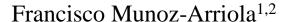


Pioneering new frontiers.



"Plugin"-based architecture of software to predict corn phenotypes



Diego Jarquin³

Hallie Hohbein⁴

Parisa Sarzaeim⁴

Joseph Carter⁴

David Recic⁴

Zoe Trautman⁴

Anna Zhang⁴

Byrav Ramamurthy⁴





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² School of Natural Resources

³Department of Agronomy and Horticulture

⁴Department of Computer Sciences and Engineering

Acknowledgements





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Some ideas are associated with the USDA National Institute of Food and Agriculture, Agriculture and Food Research Initiative HATCH project NEB-21-166 Accession No. No.1009760

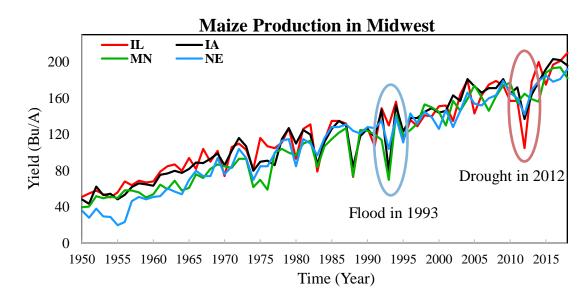
Genomes to Fields initiative

UNL's Department of Computer Sciences and Engineering Senior Design

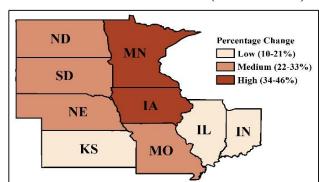
Motivation

 Consistent increase of water use efficiency, farmers revenues and yields

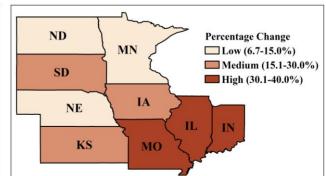
• Drops in water use efficiency, farmers revenues and yields after the occurrence of floods and droughts



Maize Yield Reduction (1992 -1993)



Maize Yield Reduction (2010 -2012)

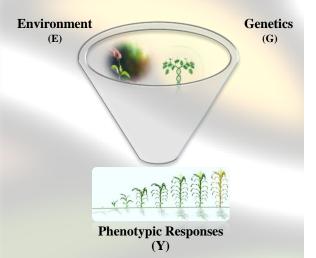


Ref: USDA NASS

Outline

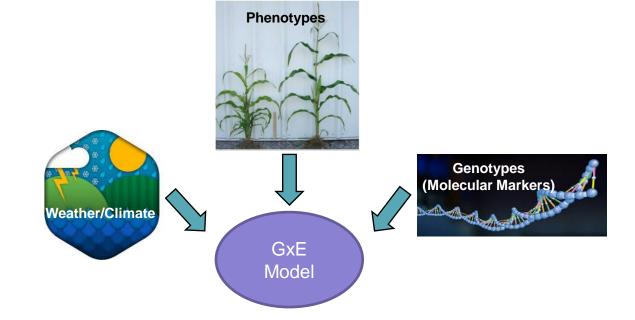
- Framework
- **G2F**
- Software Architecture
 - Preprocessing
 - Option Selection
 - Processing
 - Postprocessing
- Software Demo
- Complexities
- Conclusion
- Future Work

Framework

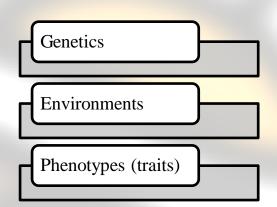




Develop a framework to collect, store, manage, and use weather/climate data to predict plant phenotypes using GxE model

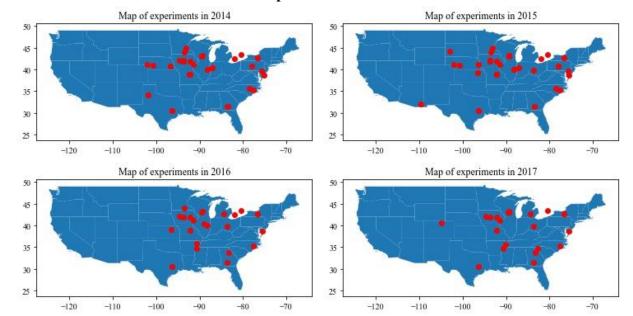


G2F



Incorporation of Environmental Information to Improve Phenotypic Predictability in Maize G2F-GxE Hybrid Project

G2F Experiments Distribution



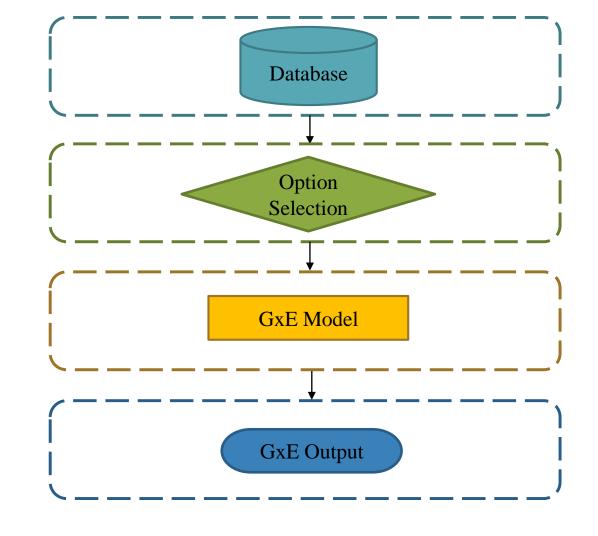
Software Architecture

Preprocessing

Selection

Processing

Postprocessing



Database



Temperature



Dew Point



Relative Humidity



Solar Radiation



Rainfall



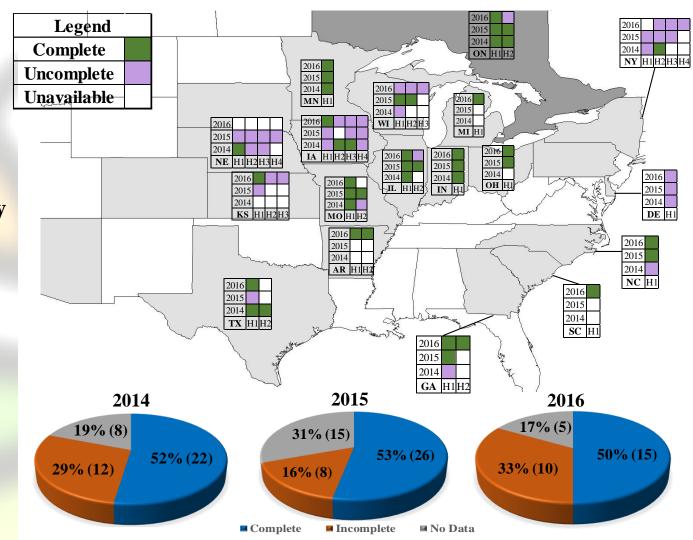
Mind Speed



Wind Direction

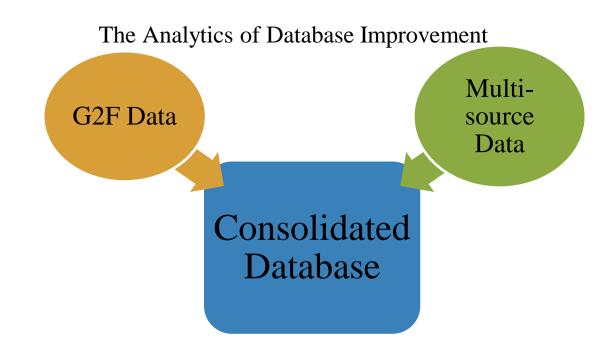


Wind Gust



Pre-processing

- 1. Integration of various data sources
- 2. Correction the data
- 3. Synthesis the data



Data consolidation



Temperature



Dew Point



Relative Humidity



Solar Radiation



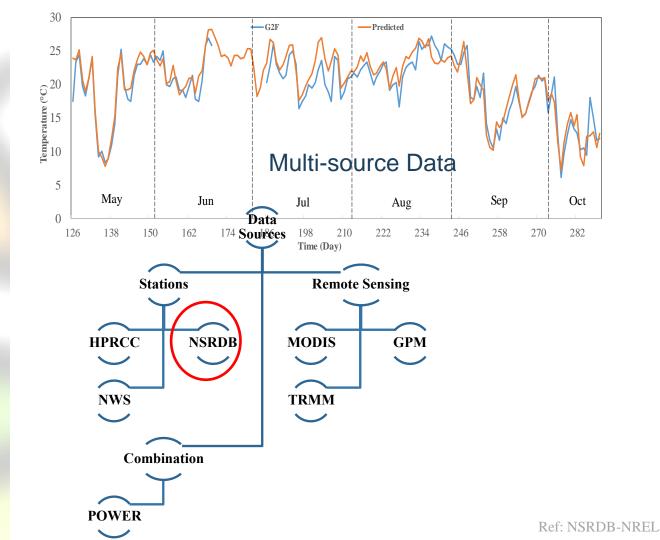
Rainfall



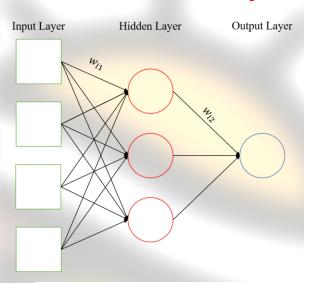
Wind Speed

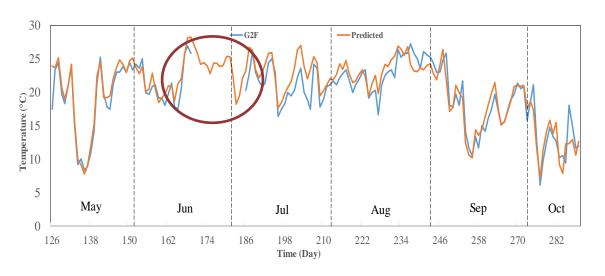


Wind Direction



Data-driven analytics





Performance Metrics

- R²
- Bias
- RMSE
- NSE

Performance Metric	Mean	Min	Max	SD
R ²	0.88	0.61	0.96	0.10
Bias	-0.52	-1.15	0.13	0.37
RMSE	1.67	1.13	3.00	0.55
NSE	0.87	0.80	0.98	0.05

Input Files

Files Control

Data Processing

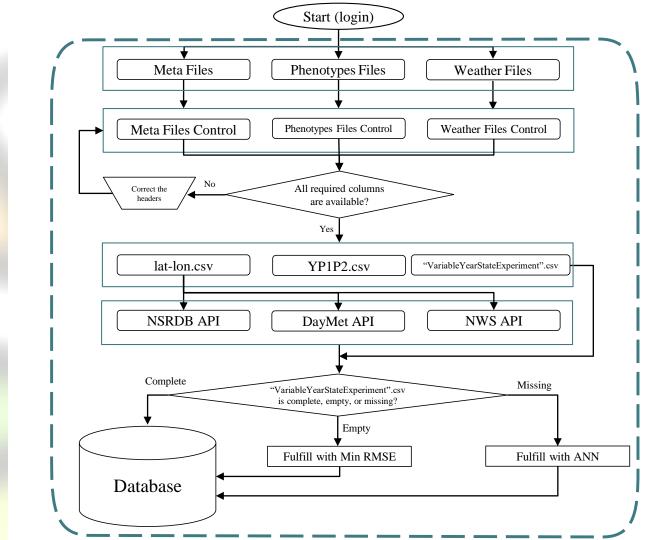
APIs

Database

Option Selection

GxE Model

GxE Output



Preprocessing

Pre-Processing

Input Files

Files Control

Data Processing

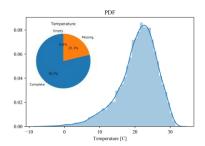
APIs

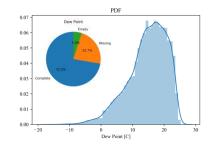
Database

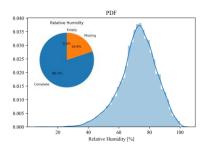
Option Selection

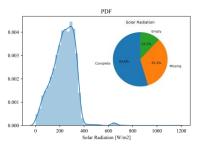
GxE Model

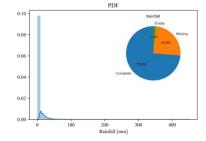
GxE Outpu

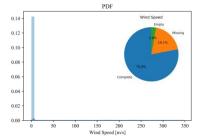


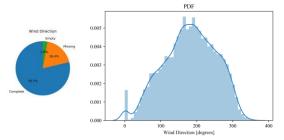


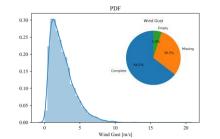








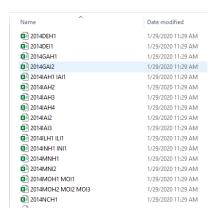




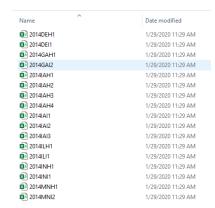
Complete Empty Missing

Pre-processing

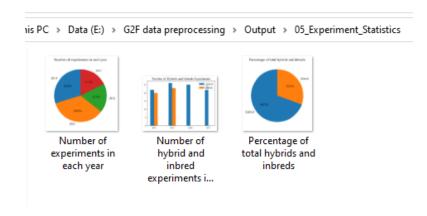
Data separation for each experiment



Correction of the Experiment names and Check the sequence of days



Charts for experiments analysis



Pre-processing

Separating data for each variable

Name Date modified **☑** D2014DEH1 1/29/2020 11:31 AM D2014DEI1 1/29/2020 11:31 AM D2014GAH1 1/29/2020 11:31 AM D2014GAI2 1/29/2020 11:31 AM **図** D2014IAH1 1/29/2020 11:31 AM D2014IAH2 1/29/2020 11:31 AM D2014IAH3 1/29/2020 11:31 AM D2014IAH4 1/29/2020 11:31 AM D2014IAI1 1/29/2020 11:31 AM D2014IAI2 1/29/2020 11:31 AM D2014IAI3 1/29/2020 11:31 AM D2014ILH1 1/29/2020 11:31 AM □ D2014ILI1 1/29/2020 11:31 AM D2014INH1 1/29/2020 11:31 AM D2014INI1 1/29/2020 11:31 AM D2014MNH1 1/29/2020 11:31 AM D2014MNI2 1/29/2020 11:31 AM **₽** D2014MOH1 1/29/2020 11:31 AM D2014MOH2 1/29/2020 11:31 AM D2014MOI1 1/29/2020 11:31 AM D2014MOI2 1/29/2020 11:31 AM **□** D2014MOI3 1/29/2020 11:31 AM D2014NCH1 1/29/2020 11:31 AM

Providing PDFs for each variable

200

400

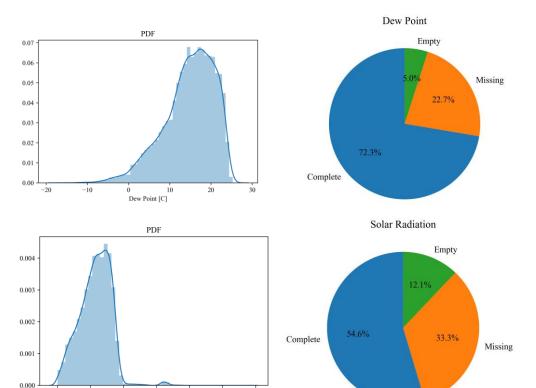
600

Solar Radiation [W/m2]

800

1000

1200



Providing charts to analyze data

availability for each variable

Input Files

Files Control

Data Processing

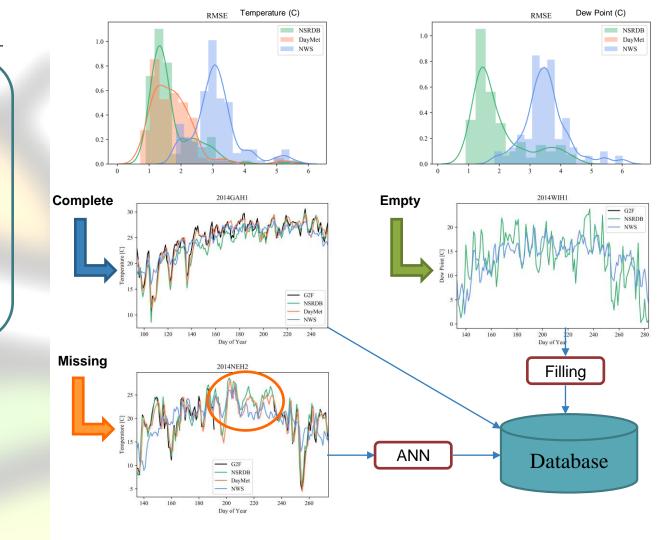
APIs

Database

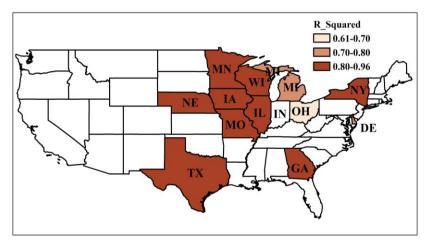
Option Selection

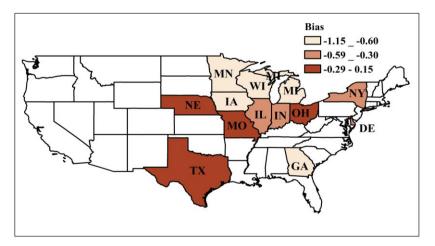
GxE Model

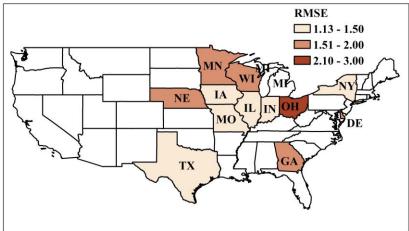
GxE Outpu

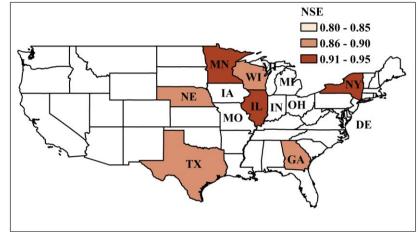


Performance Metrics









Selection

Preprocessing

Select Variables

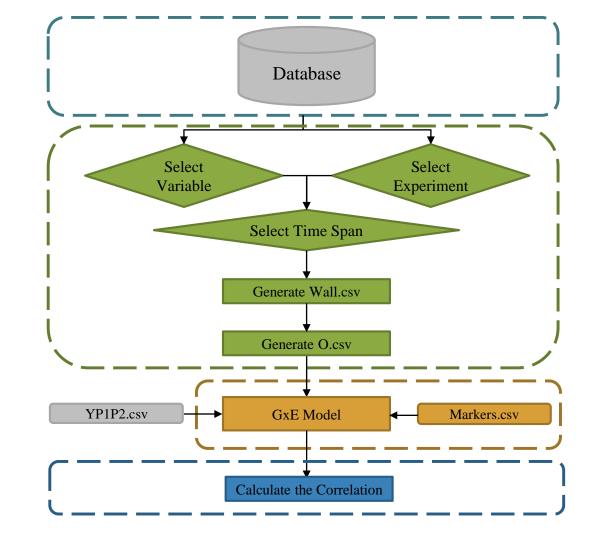
Select Experiment

Option Selection

Select Time Span

GxE Model

GxE Output



Selection

Preprocessing

Select Variables

Selection

Option

GxE Execution Select Experiment

Select Time Span

GxE Model

GxE Output

Select Variable(s) ▼

- Temperature (C)
- Dew Point (C)
- Relative Humidity (%)
- Solar Radiation (W/m2)
- Rainfall (mm)
- Wind Speed (m/s)
- Wind Direction (degrees)
- Pressure (mb)
- Precipitable water (mm)

Select Experiment(s) ▼

- 2014IAH3
- 2015NEH3
- □ 2017MOH1

-

.

•

Start day =

		Tested Genotypes	
		YES	NO
Tested Environments	YES	CV2	CV1
	NO	CV0	CV00

CV00: Predicting performance of unobserved lines in unobserved environments;

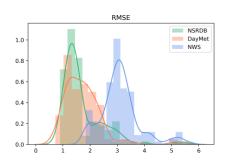
CV0: Predicting performance of unobserved environments;

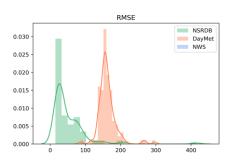
CV1: Predicting performance of new developed lines through relationships with others;

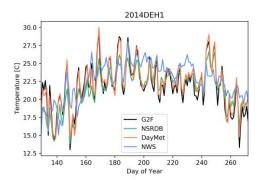
CV2: Predicting Performance of Lines Captured in

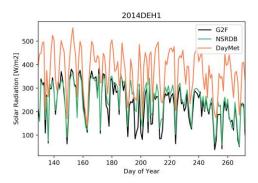
Other Environments

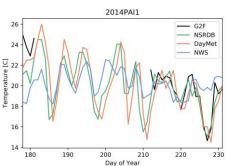
Selection

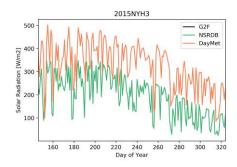












Post-Processing

Preprocessing

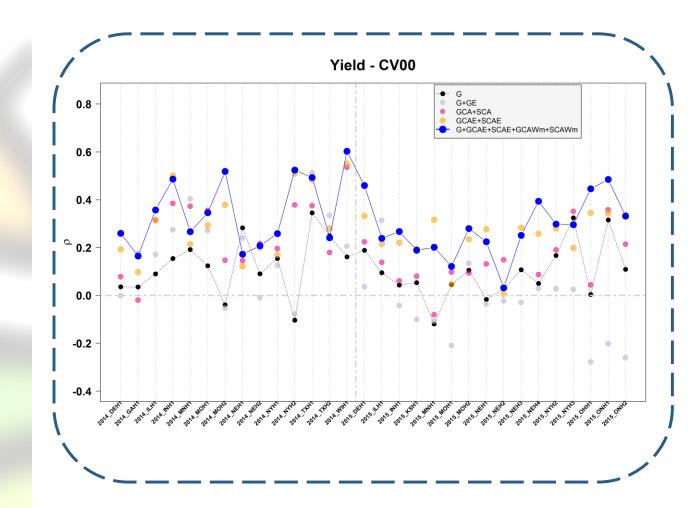
Select Variables

Select Experiment

Select Time Span

GxE Model

GxE Output



GxE Predictability

Post-Processing

Preprocessing

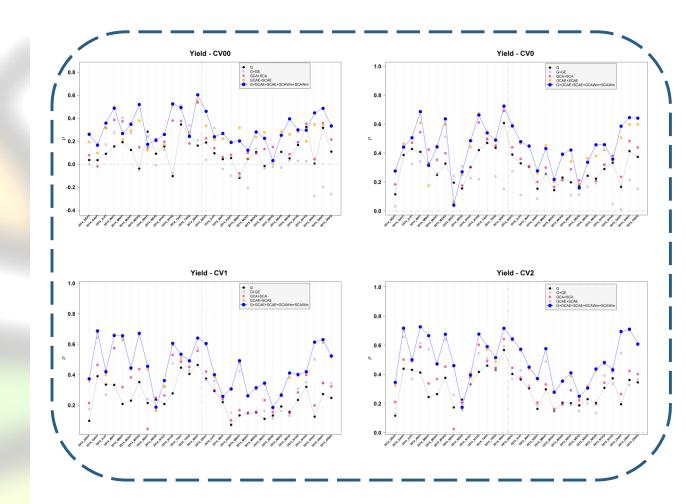
Select Variables

Select Experiment

Select Time Span

GxE Mode

GxE Output



GxE Predictability

A 80 A A 11 Car

Sign In

Email

Password

Forgot your password?

New User?

Sign In







(C) (085.24.03 (64-60)

Complexities



Providing AWS (Amazon Web Service) as platform for the phenotypic predictability application;



Authentication for different users;



Transferring all the data (G2F, NSRDB, DayMet, and NWS) and scripts (R and Python) to the platform;



Coupling R and Python scripts to develop an integrated software for phenotype-prediction in the G2F experiment.

Conclusions

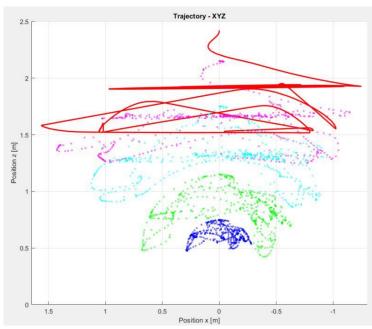
- The integration of other data sources to improve G2F database unclearly improved the predictability of phenotypes;
- Transferring and coupling the hydroclimate data analytics and GxE modeling scripts to the web service platform is feasible;
- Increasing the number of experiments may lead to a better accuracy of phenotype predictability.

Future work

- Add climatic spatial and temporal analytics of GxE predictability module;
- Add a global sensitivity of GxE accuracy module to estimate sources and propagation of uncertainty in response to various climatic (environmental) factors;
- Add the remote sensing data plugin module to increase the number of climatic variables and phenotypes in the database.

Some more future work







Team members and tasks:

- Francisco Munoz-Arriola; Team leader
- Diego Jarquin: GxE model developer; Develops R scripts for phenotypes predictions using GxE
- Hallie Hohbein: Project Manager; Takes care of project management tasks, documentation, and testing
- Parisa Sarzaeim: Hydroclimate data scientist; Develops Python scripts to manage hydroclimate database
- **Joseph Carter: Frontend/Backend Developer**; Works on user authentication, frontend development, and testing.
- **David Recic: Backed Developer**; Creates the database and works on user authentication.
- Zoe Trautman: Frontend Developer; Develops the frontend and writes documentation.
- Anna Zhang: Development Manager; In charge of AWS and helps with backend development.
- Byrav Ramamurthy and Francisco Munoz-Arriola; Computer science advisers

Thank You

This project was supported by the Agriculture and Food Research Initiative Grant number NEB-21-176 and NEB-21-166 from the USDA National Institute of Food and Agriculture, Plant Health and Production and Plant Products: Plant Breeding for Agricultural Production, A1211).

Accession Nos.1015252 and No.1009760

