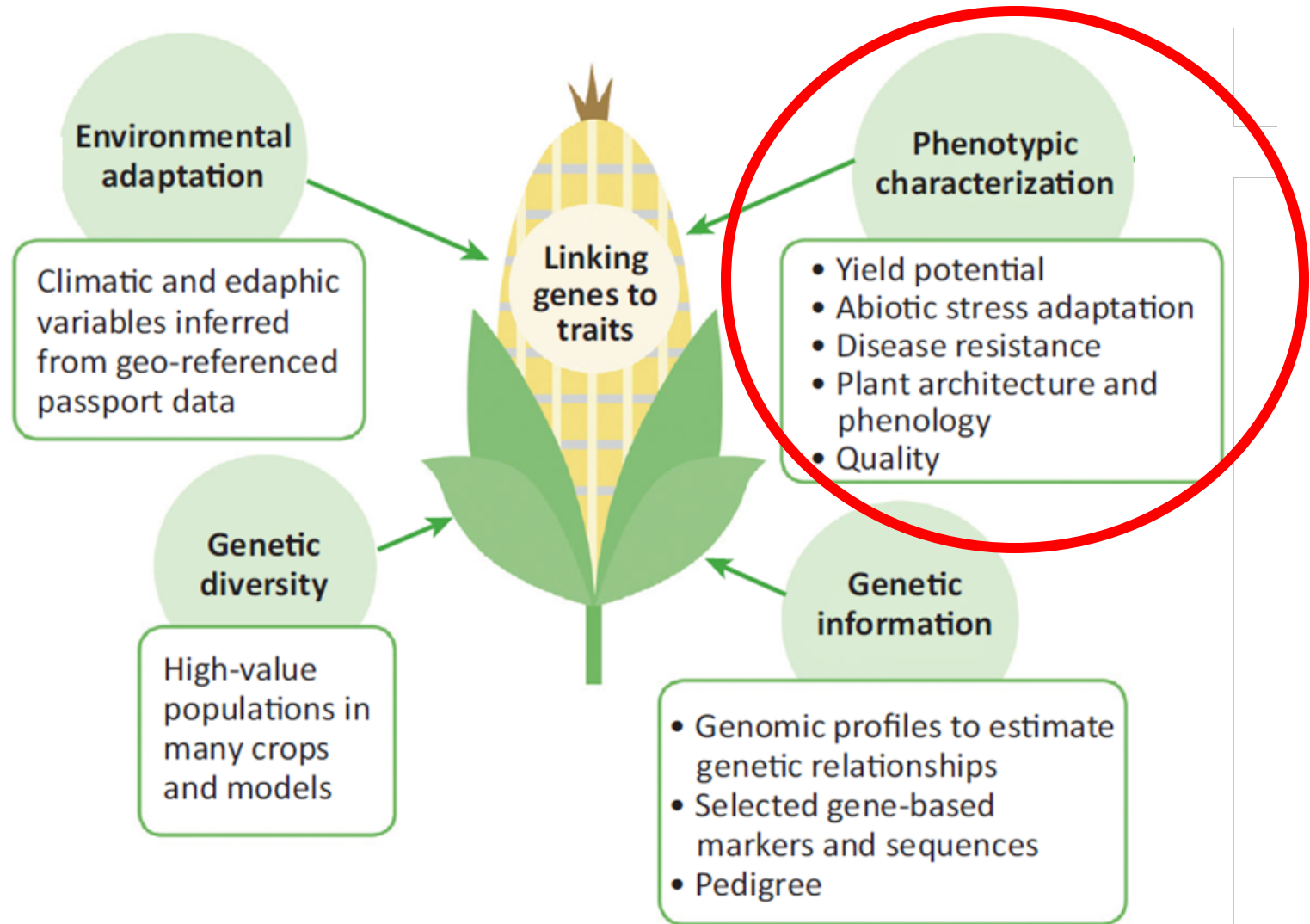




High-throughput phenotyping in maize breeding

The CIMMYT's experience

Crop breeding pillars



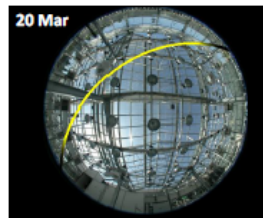
TRENDS in Plant Science

Phenotyping – still a bottleneck

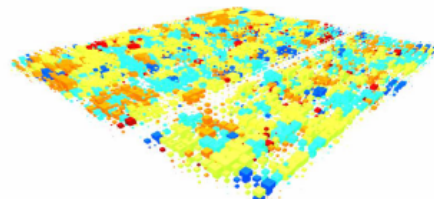
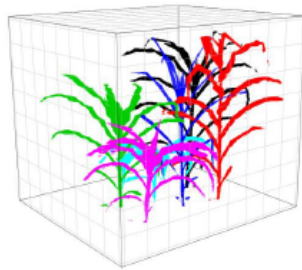
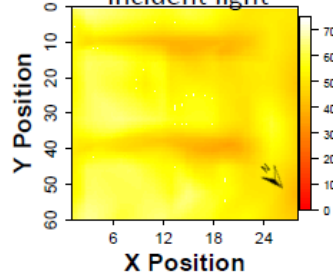
Controlled Environments



Daily sun paths



Spatial distribution of incident light



3D plant reconstruction

Cabrera-Bosquet et al. 2016 *New Phytologist*

Virtual canopy scene of 1680 plants in the glasshouse

Field



The world's first Field Scanalyzer is up and running at Rothamsted Research



A unique facility for field phenotyping has been officially launched at Rothamsted Research.

Lemmatech, Montes et al 2011, *FCR*;
Romano et al. 2012 *Comp. Elect. Agric.*

Traits



anthesis-silking interval

Bolaños and Edmeades 1996 FCR

Tools



Manuals



Bänzinger et al. 2000

Iván Ortiz-Monasterio

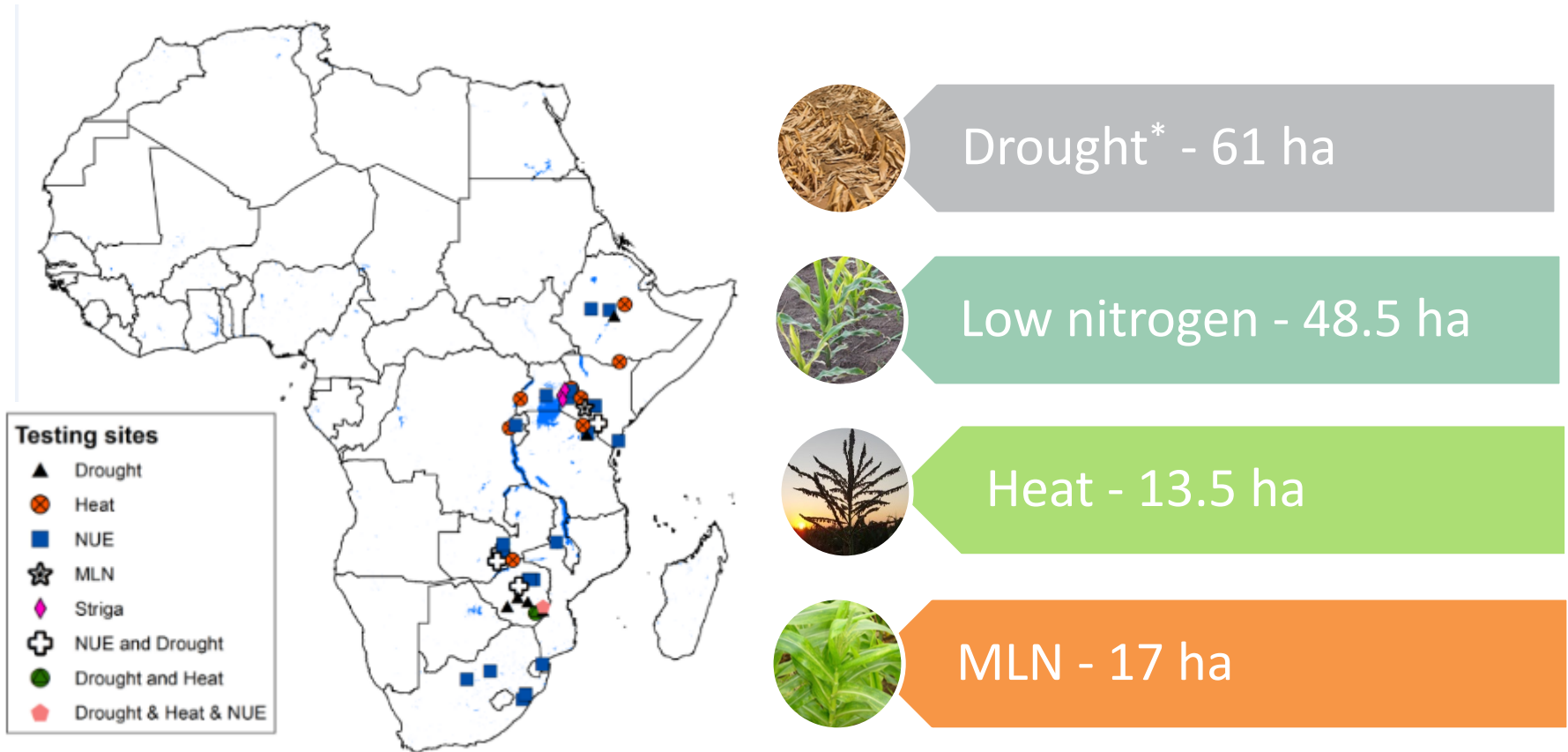


Outline

- **Field phenotyping: multilocation – facilities**
- **Proximal sensing & imaging: affordable alternatives**
 - RGB – indices
 - Other uses of RGB images
 - Lab approaches: NIRS
- **More than just phenotyping**
- **Conclusion**



Large testing network



*Including CFT sites

Cairns updated from Prasanna et al. 2013

Reducing field variability: managed growth conditions

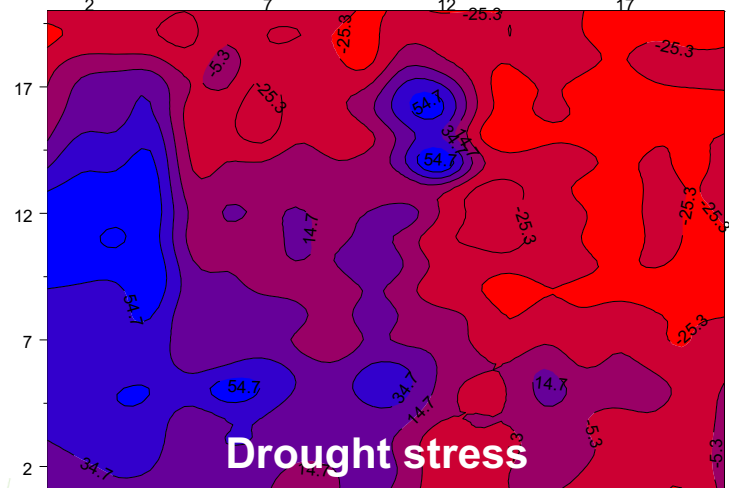
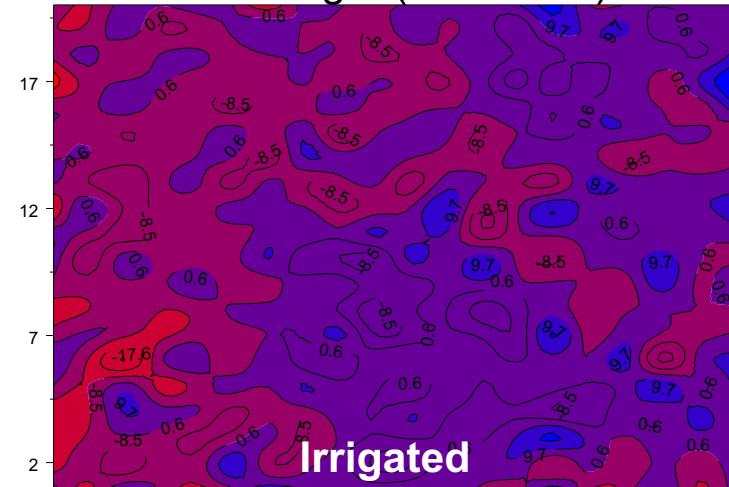
Variance components [†]	Well-watered	Drought stress	Combined drought and heat stress
σ_g^2	0.35	0.12	0.07
$\sigma_{g \times e}^2$	0.24	0.36	0.12
σ_e^2	0.48	0.39	0.18
No. of locations	7	7	3
H	0.84	0.64	0.50

Cairns et al. 2013 Crop Sci.

Test environment	Variance components [†]		
	σ_g^2	σ_{ge}^2	σ_e^2
Early maturity group			
Optimal	28.02 ± 11.14	24.17 ± 8.24	47.81 ± 13.95
Managed drought	14.39 ± 9.30	14.58 ± 4.17	71.04 ± 8.08
Random abiotic stress	10.29 ± 8.32	23.37 ± 11.76	66.34 ± 14.75
Low N	19.01 ± 10.66	23.86 ± 11.30	57.13 ± 14.18
Late maturity group			
Optimal	22.26 ± 4.50	22.41 ± 7.11	55.34 ± 7.85
Managed drought	17.57 ± 9.43	15.72 ± 8.33	66.70 ± 13.52
Random abiotic stress	10.28 ± 7.28	18.25 ± 6.39	71.47 ± 11.23
Low N	15.69 ± 6.95	15.35 ± 4.77	68.95 ± 8.84

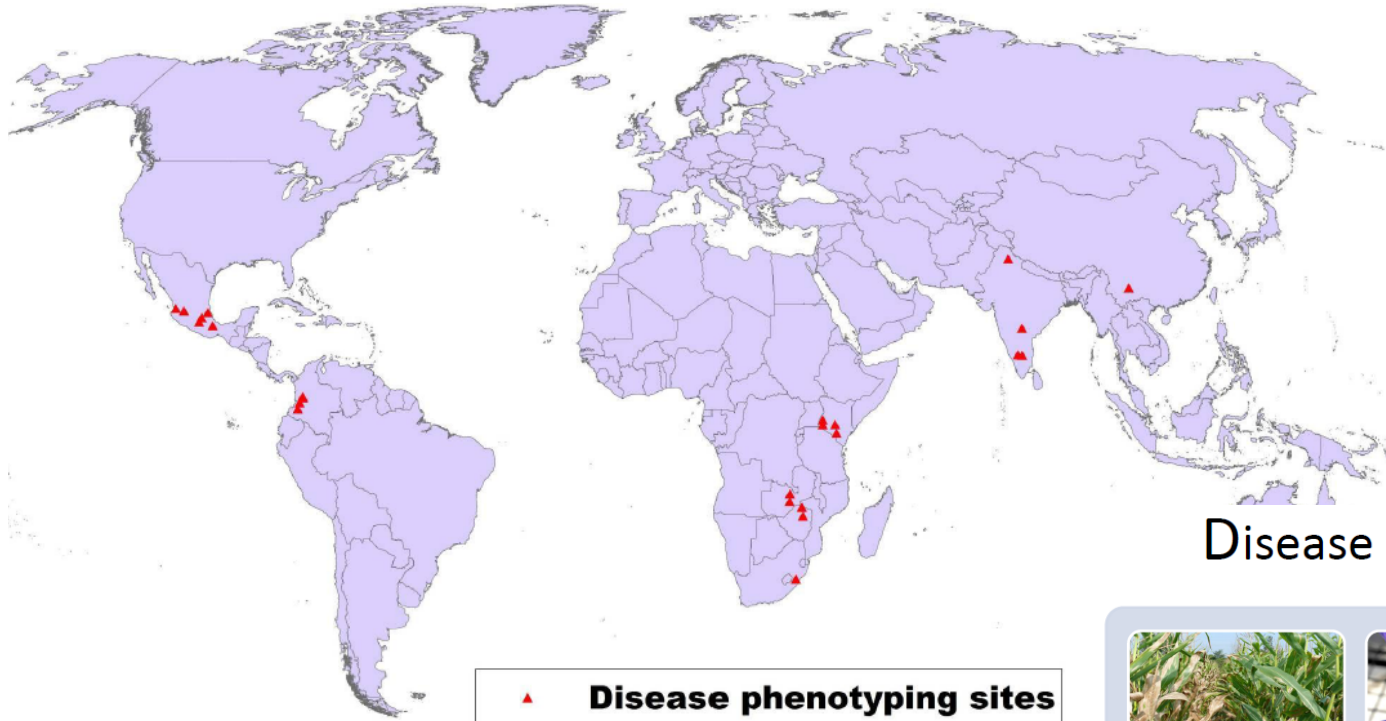
Weber et al. 2012 Crop. Sci.

Plant Height (Residuals)



Burgueño and Wilcox

Disease phenotyping sites



▲ **Disease phenotyping sites**

Disease stress imposition



Natural
infestation
Hotspot



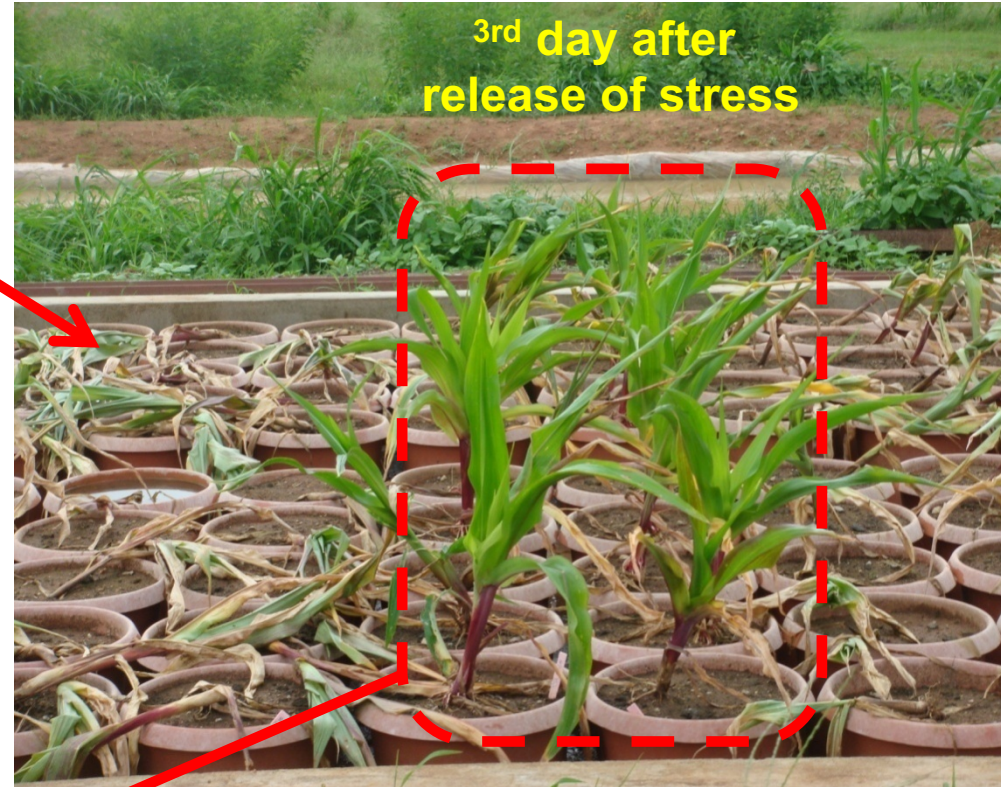
Artificial
infestation
Inoculation



Facilitated
infestation
Misting

Adapted from Cairns

Water-logging at vegetative growth stage



Adapted from Zaidi



Root phenotyping

- ▶ **Structural traits:** root depth, length, volume, root-length density, dry weight
- ▶ **Functional traits:** water use during stress (WU) & Transpiration efficiency (TE)



Adapted from Zaidi

Visual scores: e.g. canopy senescence

Measurement:

- score from 0-10, divide the % of estimated total leaf area that is dead by 10
- initiation & rate of canopy senescence



1 (10%)



3 (30%)



5 (50%)

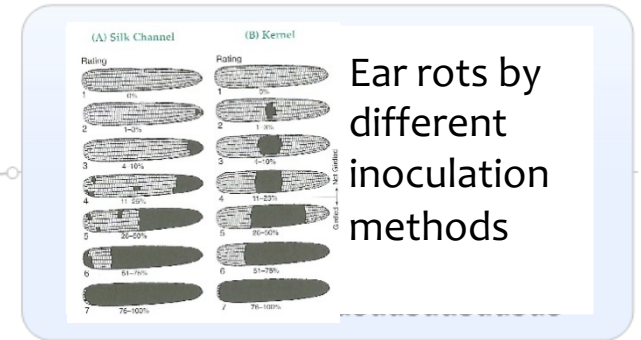
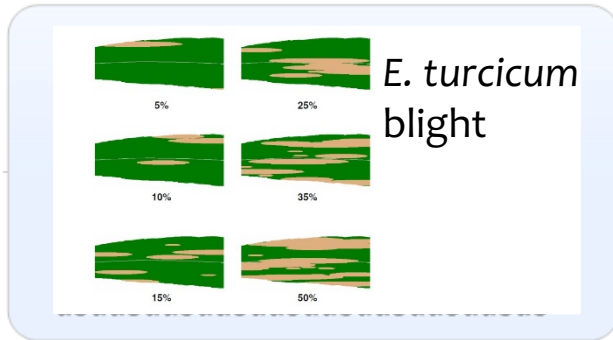


7 (70%)

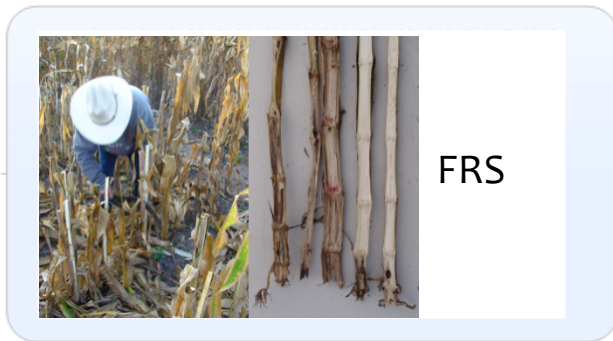


9 (90%)

Visual scores



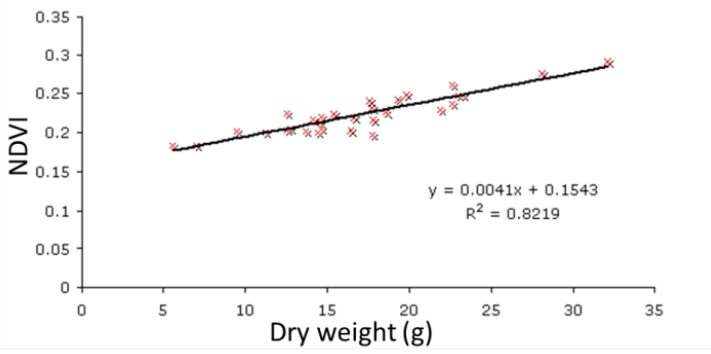
Disease scoring



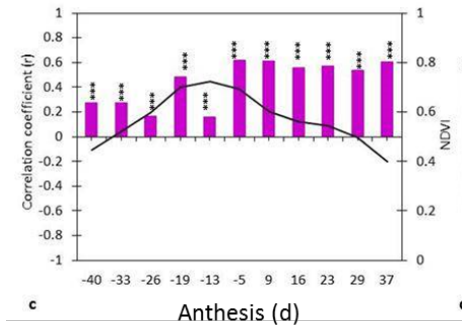
Standard area diagrams and scales

Phenotyping tools: remote sensing approaches

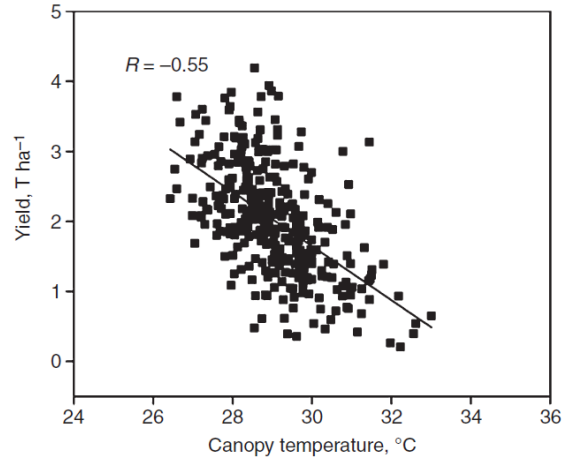
Biomass



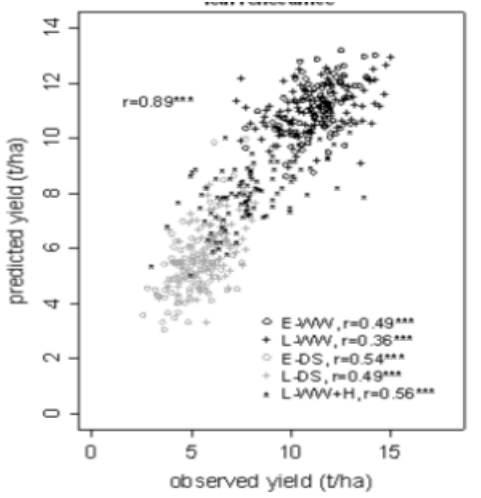
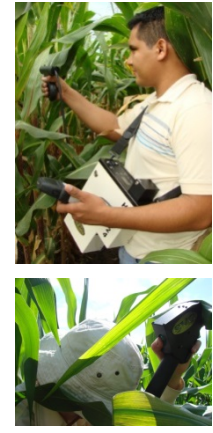
Senescence



Plant water status



Grain yield prediction



Masuka *et al.* 2012; Romano *et al.* 2012; Weber *et al.* 2012; Zia *et al.* 2013; Cairns *et al.* 2013; Zaman-Allah *et al.* 2015; Vergara *et al.* 2016



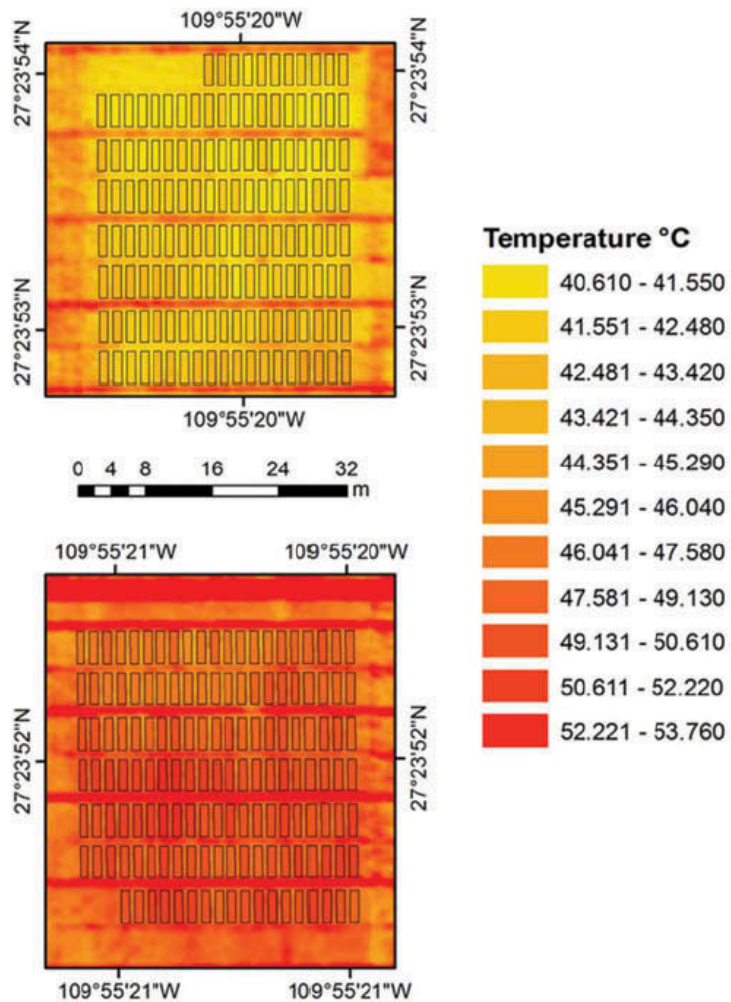


TABLE 5 Expected genetic gains for indices S1 (grain yield, canopy temperature, NDVI, anthesis silking interval), S2 (anthesis silking interval, canopy temperature, and NDVI) and S3 (canopy temperature and NDVI) on grain yield (GY) under heat (HS) and combined heat and drought stress (HS + DS)

Index	GY
HS	
S1	0.486
S2	0.322
S3	0.237
HS+DS	
S1	0.015
S2	0.002
S3	-0.027

Affordable methods for biotic and abiotic stress phenotyping

Data collection

Present: **Visual scoring**

Future: **Canopy digital imaging**



Score = 8
5% leaf loss

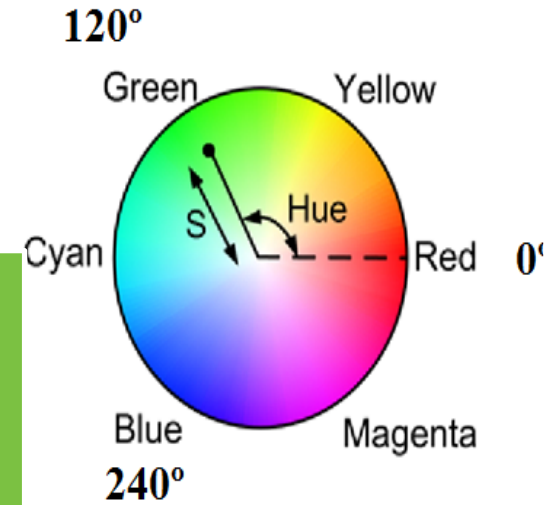
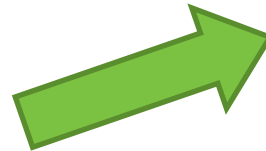
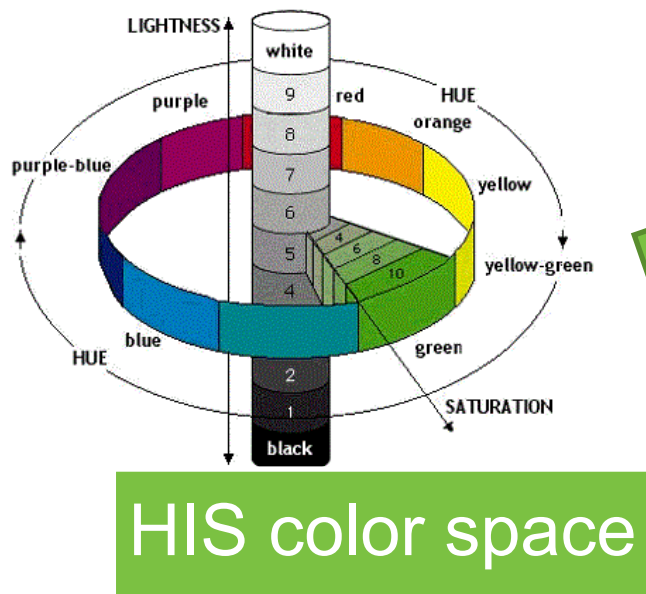
Score = 5
35% leaf loss

Score = 2
80% leaf loss



Adapted from Cairns

RGB – indices



Green Area (**GA**)

(% pixels with $60^\circ < \text{Hue} < 120^\circ$)

Greener Area (**GAA**)

(% pixels with $80^\circ < \text{Hue} < 120^\circ$)



RGB, Green Area, Greener Green Area

MLN plot score 3.0



Maize Leaf Plot RGB



GA (healthy pixels)



GGA (very healthy pixels)

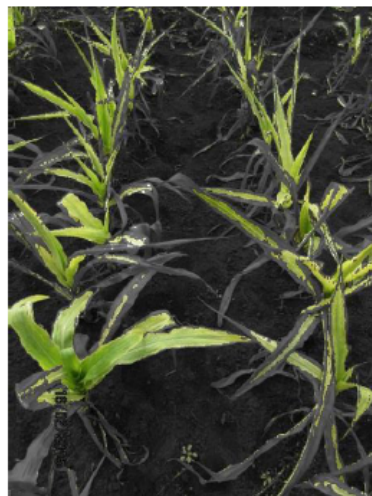


NGRDI (vigor index)

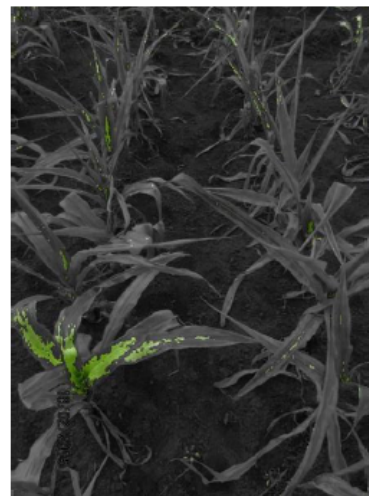
MLN plot score 4.0



Maize Leaf Plot RGB



GA (healthy pixels)



GGA (very healthy pixels)



NGRDI (vigor index)



RGB image processing: vegetation indices



CIMMYT Maize Scanner for RGB field-based phenotyping (released at <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)

Kefauver *et al.*



RGB image processing: image segmentation

ValladolidGX7_May28_50m_Regadio_stitch.png (16.7%)
2660x4592 pixels; RGB; 47MB

(Fiji Is Just) ImageJ
File Edit Image Process Analyze Plugins Window Help
(Fiji Is Just) ImageJ 2.0.0-rc-49/1.51d; Java 1.8.0_91 [32-bit];

Mosaic Tool
BreedPix

Mosaic Image: C:\Users\FisiologiaVegetal\Desktop\FIJI Breedpix\Valladolid\Valladolid May 28\ValladolidGX7_May28_50m_Regadio_stitch.png
Set Scale: 2660x4592 pixels

Replicate ID: A1
Replicate Area: Dimensions = (414,582) (522,4158) (1110,4128) (996,576) Capture
Plot Height: 95 pixel
Plot Spacing: 43 pixel
Plot Numbering: Start At: 1 Top Down Bottom Up Disable
Plot Skew: South West South East

Draw Plots
Preview: Plot 1

Plot 1
Plot 2
Plot 3
Plot 4
Plot 5
Plot 6
Plot 7
Plot 8
Plot 9
Plot 10
Plot 11
Plot 12
Plot 13
Plot 14
Plot 15
Plot 16
Plot 17
Plot 18
Plot 19
Plot 20
Plot 21
Plot 22
Plot 23
Plot 24
Plot 25
Plot 26

Copy All to ROI Manager
Update From ROI Manager

12. Once all the plots have been modified, Right click on the plot list and select "Update from ROI Manager". All the plots will be updated based on the changes done in the ROI manager

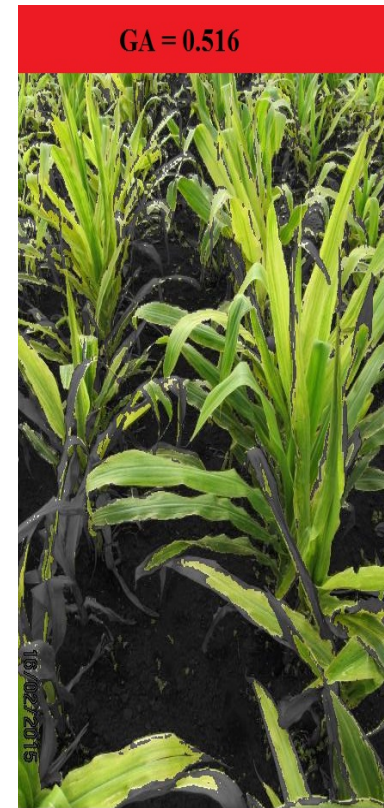
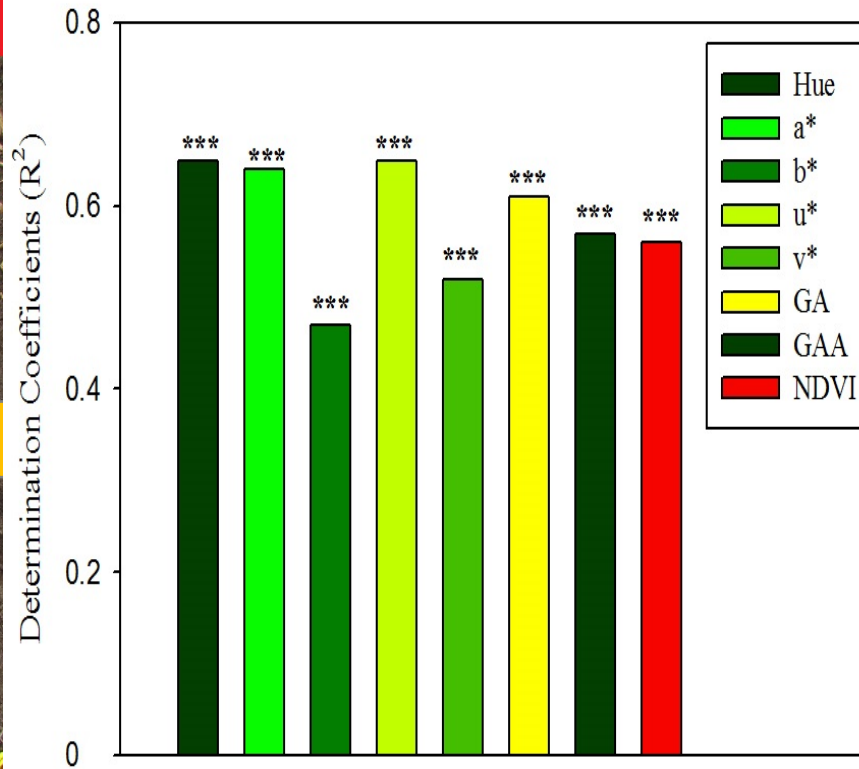
13. Once all the plots are correct, we can proceed to click on the "Add Replicate" button, to finish the replicate



RGB vs Spectral indices

Maize Lethal Necrosis evaluation in Kenya

MLN Index

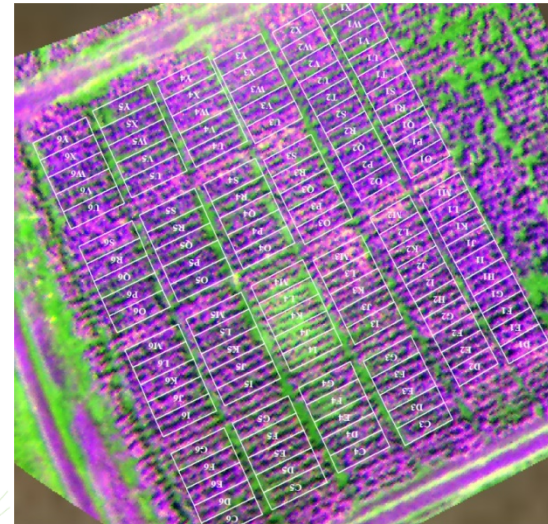


Vegetation Indexes

***, $P < 0.001$; **, $P < 0.01$; *, $P < 0.05$; ns, not significant

RGB vs Spect. indices

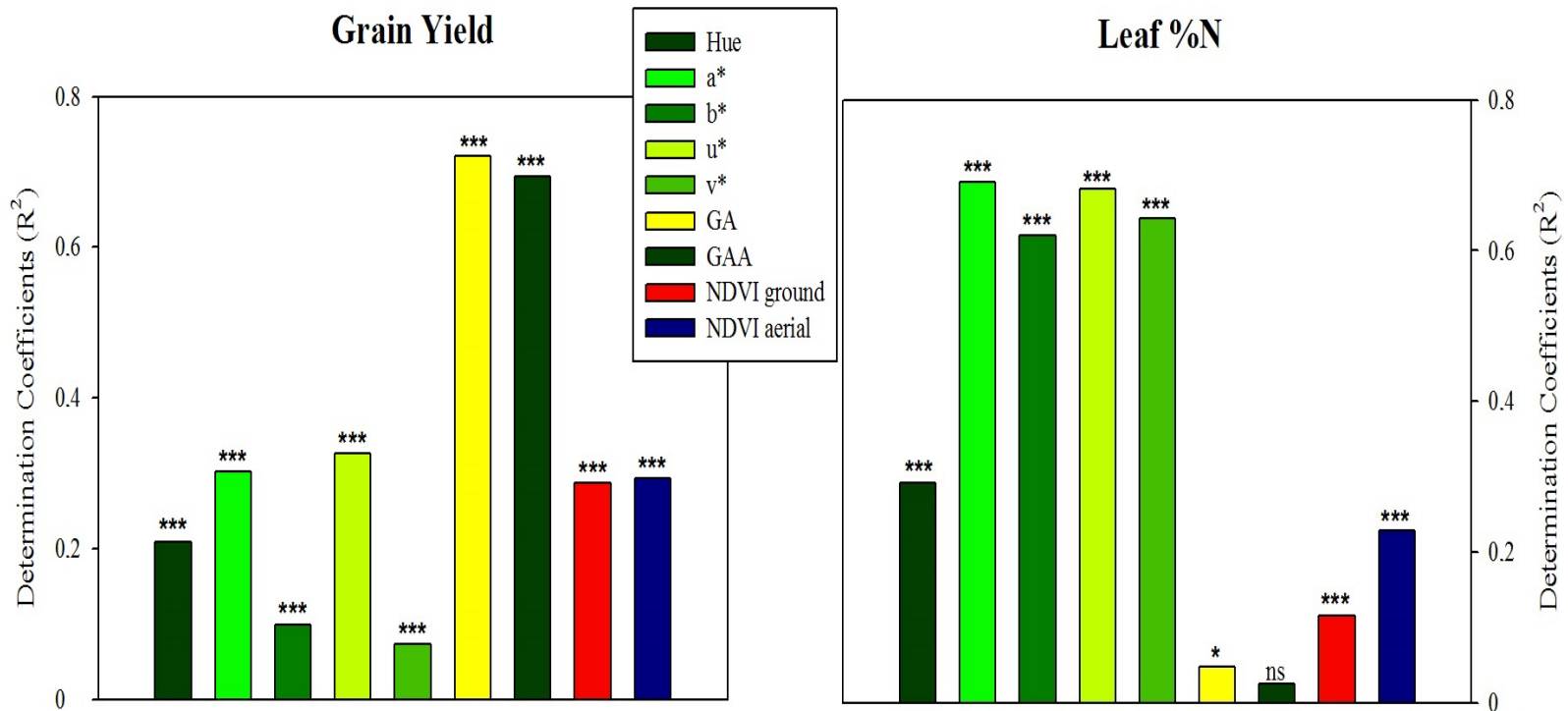
N fertilization treatments in Zimbabwe



CIMMYT's South Africa regional station, Harare

RGB vs Spectral indices

N fertilization treatments in Zimbabwe



Vegetation Indexes

Zaman-Allah et al. 2015. *Plant Methods*

Vergara et al. 2016 *Frontiers in Science*

***, $P < 0.001$; **, $P < 0.01$; *, $P < 0.05$; ns, not significant

Other applications of RGB images: harvest index

Kernels



- Size
- Uniformity/abortion rate
- Kernel weight
- Rot index

Ears

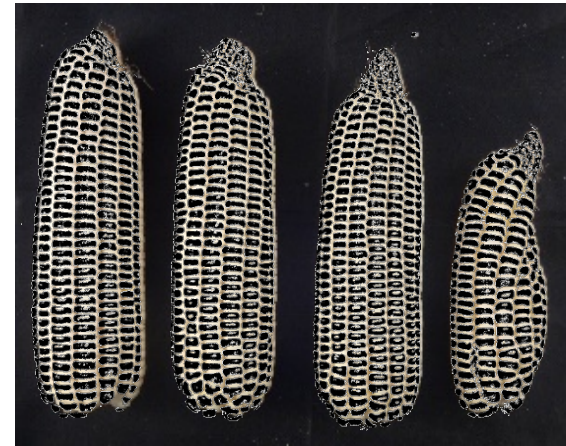


- Number
- Rows
- Size

Stress stage



- Flowering stress
- Grain filling stress



NEAR INFRARED REFLECTANCE SPECTROSCOPY (NIRS)

VERIFICATION

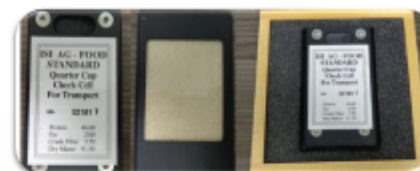
Advanced diagnostic

- Wavelength linearization
- Gain test
- Self test

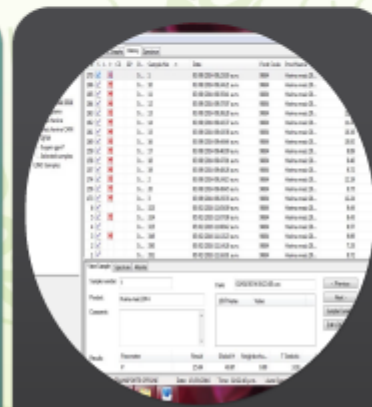
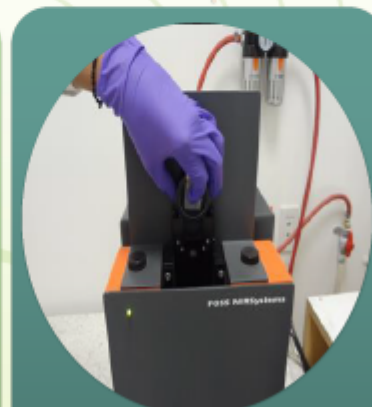
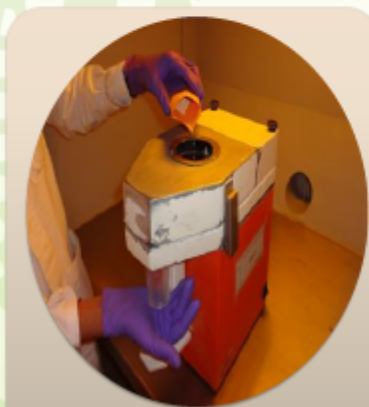
Routine diagnostics

- Performance test

Check cell

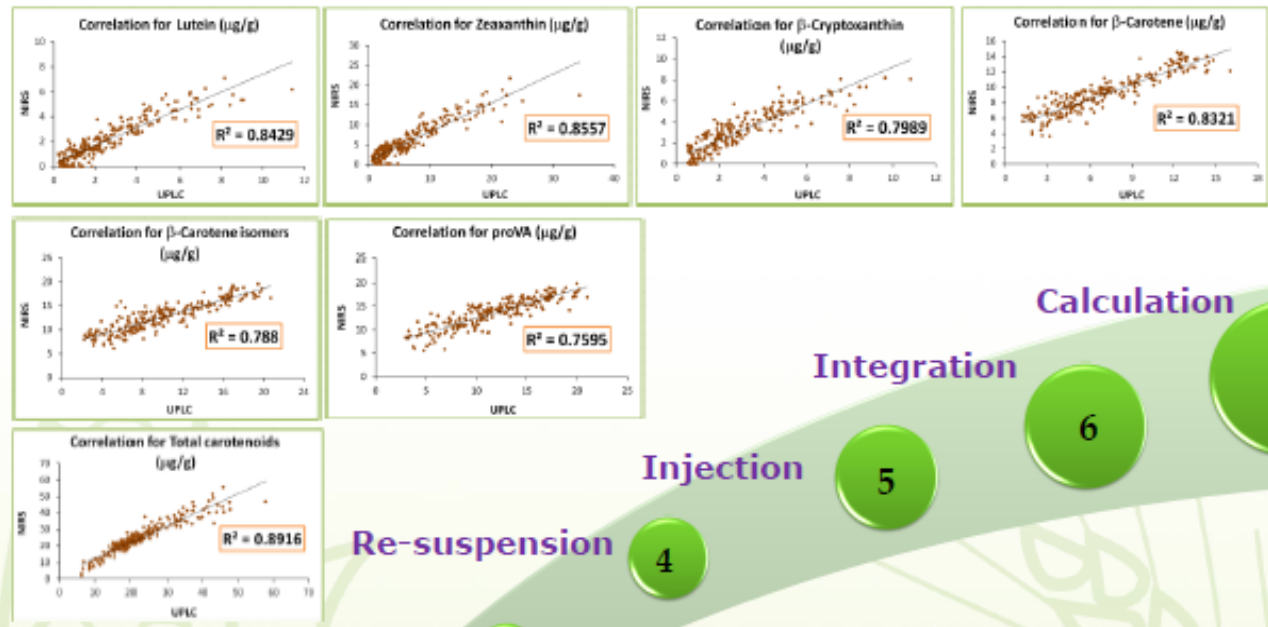


SAMPLE ANALYSIS (Starch, protein, oil, provitamin A, lysine, tryptophan, anthocyanins)



Total time: ~ 4 min

UPLC and NIR for carotenoids in maize



UPLC



Factor	UPLC	NIRS
Accuracy	☹️ Quant. Qual.	😊 Quant.
Time of analysis	☹️ ~3 h	😊 ~4 min
Cost/sample	😊 37.10 US	😊 2.62 US
Risk	😊	😊
Waste generation	☹️	😊
HR	😊 3 people	😊 1 person



NIRS

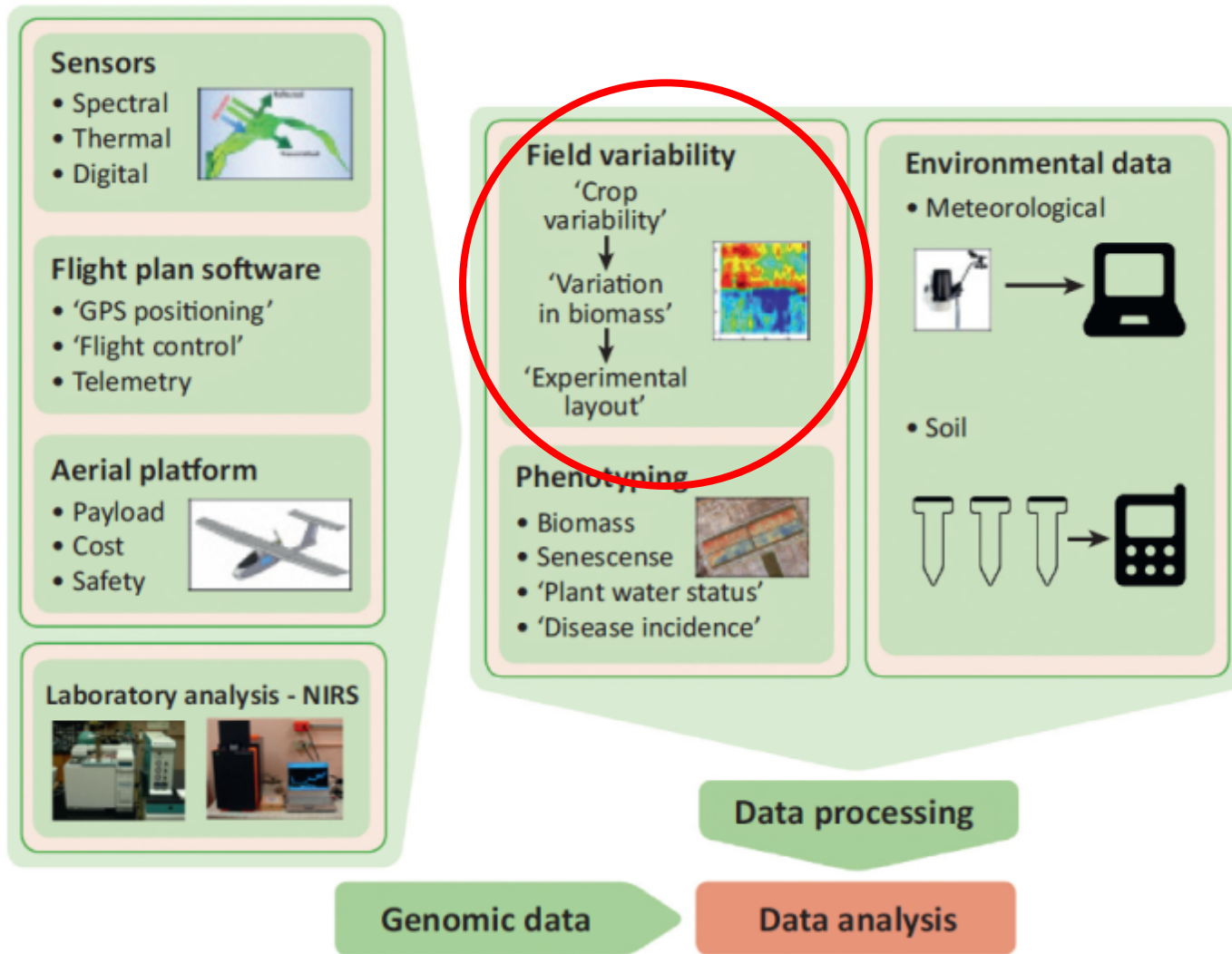
NIRS as alternative to the analysis of other traits



Technique	IRMS		EA	AACC Method	NIRS-prediction			
Parameter	$\delta^{13}\text{C}$	$\delta^{18}\text{O}$	N content	Ash content	$\delta^{13}\text{C}$	$\delta^{15}\text{N}$	Ash	N
Cost per sample	10\$	20\$	3\$	1.5\$	1.0\$			
Time	<10 min	<10 min	<10 min	≈24 h	≈3 min			
Equipment	EA-IRMS		EA	Muffle furnace	NIR spectrometer			

Cabrera-Bosquet et al. 2012 J. Agric. Chem.

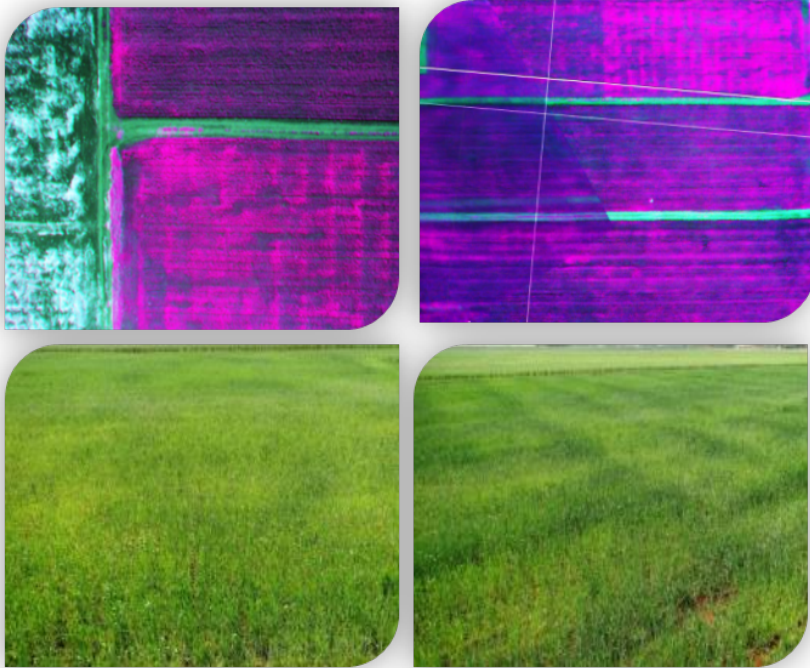
More than traits and tools



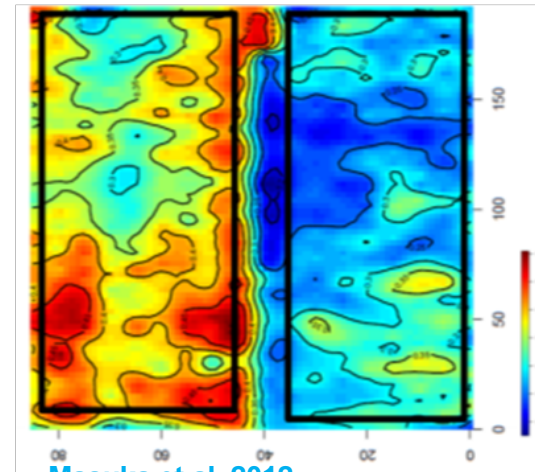
Araus and Cairns 2014

TRENDS in Plant Science

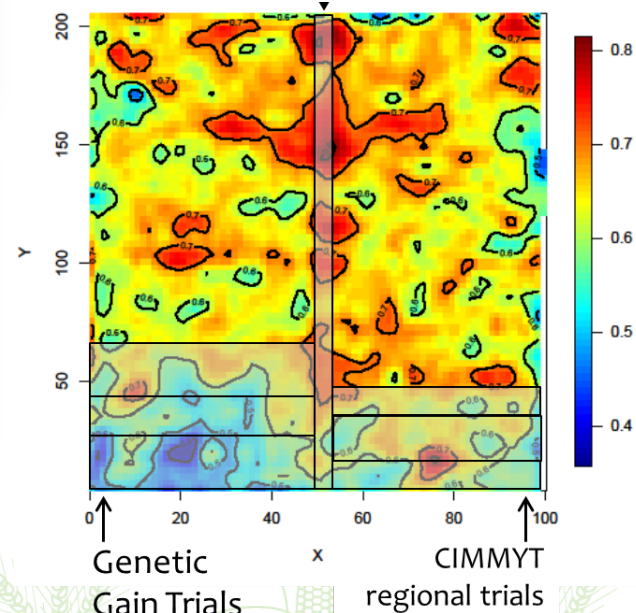
Measuring / reducing spatial variability



Cairns and Zaman-Allah



Masuka et al. 2012



Das and Crossa

Harmonized phenotyping protocols for abiotic stresses



PHENOTYPING FOR ABIOTIC
STRESS TOLERANCE IN MAIZE:

DROUGHT STRESS

M. Zaman-Allah, P.H. Zaidi, S. Trachsel,
J.E. Cairns, M.T. Vinayan and K. Seetharam



PHENOTYPING FOR ABIOTIC
STRESS TOLERANCE IN MAIZE:

HEAT STRESS

P.H. Zaidi, M. Zaman-Allah, S. Trachsel, K. Seetharam,
J.E. Cairns and M.T. Vinayan



PHENOTYPING FOR ABIOTIC
STRESS TOLERANCE IN MAIZE:

WATERLOGGING STRESS

P.H. Zaidi, M.T. Vinayan and K. Seetharam
CIMMYT Asia Maize Program, Hyderabad, India



Conclusions

- **Large multilocation networks**
- **Emphasis in site quality – logistics for particular stresses**
- **Flexible phenotyping platforms**
- **Affordable, high-throughput phenotyping technologies**



Acknowledgements

SHAWN C. KEFAUVER, OMAR VERGARA,

University of Barcelona, Section of Plant Physiology, Barcelona, Spain

BODDUPALLI M PRASANNA

International Maize and Wheat Improvement Center (CIMMYT), ICRAF House, Nairobi, Kenya

**NATALIA PALACIOS, SAMUEL TRACHSEL, IVAN ORTIZ-MONASTERIO,
JUAN BURGEÑO**

CIMMYT , El Batan, Texcoco, Mexico

MAINASSARA A. ZAMAN-ALLAH, JILL CAIRNS

CIMMYT, SE Africa Regional Office, Harare, Zimbabwe



Many thanks! – ¡Muchas gracias!

