



ISMB 2022 Panel: Digital Agriculture at Scale

Addie Thompson
Michigan State University
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Background

- Iowa State University
 - *B.S. Genetics*
- University of Minnesota
 - *Ph.D. Plant Breeding and Molecular Genetics*
- Purdue University
 - *Postdoc in maize/sorghum quantitative genetics and phenomics*



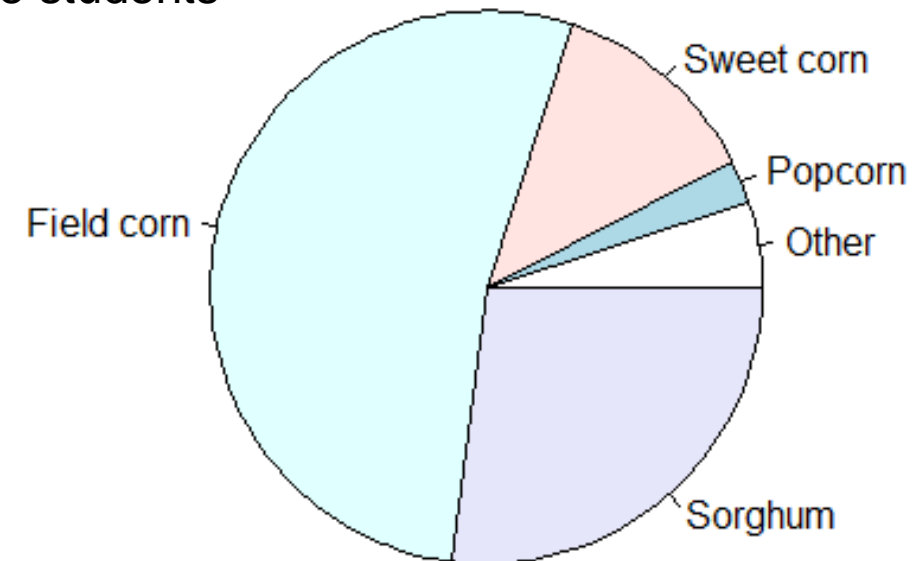
Current Appointment – Michigan State

Research

- **Genetic mapping and genomic prediction** for both basic (genetic mechanisms) and applied (plant breeding) outcomes;
- **High-throughput phenotyping** to increase efficiency and throughput of trait data collection; and
- **Innovation in modeling and prediction** approaches to maximize the impact of data and information.

Teaching

- Transdisciplinary skillsets and communication
- Hands-on, project-based, experiential learning
- Real-world job prep for graduate students





Current projects in maize and sorghum



Ruijuan Tan



Anuradha Singh



Zhongjie Ji

...painstakingly
measuring
thousands of
plants

We get great
datasets...but it is
SO SLOW

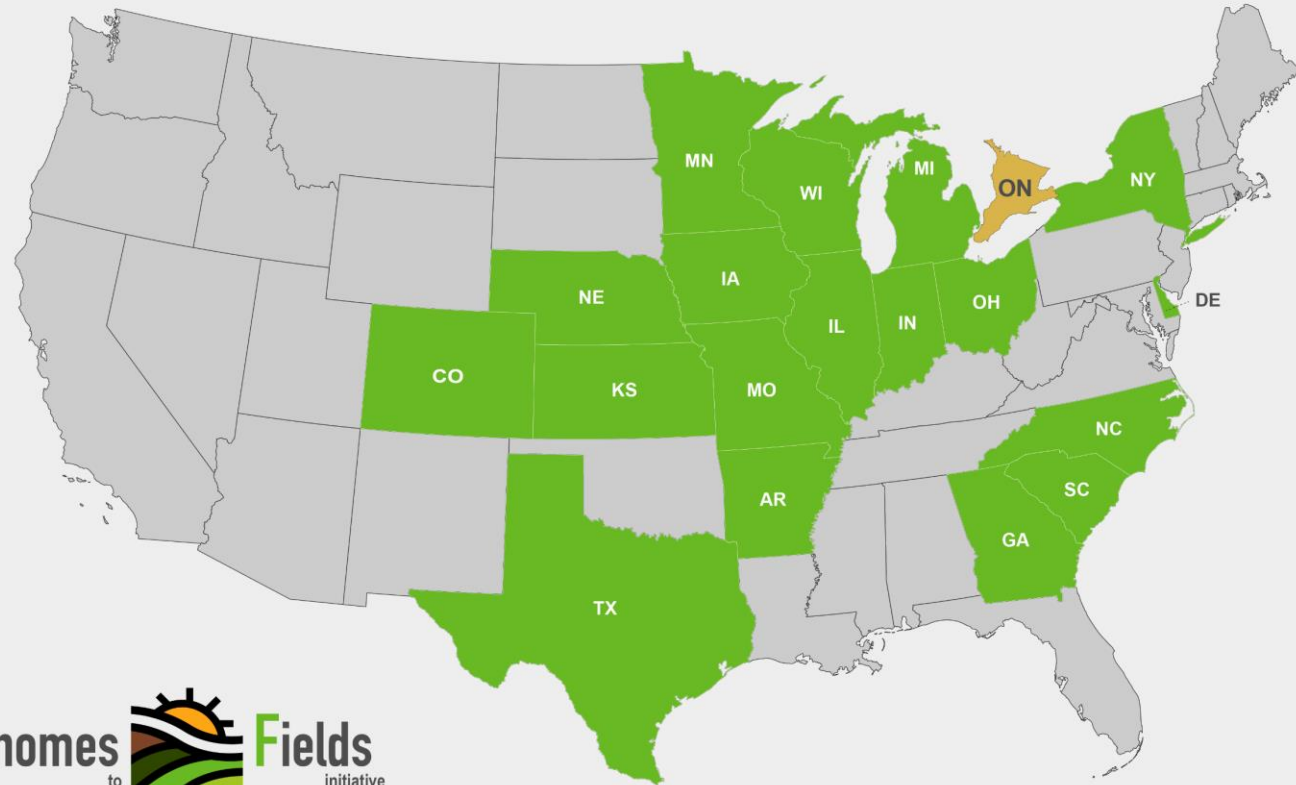
2019 Academic & Federal Institutions

Arkansas State University
Clemson University
Colorado State University
Cornell University
Iowa State University
Kansas State University

Michigan State University
North Carolina State University
Ohio State University
Purdue University
Texas A&M University
University of Delaware

University of Georgia
University of Guelph
University of Illinois
University of Minnesota
University of Missouri
University of Nebraska

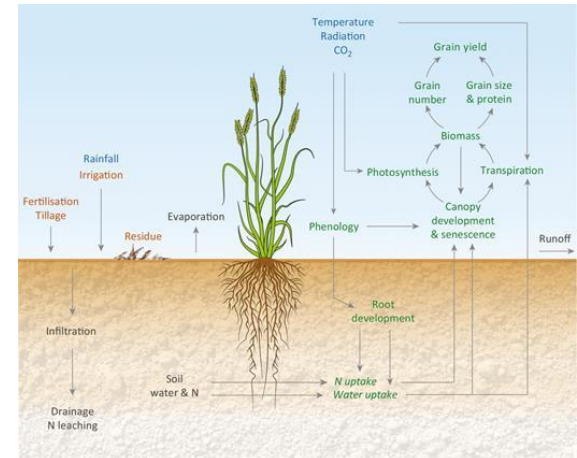
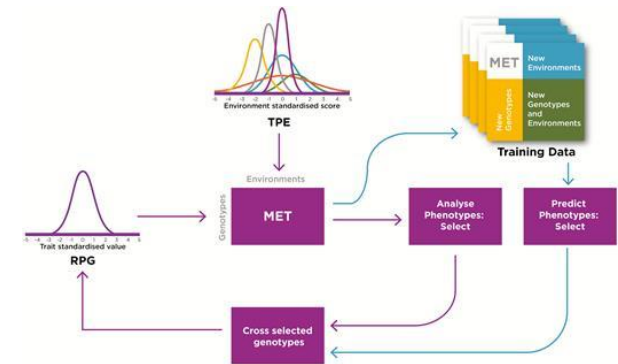
University of Wisconsin
USDA-ARS
Georg-August-Universität Göttingen
(Not shown on map)



2019 GxE Principal Investigators and Academic & Federal Institutions in the United States and Canada

Our Big Question

- From a plant breeding and genetics perspective, we want to predict how a variety will perform to assess its usefulness.
- We can use genetics to predict phenotype (Genomic Prediction), but this does not perform well in new environments.
- We can use physiological modeling (Crop Growth Models) to simulate varieties in different environments, but this relies on many [tedious] hand-measured phenotypes to parameterize the models.
- ***Can we acquire those phenotypes some other way?***



Sensor Technology

Measured Traits

Modeled Traits

Composite Traits

LiDAR

Stalk diameter

Lodging

Height distribution and canopy structure

Tillering propensity

Leaf angle

Leaf number and size

Plant count and spacing

Total biomass

Reconstructed RGB

Flowering time

Ear size and number

Canopy cover

Grain color

Yield

Biochemical composition

Vegetative indices for plant vigor and stress

Hyperspectral Imaging

Nitrogen content

Water content

Drought tolerance

Thermal Imaging

Leaf temperature

Stomatal conductance

Our Team (2019)

Plant, Soil & Microbial Sciences



Thompson



Chilvers



Olson



Douches



Weebade



Cichy

USDA-ARS

Horticulture



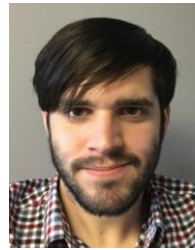
Brainard



Edger



Hayden



VanBuren



Hollender

Biochemistry and Mol. Bio.



Grotewold

Computational Math, Sci, and Eng



Xie

MSU IT



Lim



de los Campos



Shiu

Geography, Environ., & Spatial Sci.



Bunting

Statistics and Probability

Electrical and Computer Eng.



Biswas



Morris

Comp Sci & Eng



Liu



Ross



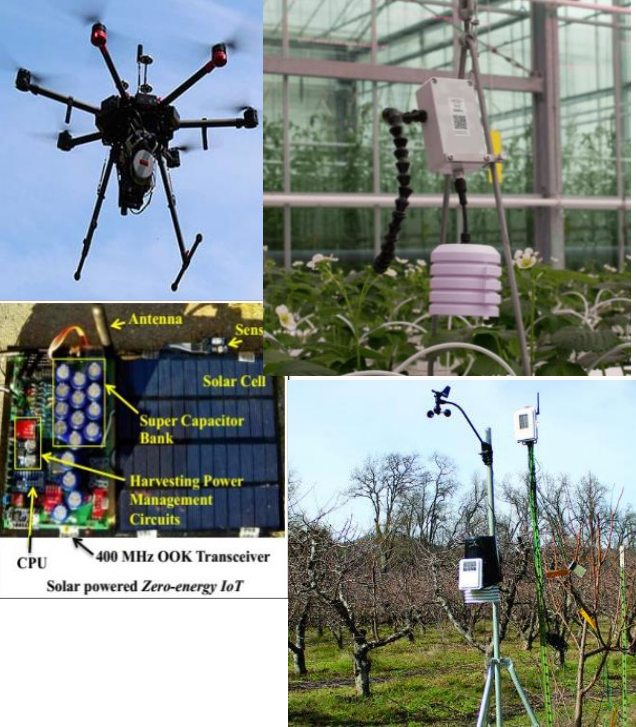
Tong



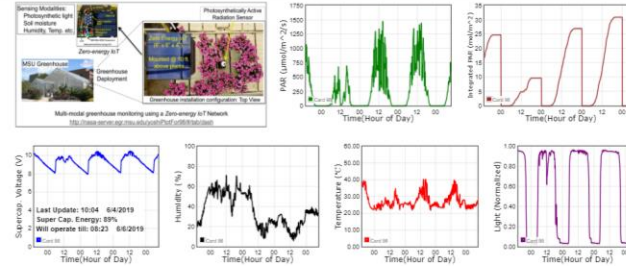
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Our Vision (2019)

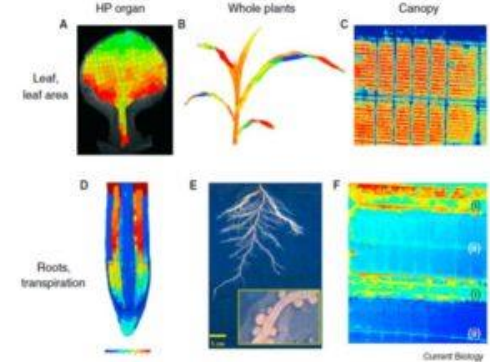
Multi-scale platforms



Network protocols for data collection



Big data, predictive analytics, machine learning and AI



Commercial sensors

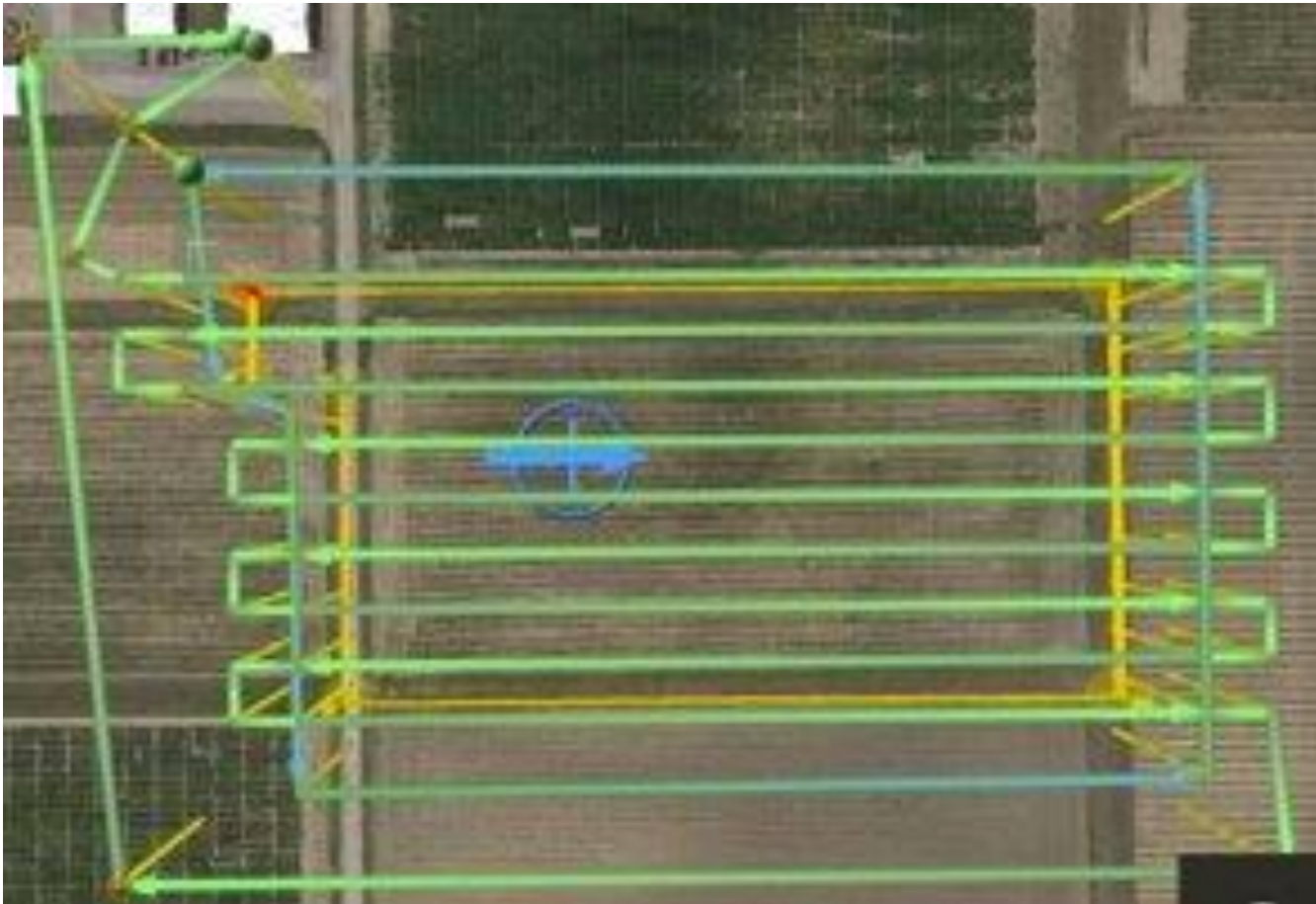


Enhanced outcomes for plant science and breeding

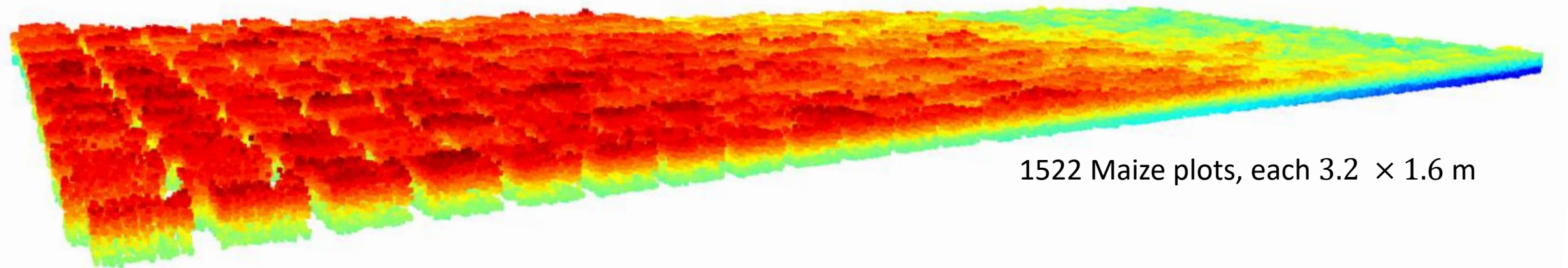
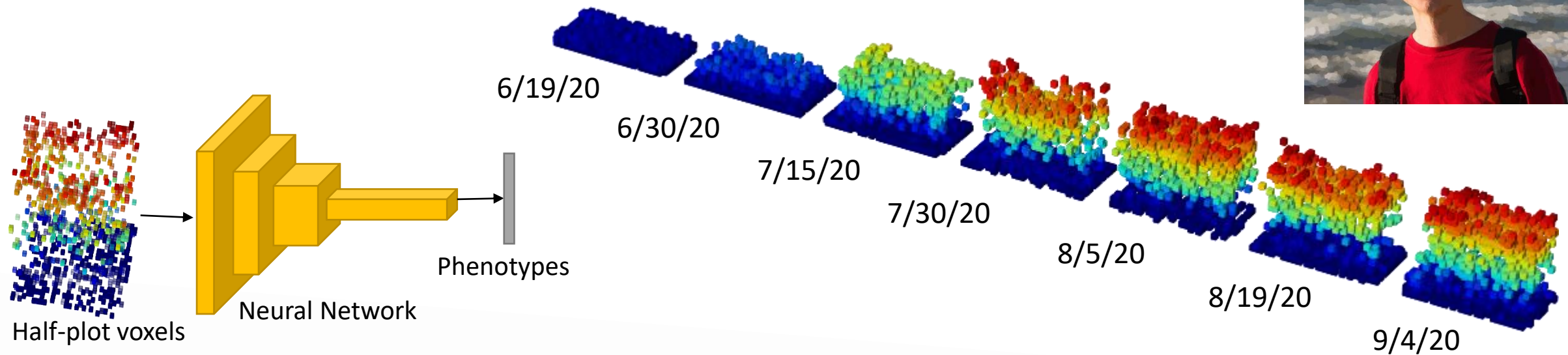


UAV-Measured Data

- LiDAR, VNIR-SWIR, RGB, MultiSpec

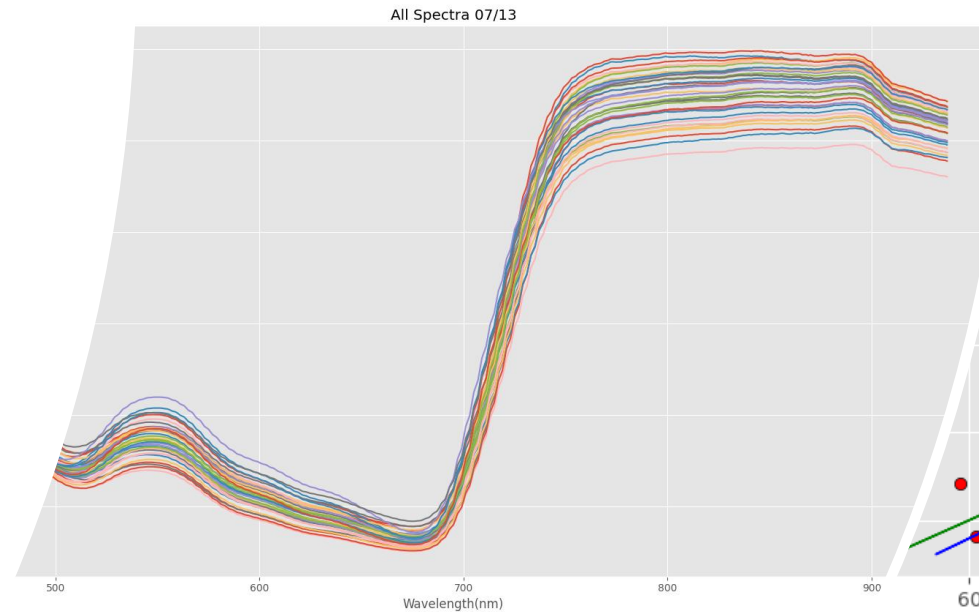
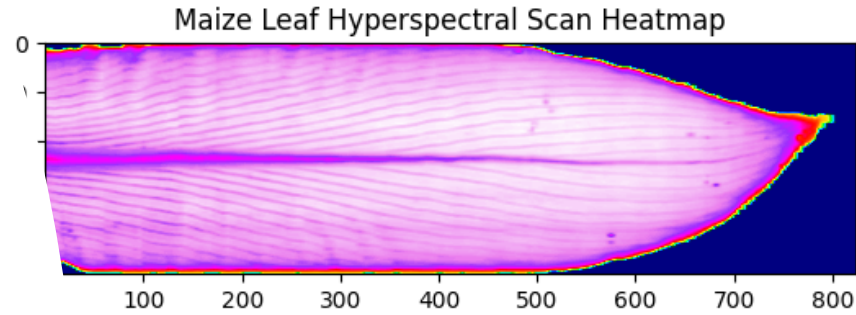


Using LiDAR data to predict plant traits

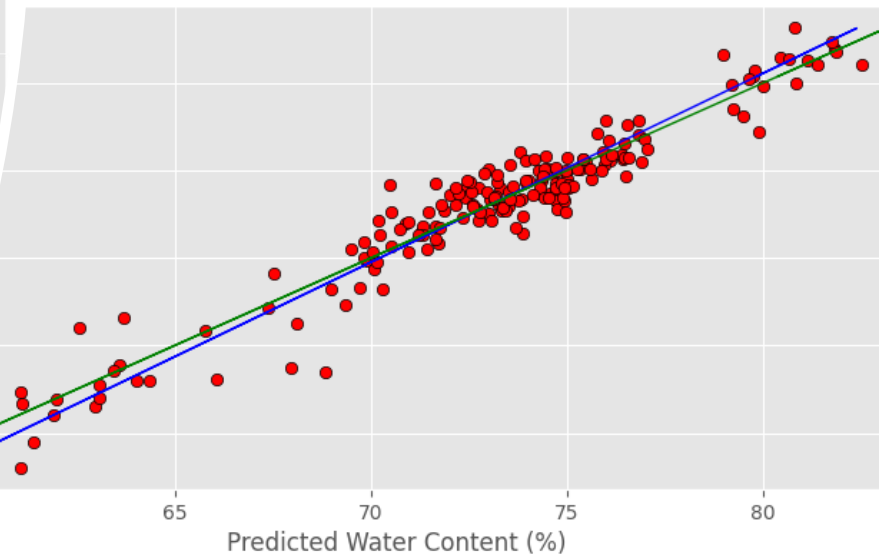


Training computers to “see” plant physiology

- Reflected and transmitted light changes based on biochemical properties
- Machine learning techniques to train and predict important traits



RPD: 2.6226579146137023
 R^2 (CV): 0.853838514517961
Num Components: 17
Predicted Water Content - 07/28 removed



Improving detection of onset and severity of tar spot disease in maize

- Collect data throughout the season and relate back to disease onset and severity
- Can we identify disease before it is abundantly visible in the field?
 - Enables more management options



Other Research Projects

- Multi-objective genomic mating optimization for breeding programs
- Physiological impacts of N and biostimulant
- Functional genomics and modeling in sorghum
- Assessment and prediction of flavonoid content in maize kernels and leaves



Collaboration

Communication

Creativity



“The world needs different kinds of minds to work together. When different kinds of minds work together effectively, there can be great successes. They complement each other’s skills.” – Temple Grandin

- Find common language with collaborators – this can take time

College of Natural Science

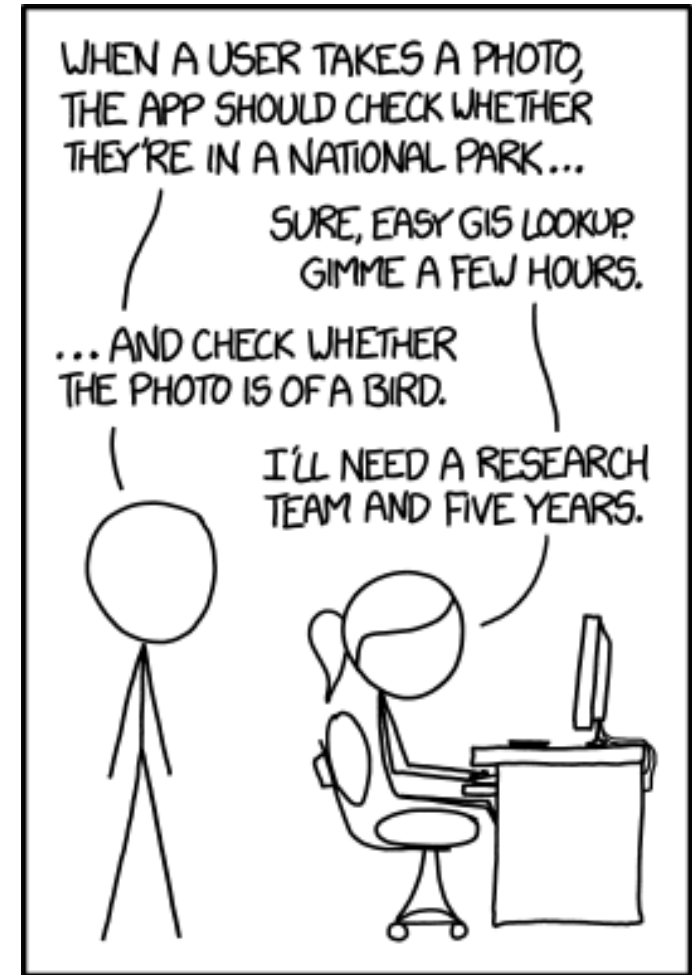
NRT-IMPACTS

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IMPACTS

Interfacing Computational & Plant Sciences

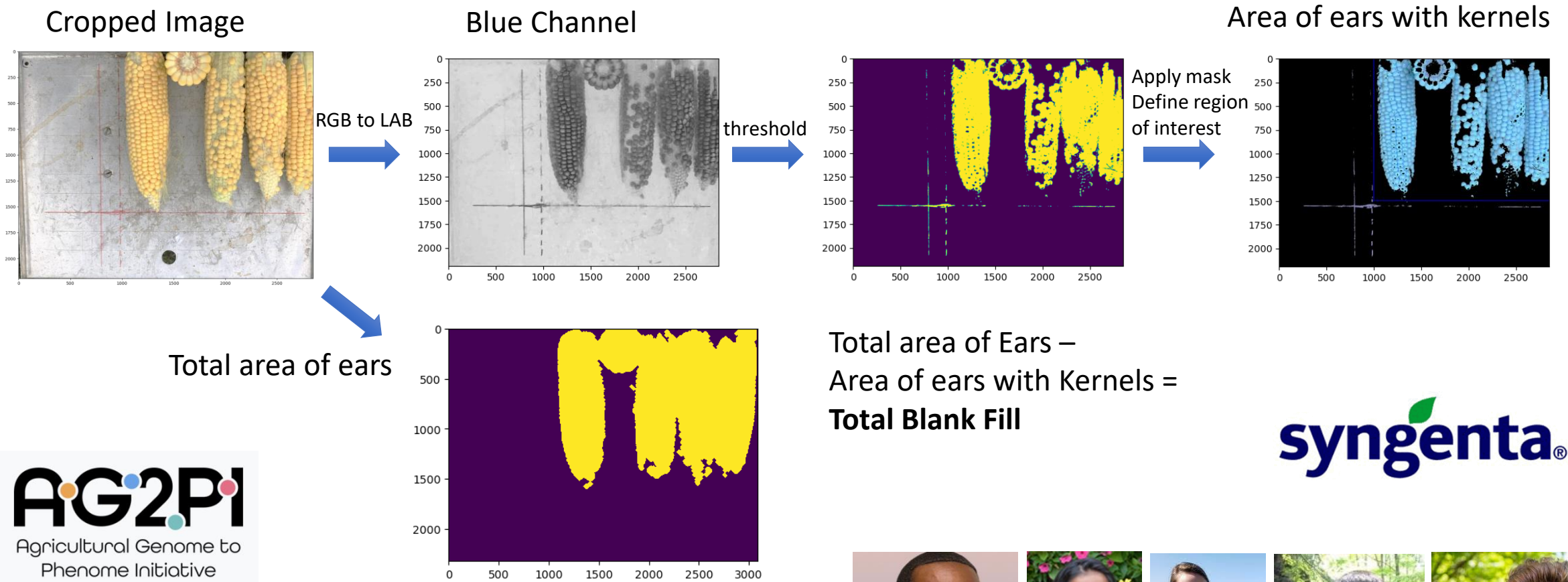
CSS 844:
Project-oriented
Team-based
Interdisciplinary
Communication skills



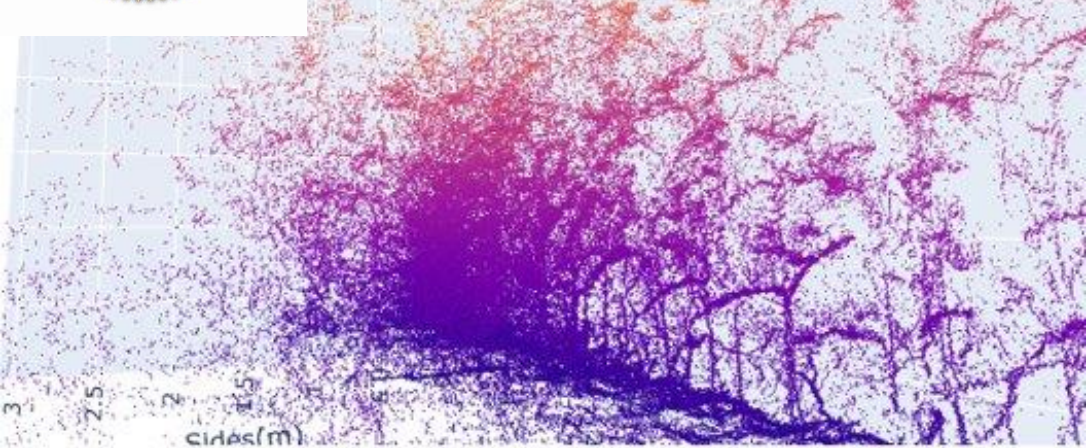
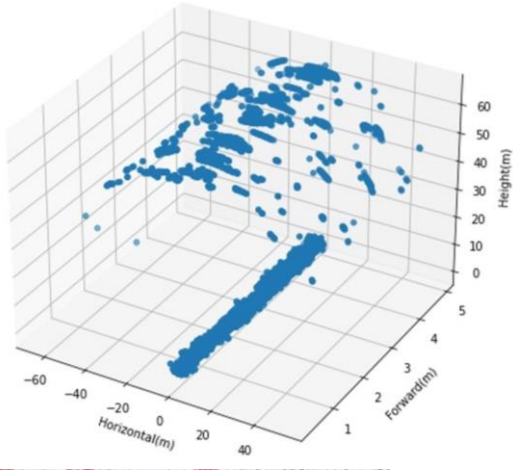
IN CS, IT CAN BE HARD TO EXPLAIN
THE DIFFERENCE BETWEEN THE EASY
AND THE VIRTUALLY IMPOSSIBLE.

From xkcd

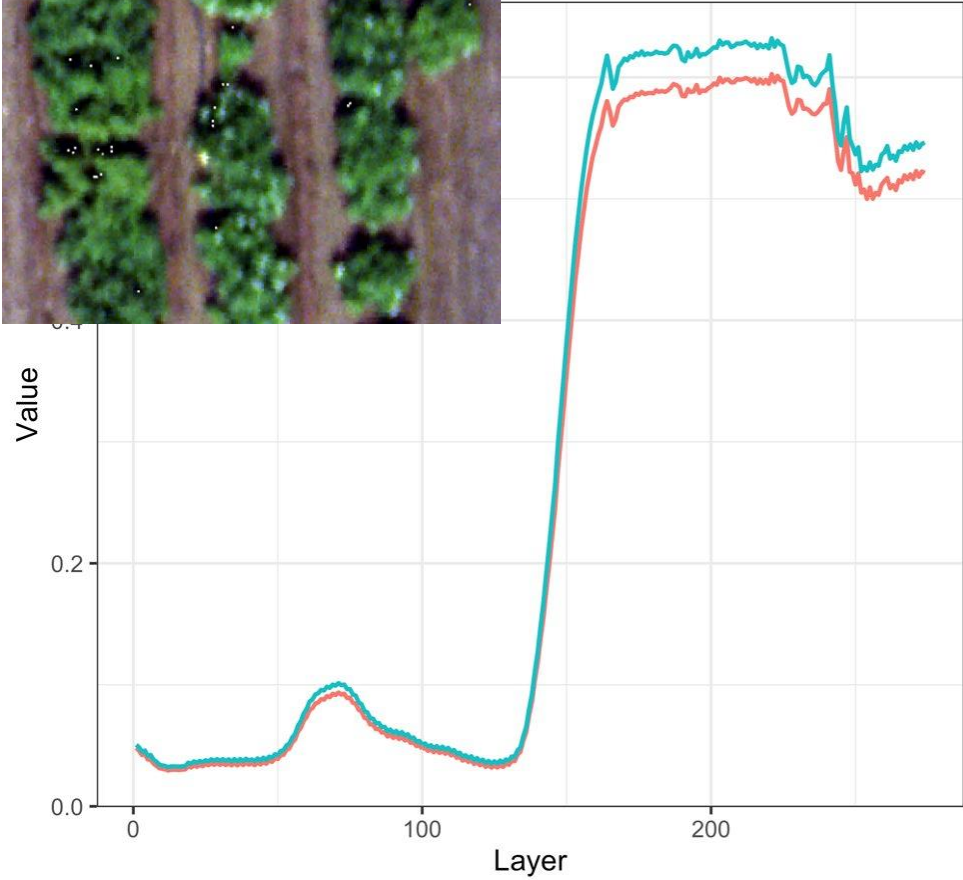
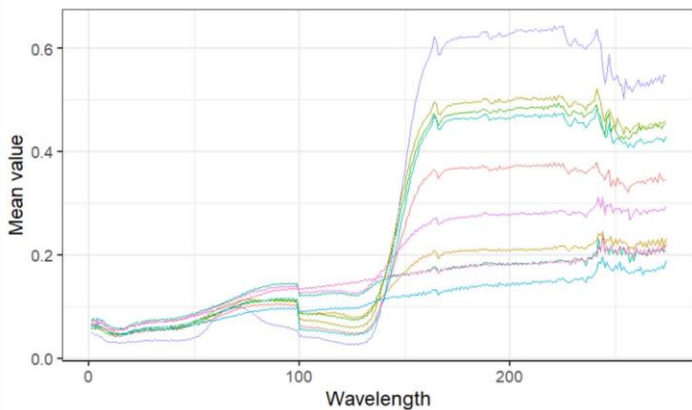
CSS 844: Frontiers in Computational and Plant Sciences – Spring 2021 Module 2



2022: Module 2



Asia Hightower, Brandon Webster, Joanne Thomson, & Nick Johnson



Kara Dobson, Sidney Sitar, MacKenzie Jacobs, Claudia Miranda

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2021 GRANTS4AG AWARDEES

Bayer Crop Science



MICHIGAN STATE UNIVERSITY

Plant Resilience Institute



Strategic Partnership Grant
Research and Innovation
MICHIGAN STATE UNIVERSITY

MICHIGAN STATE UNIVERSITY

AgBioResearch

AG2PI

Agricultural Genome to
Phenome Initiative



U.S. DEPARTMENT OF
ENERGY

Office of
Science



Maximum Farming
powered by Ag Spectrum



IOWA CORN

the

Genomes

to



Fields

initiative

Thank You

