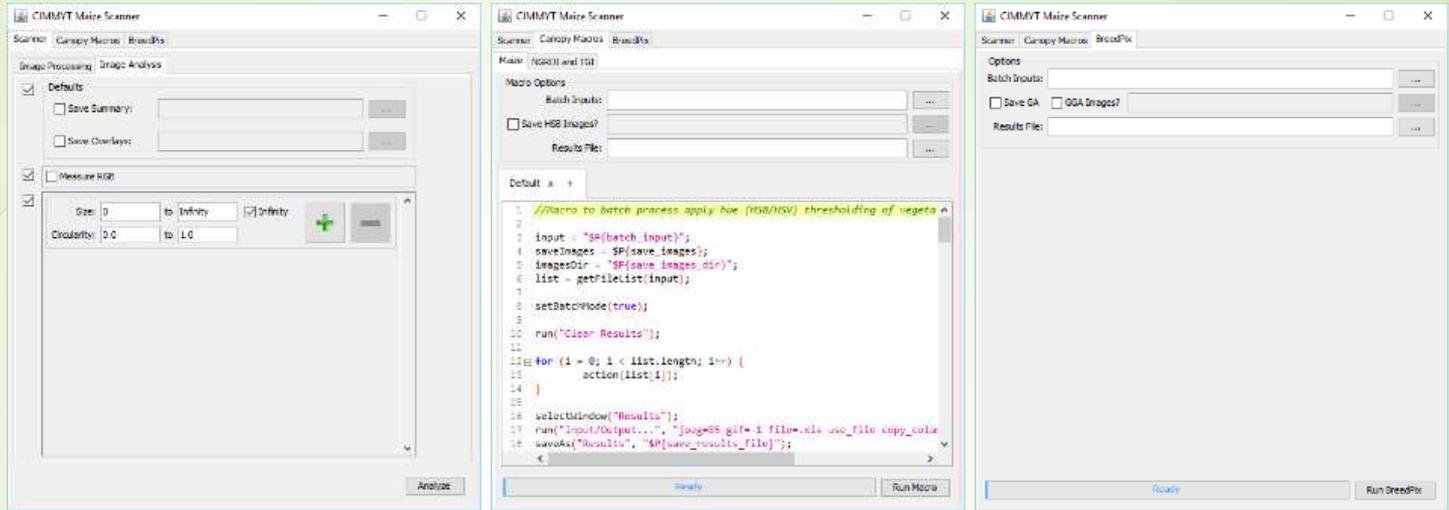
$$\text{Triangular Greenness Index (TGI)} = -0.5 [190(R670 - R550) - 120(R670 - R480)]$$

(Hunt et al., 2012)

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

(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)



## CIMMYT Maize Scanner

RGB field-based phenotyping

(<http://github.com/george-haddad/CIMMYT>)

Calculates a number of RGB based indexes for estimating disease impacts, crop vigor, LAI, biomass at the leaf and canopy scale, including Breedpix (GA and GGA), Triangle Greenness Index (TGI), and Normalized Green Red Difference Index (NGRDI)

# University of Barcelona HTPP



Multispectral Tetracam MCA11+ILS



TEAX ThermalCapture  
FLIR Tau 640 Camera



Lumix GX7 16 MP RGB

Also AIRINOV MultiSpec 4 NDVI/PRI, SonyQX1 and Tetracam MCA-6 custom

# Improved processing techniques: 3D models and image segmentation

Computer intensive processes should be batched remotely.

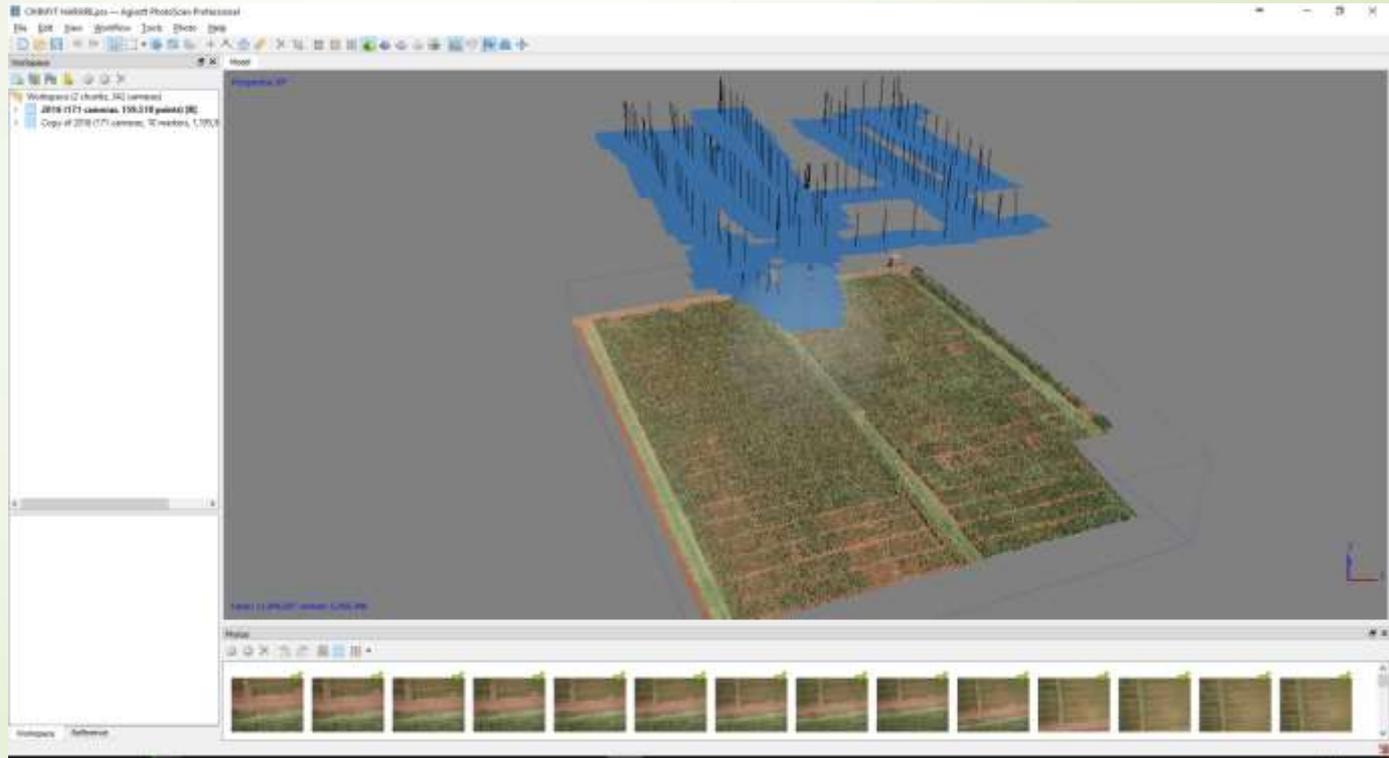
Semi-automatic process should be as efficient and consistent as possible.

Agisoft Photoscan 3D and University of Barcelona MosaicTool built as a plug-in within FIJI (ImageJ)



# 3D modelling example, low N trial, 2016 Maize phenotyping, Zimbabwe

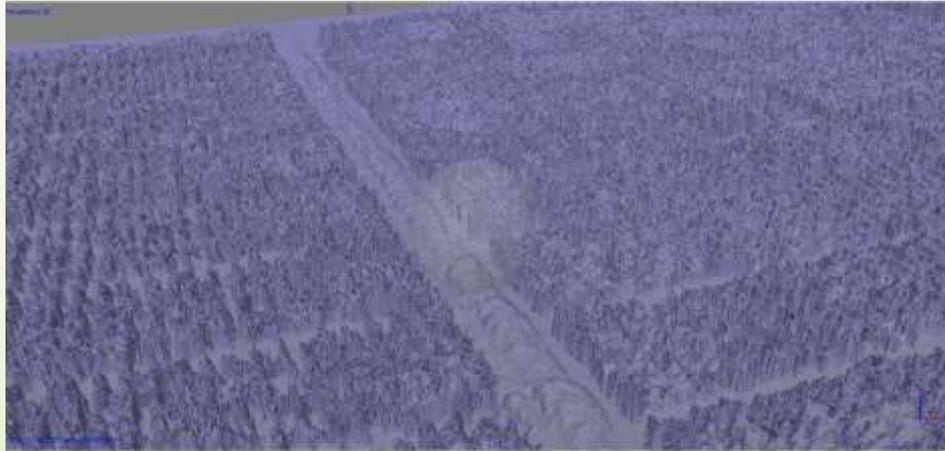
Agisoft Photoscan Dense cloud + camera alignment



## Dense cloud creation and dense cloud classification (DSM)



3D wireframe “mesh” and triangulated surface model



## Benefits of 3D models for UAV plant phenotyping

- Orthorectification of UAV movement/terrain effects
- 3D related data products

## Costs of 3D models for UAV plant phenotyping

- High computing requirements (CPU+GPU)
- Commercial software options (educational discounts can be great)
- Required high percentage of image overlap (continuous capture)

## Dense cloud model over 3D mesh with texture



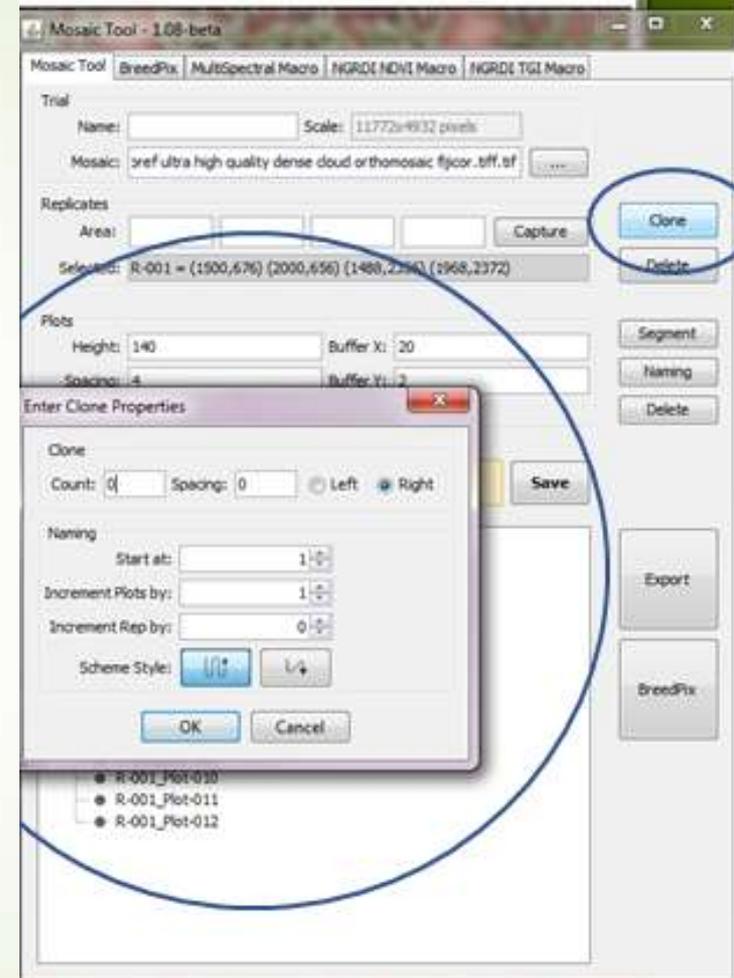
200 million points per ha, 1.5 cm xy

# MosaicTool: University of Barcelona FIJI Plugin (ImageJ)



Allows for segmenting (and processing!) exact same ground spatial area for hundreds of plots in less than one student/hour. Supports RGB, Multispectral (4-12 bands) and TIR sensor indexes.

Field season post-processing down from 9 months to 6 weeks!!!



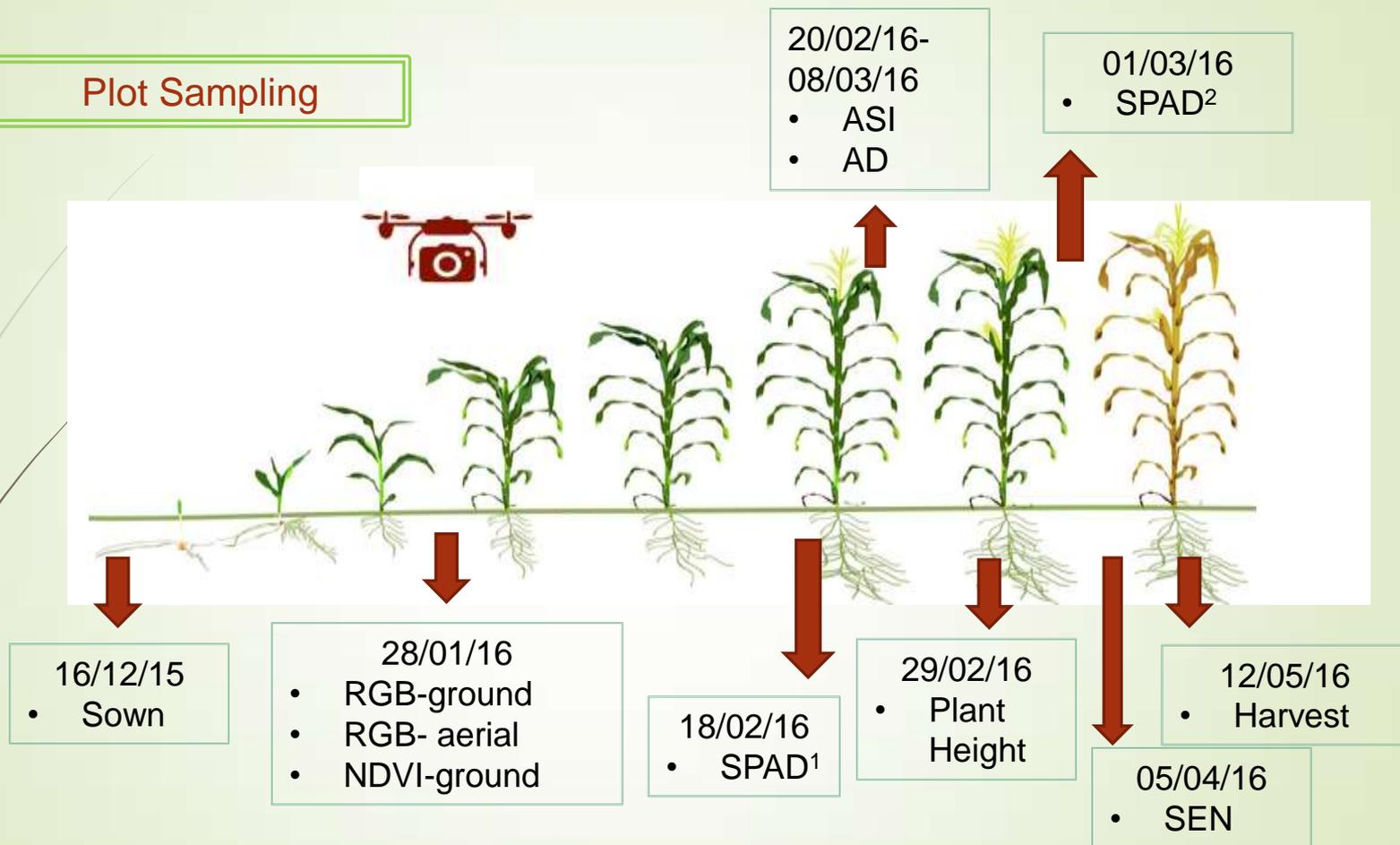
Kefauver et al. In preparation

# RGB comparisons from UAV and ground for low N phenotyping Zimbabwe

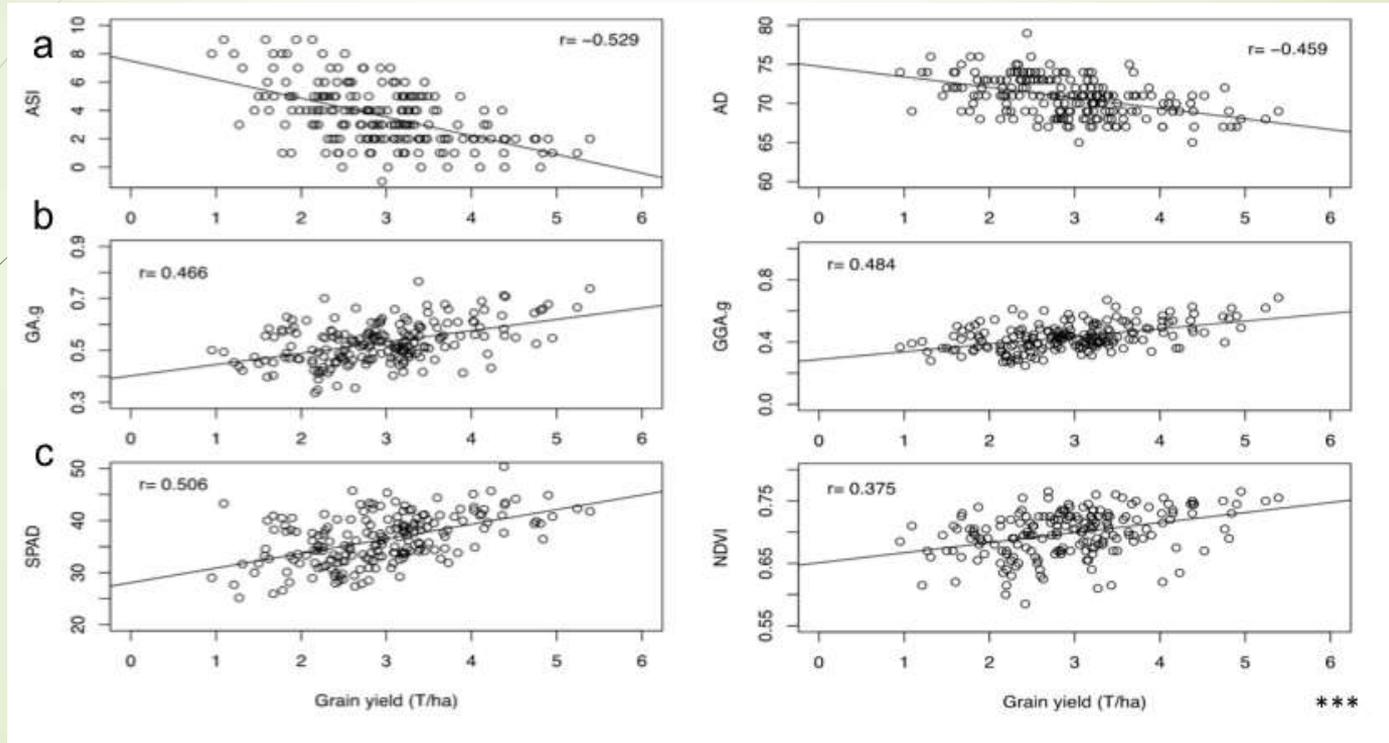


Figure 3. Examples of the differences in resolution between the images of maize taken at ground level and aerial level. A) maize plants at ground level. B) maize image at ground level with GA (Green Area). C) maize image at ground showing GGA (Greener Green Area). D) maize image at aerially. E) maize aerial image with GA (Green Area). F) maize aerial image with GGA (Greener Green Area).

## Plot Sampling



# RGB comparisons from UAV and ground for low N phenotyping Zimbabwe



# Multivariate models assessments

Estimations of Grain Yield (GY) with different categories of remote sensing traits including: RGB ground and aerial level and agronomic data, such as ASI (Anthesis Silking Interval), AD (Anthesis Data), MOI (Moisture), SEN (Canopy Senescence) and PH (Plant Height).

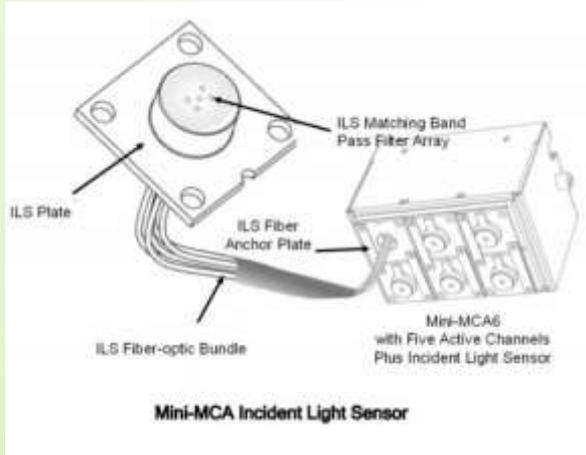
	R <sup>2</sup>	P
RGB ground	0.249	***
RGB ground+ Field sensors	0.403	***
RGB aerial	0.239	***
RGB aerial + Field sensors	0.384	***
Agronomic + RGB ground	0.559	***
Agronomic + RGB aerial	0.560	***

\*\*\*,  $P < 0.001$

# Multispectral sensors for UAVs

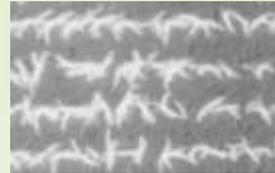
Working with multiple sensors each with a specific bandpass filter. Tetracam sensors with real-time calibration to reflectance with a matching micro filter array in the incident light sensor.

Mosaic ortho-rectification of 11 bands using Agisoft PhotoScan, which permits 11 band multispectral 3D point clouds.



# Multispectral Indexes

Normalized Difference Vegetation Index (NDVI)	$(R_{840}-R_{670})/(R_{840}+R_{670})$	(Rouse et al., 1973)
Soil Adjusted Vegetation Index (SAVI)	$(R_{840}-R_{670})/(R_{840}+R_{670}+L)*(1+0.5)$	(A. R. Huete, 1988)
Optimized soil-adjusted vegetation index (OSAVI)	$((1+0.16)*(R_{780}-R_{670}))/((R_{780}+R_{670}+0.16))$	(Rondeaux et al., 1996)
Renormalized Difference Vegetation Index (RDVI)	$(R_{840} - R_{670}) / ((R_{840} + R_{670})^{1/2})$	(Roujean and Breon, 1995)
Enhanced Vegetation Index (EVI)	$2.5*(R_{840}-R_{670})/(R_{840}+(6*R_{670})-(7.5*R_{450})+1)$	(Huete et al., 2002)
Photochemical Reflectance Index (PRI)	$(R_{550}-R_{570})/(R_{550}+R_{570})$	(Gamon et al., 1997)
Modified Chlorophyll Absorption Ratio Index (MCARI)	$[(R_{700}-R_{670})-0.2*(R_{700}-R_{550})]*R_{700}/R_{670}$	(Daughtry, 2000)
Transformed Chlorophyll Absorption in Reflectance Index (TCARI)	$3*(R_{700}-R_{670})-0.2*(R_{700}-R_{550})*(R_{700}/R_{670})$	(Haboudane et al., 2002)
Anthocyanin Reflectance Index 2 (ARI2)	$R_{840}*(1/R_{550}-1/R_{700})$	(Gitelson et al., 2001)
Carotenoid Reflectance Index 2 (CRI2)	$1/R_{550}-1/R_{700}$	(Gitelson et al., 2002)
Water Band Index (WBI)	$R_{950}/R_{900}$	(Peñuelas et al., 1993)



Soil mask

NDVI 0.4-1



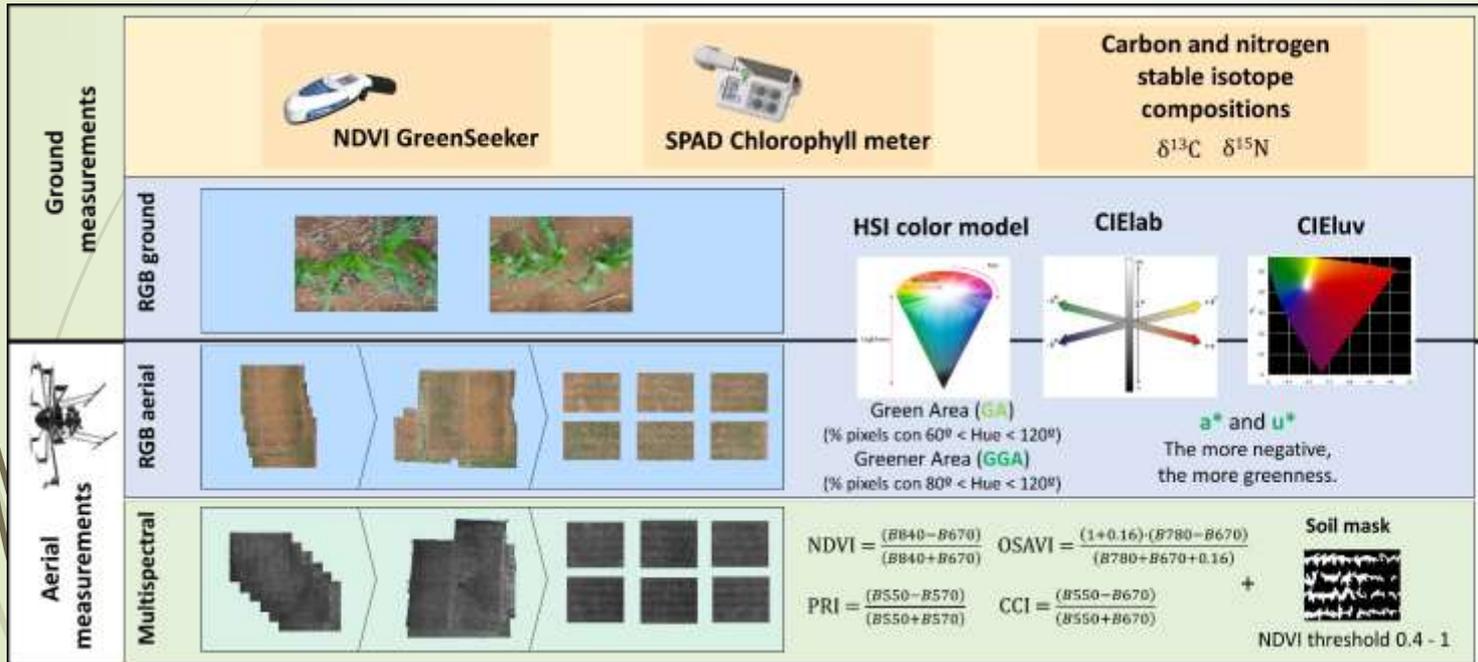
# RGB and multispectral measurements from the UAV and ground compared against yield in low and optimum Phosphorous treatments



	Ground	Aerial
RGB		
Hue		
GA		
a*		

**Figure 2.** Examples of the differences in resolution between the images taken at ground level and at aerially.

# Comparisons of Sensors for Ground and Aerial (UAV) Phenotyping



Data comparison

Platform level:  
Ground vs. Aerial

Sensors:  
RGB vs.  
Multispectral

Data application

Management effects

Yield prediction

# RGB and multispectral measurements from the UAV and ground compared against yield in low and optimum Phosphorous treatments

	Grain yield			P content		
	NPF	OP	Comb.	NPF	OP	Comb.
<b>RGB INDICES/GROUND</b>						
Intensity	0.194	-0.217	-0.084	-0.014	-0.067	-0.041
Hue	0.777***	0.732***	0.827***	0.336	-0.370	0.594*
Saturation	0.468*	-0.027	-0.179	0.065	0.247	-0.429*
Lightness	0.459*	-0.014	0.205	0.086	-0.152	0.126
a*	-0.601**	-0.725***	-0.818***	-0.334	0.405*	-0.643**
b*	0.572**	0.228	0.171	0.110	-0.020	-0.157
u*	-0.300	-0.729***	-0.786***	-0.267	0.425*	-0.667**
v*	0.642***	0.362	0.434**	0.151	-0.152	0.094
GA	0.816***	0.817***	0.878***	0.111	-0.369	0.707**
GGA	0.822***	0.816***	0.877***	0.122	-0.367	0.711**
<b>RGB INDICES/AERIAL</b>						
Intensity	-0.223	-0.715***	-0.620***	0.166	0.021	-0.359
Hue	0.731***	0.796***	0.888***	-0.062	-0.361	0.624**
Saturation	0.149	0.266	-0.235	-0.539*	-0.112	-0.581*
Lightness	-0.102	-0.653***	-0.526***	0.109	-0.047	-0.316
a*	-0.856***	-0.784***	-0.883***	-0.284	0.339	-0.750**
b*	0.192	0.002	-0.292*	-0.466*	-0.221	-0.575*
u*	-0.830***	-0.777***	-0.873***	-0.424*	0.284	-0.777**
v*	0.318	0.084	0.016	-0.333	-0.337	-0.283
GA	0.837***	0.814***	0.891***	0.139	-0.343	0.693**
GGA	0.790***	0.752***	0.837***	0.206	-0.309	0.697**

Correlations were studied across plots within the non-phosphorus fertilization (NPF) and the optimal phosphorus (OP) trials, as well as both in combination (Comb.). Levels of significance: \*P < 0.05; \*\*P < 0.01; \*\*\*P < 0.001.

	Grain Yield			P Content		
	NPF	OP	Comb.	NPF	OP	Comb.
<b>MULTISPECTRAL INDICES</b>						
NDVI <sub>ground</sub>	0.734***	0.711***	0.883***	0.058	-0.423*	0.869***
NDVI	0.629***	0.643***	0.823***	0.324	-0.347	0.800***
SAVI	0.662***	0.644***	0.823***	0.159	-0.269	0.790***
OSAVI	0.657**	0.655**	0.829***	0.216	-0.303	0.797***
RDVI	0.668***	0.650***	0.829***	0.159	-0.286	0.795***
EV	0.613***	0.529**	0.798***	0.119	-0.220	0.782***
PPW	0.039	0.312	0.406**	0.428*	0.032	0.466*
MCARI	0.368	-0.019	0.452**	-0.036	-0.033	0.463*
TCARI	0.172	-0.200	0.238	-0.147	0.055	0.314
TCARI/OSAVI	-0.401*	-0.618**	-0.748***	-0.368	0.283	-0.700***
ARI2	-0.012	0.286	-0.133	-0.286	-0.002	-0.363
CR2	0.016	0.359	-0.091	-0.162	-0.054	-0.364
WEV	0.241	0.595**	0.598***	-0.014	-0.084	0.414*
<b>MULTISPECTRAL BANDS</b>						
B450	-0.348	-0.888***	-0.638***	-0.459	0.318	-0.383
B550	0.261	-0.905***	-0.102	-0.205	0.371	0.036
B570	0.032	-0.529**	-0.419**	-0.498*	0.212	-0.354
B670	-0.302	-0.568**	-0.739***	-0.540*	0.398	-0.731***
B700	-0.116	-0.525**	-0.602***	-0.463*	0.324	-0.567**
B720	0.269	-0.046	0.153	-0.319	0.125	0.047
B790	0.465*	0.477*	0.741***	-0.020	-0.122	0.688***
B840	0.496*	0.550**	0.779***	0.010	-0.137	0.744***
B880	0.442*	0.492*	0.753***	-0.051	-0.129	0.736***
B900	0.425*	0.537**	0.751***	-0.053	-0.083	0.739***
B950	0.390*	0.411*	0.724***	-0.024	-0.091	0.741***

Correlations were studied across plots within the non-phosphorus fertilization (NPF) and the optimal phosphorus (OP) trials, as well as both in combination (Comb.). Levels of significance: \*P < 0.05; \*\*P < 0.01; \*\*\*P < 0.001.



# RGB and multispectral measurements from the UAV and ground compared against grain yield in low and optimum phosphorous treatments

		Non-phosphorus application		Optimum phosphorus	
<b>Grain yield</b>	RGB ground indices	GY = -1.26 - 1.38* + 0.72 ·U* - 440.53 ·GA + 469.95 ·GGA		GY = 4.80 + 13.26 ·GGA	
		<b>R<sup>2</sup> = 0.762***</b>	RSE = 0.535	R <sup>2</sup> = 0.665***	RSE = 0.593
	RGB aerial indices	GY = 12.90 - 0.77 ·α*		GY = 12.30 - 0.12 ·Lightness + 9.97 ·GA	
		<b>R<sup>2</sup> = 0.732***</b>	RSE = 0.531	R <sup>2</sup> = 0.704***	RSE = 0.569
	Multispectral bands	GY = 7.07 + 0.69 ·B840 - 0.69 ·B900		GY = 12.40 - 0.89 ·B450 + 0.37 ·B670 + 0.14 ·B900	
		R <sup>2</sup> = 0.360**	RSE = 0.838	R <sup>2</sup> = 0.668***	RSE = 0.616
	Multispectral indices	GY = -0.52 + 11.95 ·SAVI		GY = -19.27 - 1476.63 ·NDVI + 1011.59 ·SAVI + 16.54 ·WBI - 14.96 ·OSAVI	
		R <sup>2</sup> = 0.396**	RSE = 0.798	R <sup>2</sup> = 0.558*	RSE = 0.728



# Conclusions Low N and P Maize Phenotyping with UAVs

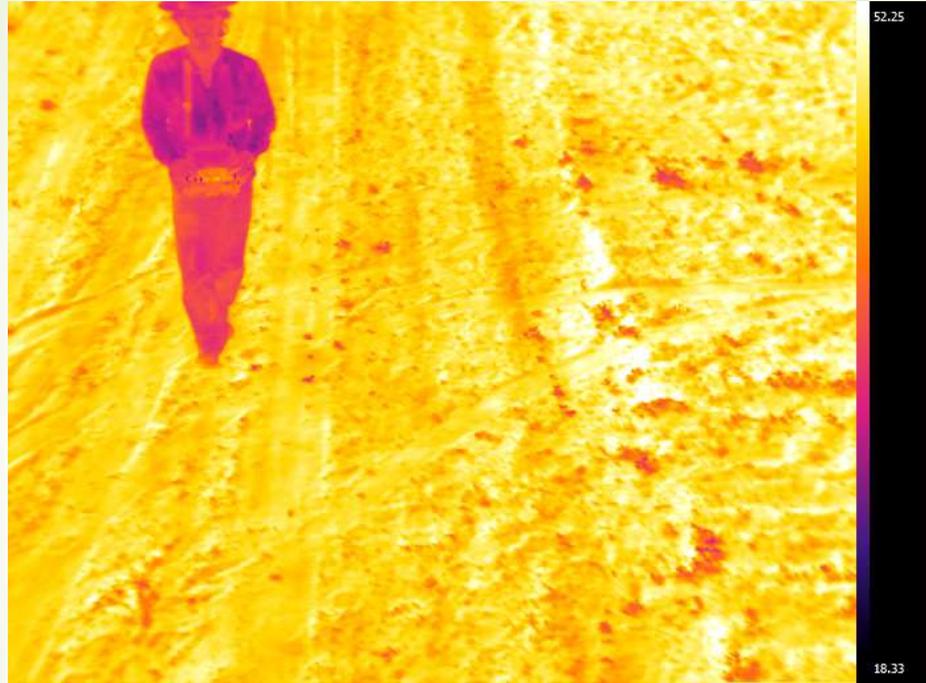
- ▶ Maize breeding shows promise for improving GY in low N environments, variability in performance suggests potential for further improvements.
- ▶ RGB sensors are functional technology from the ground or UAV, similar to SPAD or NDVI, but complimentary to some agronomic measurements.
- ▶ Growth stage timing is critical in order to optimize HTPP benefits to plant breeding. Gains with new technologies 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.

# In collaboration with:



## Thank you for your attention

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