

	<p>LOUISIANA STATE UNIVERSITY College of Agriculture School of Plant, Environmental, and Soil Sciences AGRO 7076 HTP in Plant Breeding</p>	
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Enviromics

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Prediction-based model

$$RS = \frac{i \cdot r_{aP} \cdot \sigma_a}{T}$$

$$r_{aP} = \sqrt{h_a^2}$$

$$y = u + X\beta + Zg + Wge + \varepsilon$$



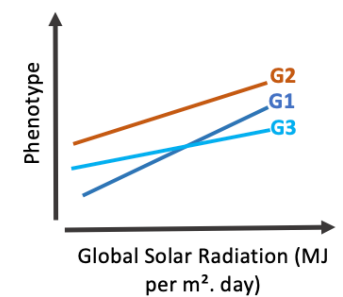
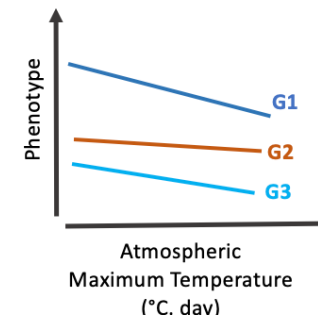
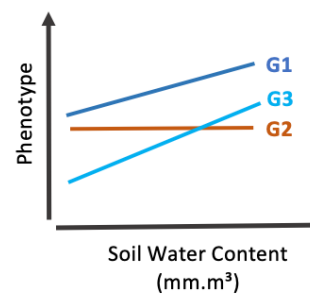
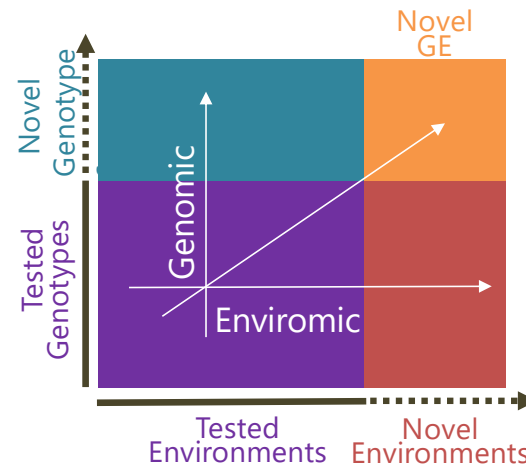
Envirotyping (W):

- Virtually increase the MET
- Allows to better predict gxe
- Optimize cultivar recommendation
- Thus, increase h

Current challenges:

1. Obtain high resolution data
2. Translate information

MET Prediction (Enviromic + Genomic)

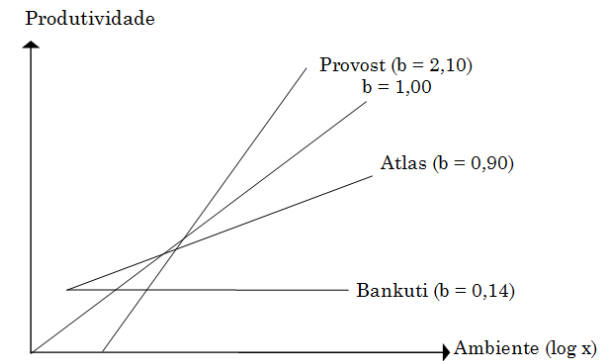
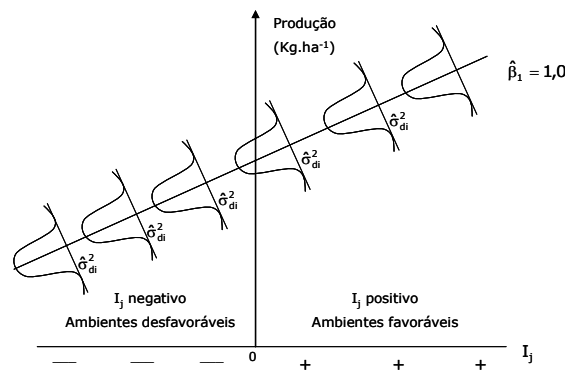
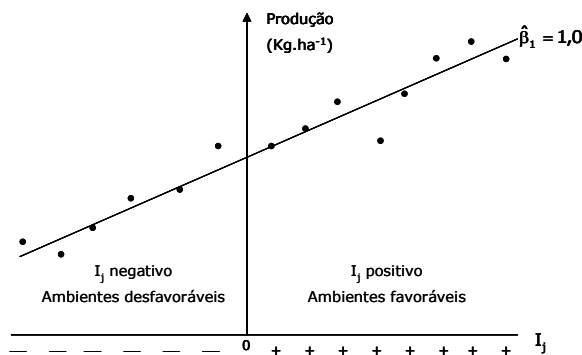


Adaptability and stability – Regression methods

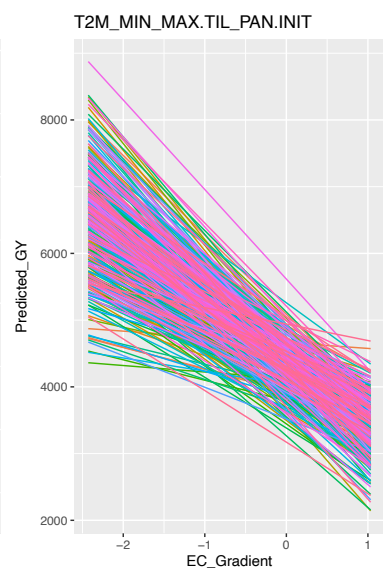
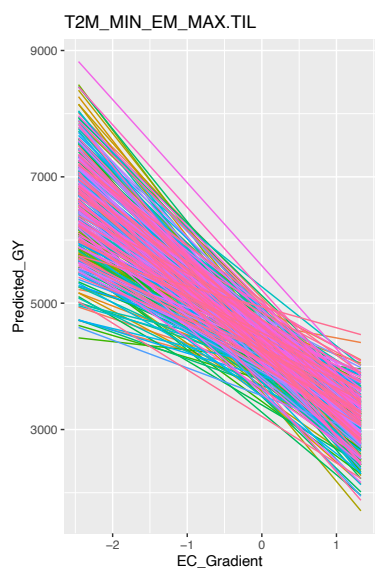
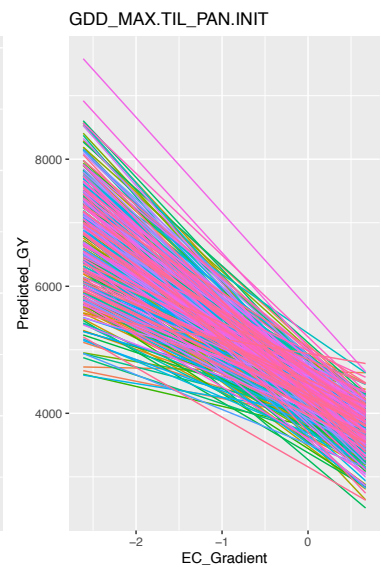
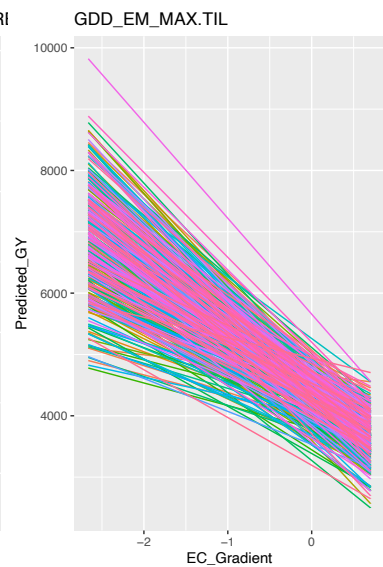
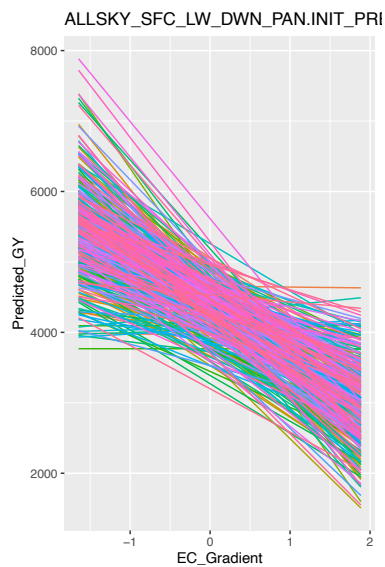
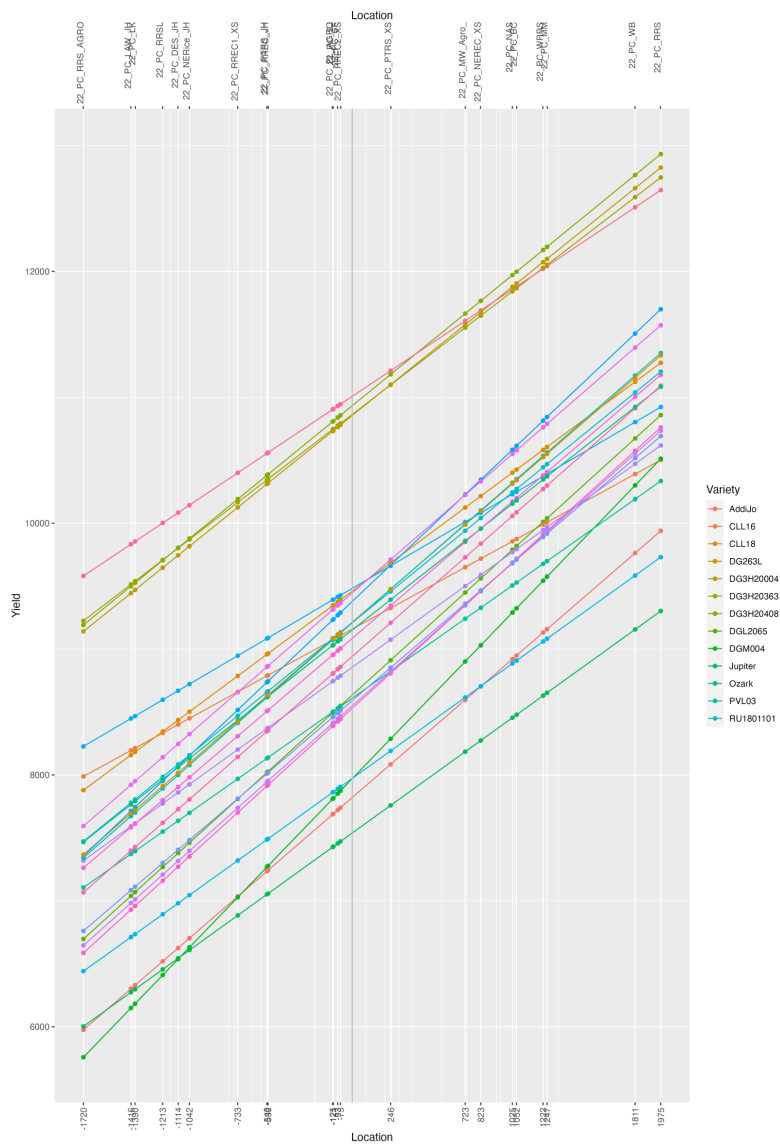
- **Adaptability:** ability to take advantage of environmental variations
- **Stability:** predictable behavior in the face of variations in the environment
- **Finlay K, Wilkinson G (1963):** Linear regression coefficient and the variance of the regression deviations

$$I_j = \bar{I} \cdot j - \bar{Y} \cdot j$$

$$Y_{ij} = m_i + b_i I_j + d_{ij}$$



- d_{ij} : regression deviations predictability (*stability*)
- *What would be the ideal cultivar?*
- Y_{ij} : high overall performance
- $b_i = 1,0$
- $d_{ij} = 0$



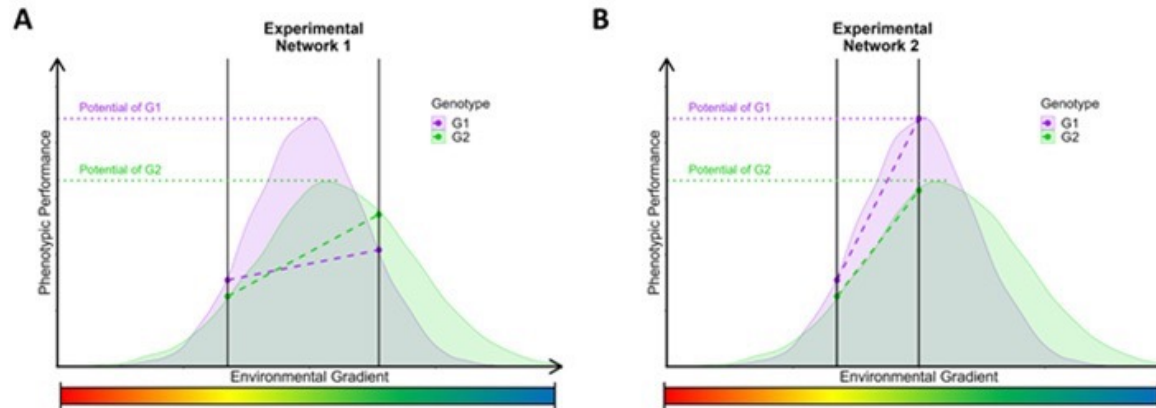
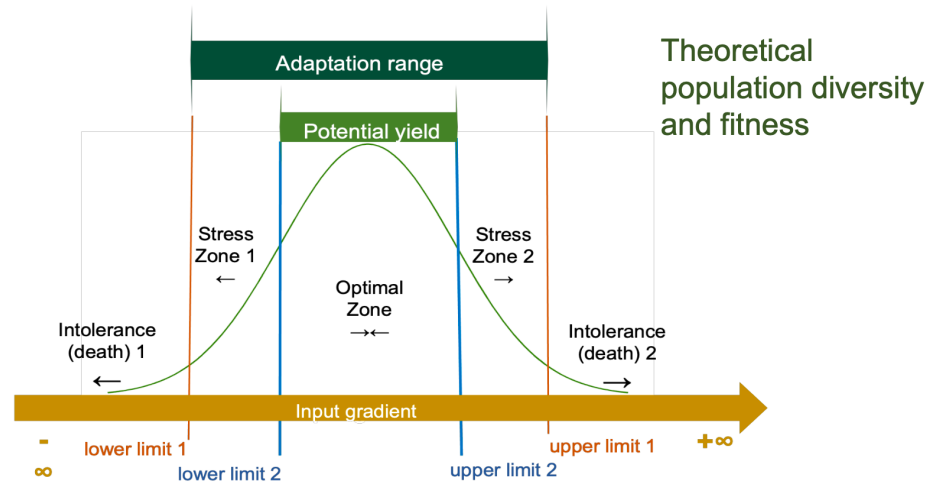


Nonlinear kernels, dominance, and envirotyping data increase the accuracy of genome-based prediction in multi-environment trials

Germano Costa-Neto ¹ · Roberto Fritsche-Neto ¹ · José Crossa ²

Theoretical gradient of some continuous environmental factor

Shelford (1931, 1932) Tolerance Limits and adaptation





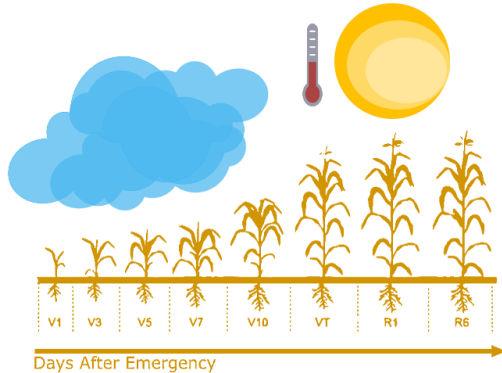
Enviromic Assembly Increases Accuracy and Reduces Costs of the Genomic Prediction for Yield Plasticity in Maize

Germano Costa-Neto^{1,2*}, Jose Crossa^{3,4†} and Roberto Fritsche-Neto^{1,5}

- Cardinals must weight EC
- **Not all** EC are important:
 - *for all traits or*
 - *during the whole cycle*

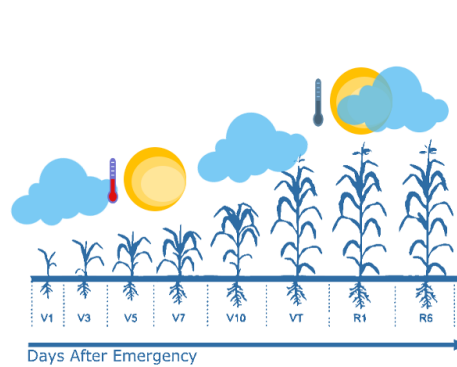
Cumulative values per croplife (**poor description**)

Environment 1



- Precipitation: 560 mm/cycle
- Temperature: 962 °C/cycle
- Radiation: 724 MJ m2/ cycle

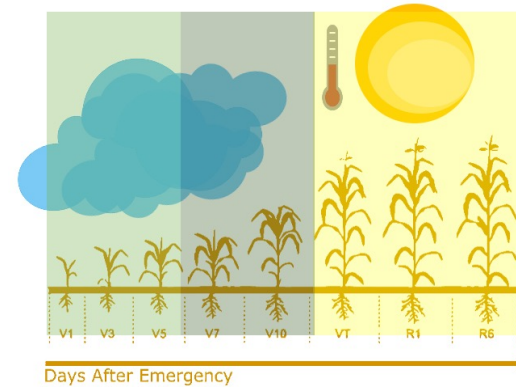
Environment 2



- Precipitation: 560 mm/cycle
- Temperature: 962 °C/cycle
- Radiation: 724 MJ m2/ cycle

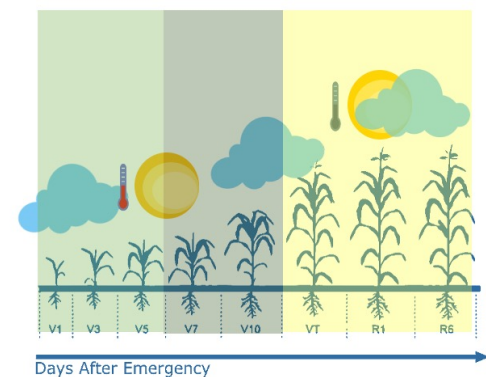
Cumulative values per stage (**better**)

Environment 1



- Precipitation (mm): 300 (T1); 260 (T2); 0 mm (T3)
- Temperature (°C): 150 (T1); 100 (T2); 712 (T3)
- Radiation (MJ/m2): 144.8 (T1); 115.8 (T2); 463.4 (T3)

Environment 2



- Precipitation (mm): 224 (T1); 168 (T2); 168 (T3)
- Temperature (°C): 150 (T1); 100 (T2); 712 (T3)
- Radiation (MJ/m2): 144.8 (T1); 115.8 (T2); 463.4 (T3)

- **Crop-specific tune**

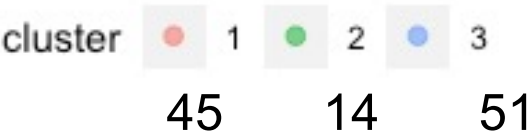
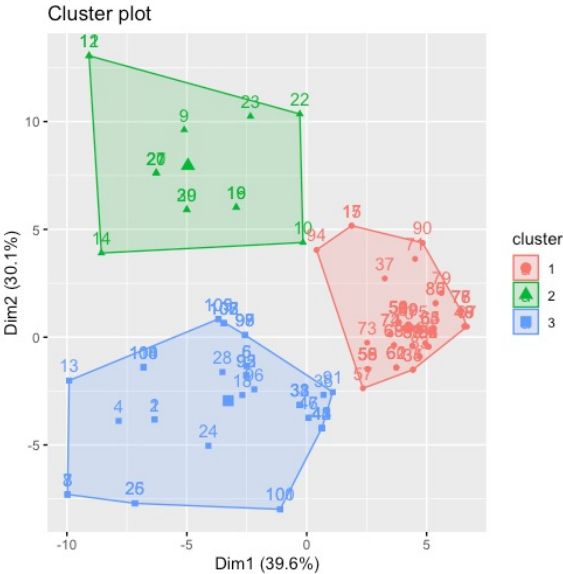
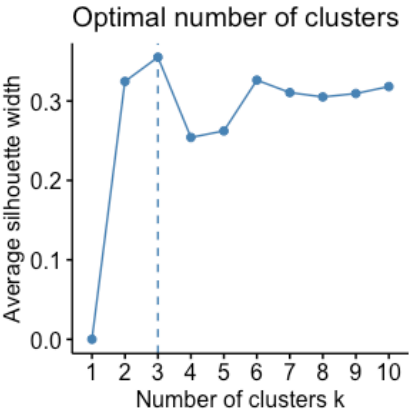
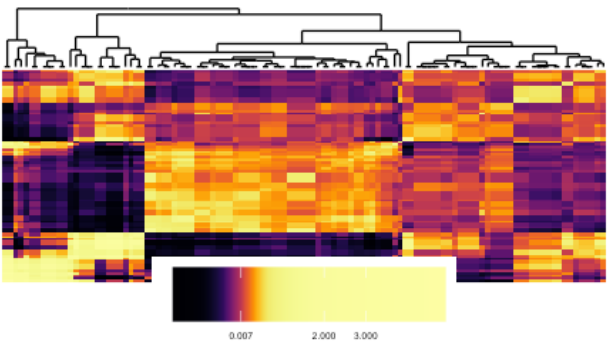
- Tbase1 = 12,
- Tbase2 = 24,
- Topt1 = 33,
- Topt2 = 37,
- Alt = 540

- **Temporal variations**

- From 0 DAE (emergence day) to 14 DAE (appearance of the first leaf, V1).
- From 15 DAE (V1) to 35 DAE (appearance of the fourth leaf, V4).
- From 36 DAE (V4) to 65 DAE (tasseling stage, VT).
- From 66 DAE (VT) to 90 DAE (kernel milk stage, R3).
- From 91 DAE (R3) to 120 DAE (physiological maturity).

Source	Environmental factor	Unit
NASA Power	Top-of-atmosphere insolation	$\text{MJ m}^{-2} \text{d}^{-1}$
	Average insolation incident on a horizontal surface	$\text{MJ m}^{-2} \text{d}^{-1}$
	Average downward longwave radiative flux	$\text{MJ m}^{-2} \text{d}^{-1}$
	Wind speed at 10 m above the surface of the earth	m s^{-1}
	Minimum air temperature at 2 m above the surface of the earth	$^{\circ}\text{C d}^{-1}$
	Maximum air temperature at 2 m above the surface of the earth	$^{\circ}\text{C d}^{-1}$
	Dew-point temperature at 2 m above the surface of the earth	$^{\circ}\text{C d}^{-1}$
	Relative air humidity at 2 m above the surface of the earth	%
	Rainfall precipitation (P)	mm d^{-1}
Calculated ^a	Effect of temperature on radiation-use efficiency	–
	Evapotranspiration (ETP)	mm d^{-1}
	Atmospheric water deficit P-ETP	mm d^{-1}
	Deficit of vapor pressure	kPa d^{-1}
	Slope of saturation vapor-pressure curve	$\text{kPa } ^{\circ}\text{C}^{-1} \text{d}^{-1}$
	Temperature range	$^{\circ}\text{C d}^{-1}$
	Global solar radiation based on latitude and Julian Day	$\text{MJ m}^{-2} \text{d}^{-1}$

E.g., DS in The Philippines



EnvRtype: a software to interplay enviromics and quantitative genomics in agriculture

Germano Costa-Neto ¹, ¹ Giovanni Galli, ¹ Humberto Fanelli Carvalho ¹, José Crossa ² and Roberto Fritsche-Neto ^{1,3}



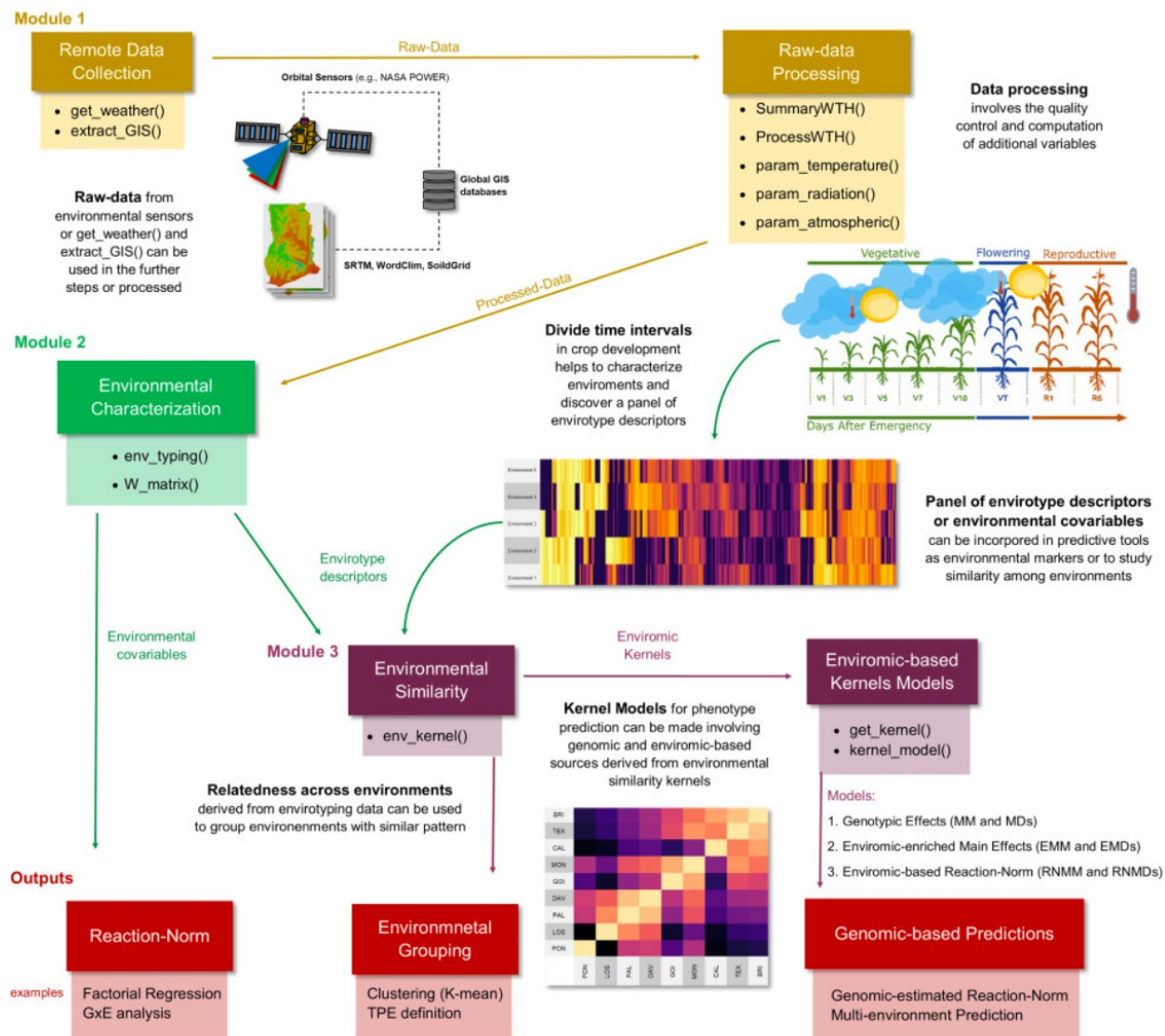
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SPECIAL SECTION: MACHINE LEARNING IN AGRICULTURE

Agronomy Journal

SoilType: An R package to interplay soil characterization in plant science

Roberto Fritsche-Neto 



Using enviromics, we can:

- **Study reaction norms – how a genotype reacts to changes in an environmental component gradient**
- **Brings more resolution to GxE studies**
- **It may reduce the total number of trials and cost - better allocate resources**
- **Define the optimal MET in advance**
- **Identify genomic regions associated with EC responsiveness**
- **Help to develop models to select more resilient genotypes for future scenarios**
- **Develop models for epidemiology – predict the disease progress**
- **The limit is your imagination...**