

IMPROVING DECISION SUPPORT SYSTEMS WITH MACHINE LEARNING:

IDENTIFYING BARRIERS TO ADOPTION

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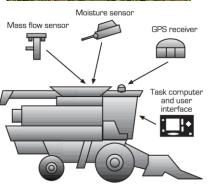
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Precision Agriculture Solves Many Challenges

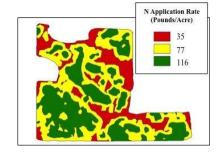
UAVs can characterize a field to identify crop health stressors before yield/economic loss occurs

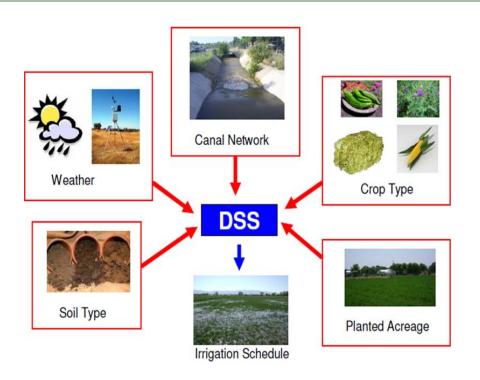
The integration of sensors into harvesters provides precise yield data in which to make well-informed management decisions and yield goals





The use of Variable Rate Fertilizer Technologies reduce cost of input and increase resource use efficiency

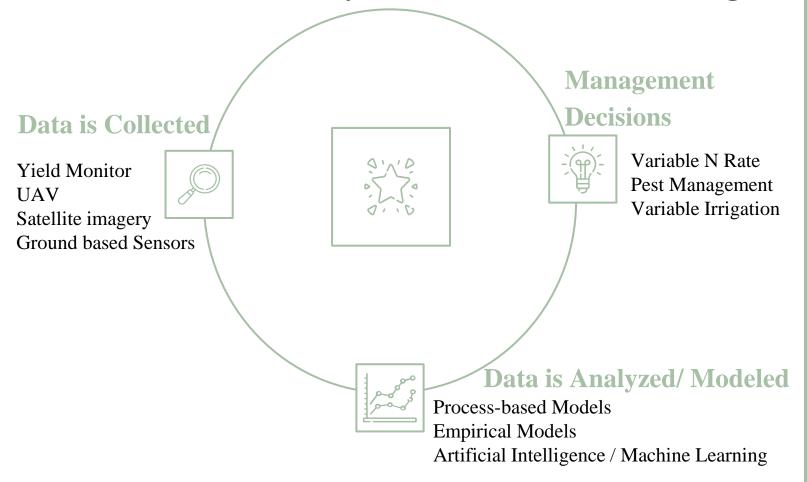


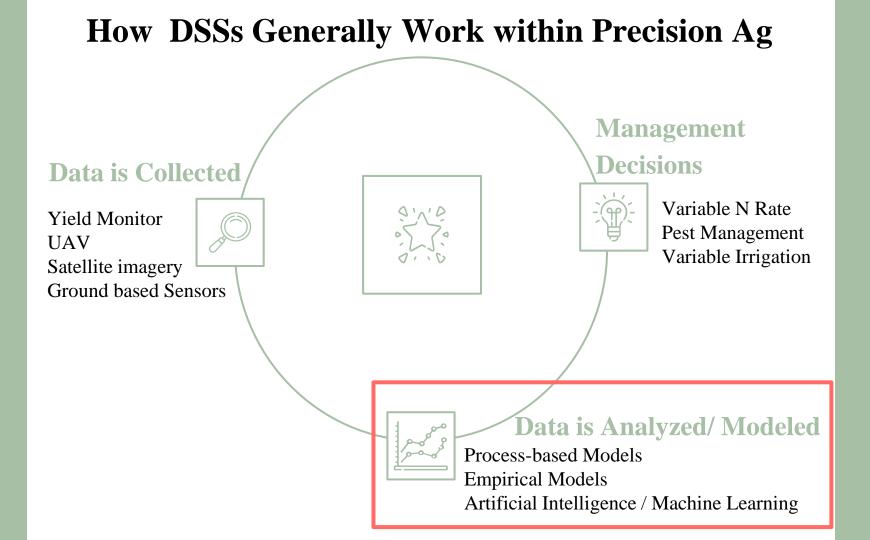


Decision Support Systems

Integrates collected data with modeling to produce actionable decisions

How DSSs Generally Work within Precision Ag





Modeling and Data Analysis to Make Decisions

Process - Based

- Chemical, physical, or biological process
- (Yield Goal x 1)-N credits= N Rate

Empirical Model

Statistical model based on observations among variables



Welcome to the

Soil test for Nitrogen analysis (lbs/acre 2-ft depth)

North Dakota Wheat Nitrogen Calculator

You will need to know the location of the farm, the general productivity of the soils, the price you contract for wheat, the cost per pound of N, the soil test nitrate-N to a depth of 2-feet, and the previous crop.

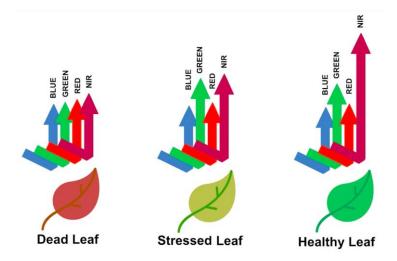
Please select the location of the farm. The map of North Dakota on this site will help you determine the region of the farm. Click on the map for a detailed view.

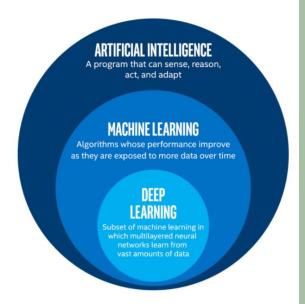


Low Productivity Medium Productivity High Productivity	No Nitrogen-supplying crop Styleam, Field Pea, Dry Bean, Lertil, Chickpea, or harvested Sweet Poa Sugarbeet with yellow-green leaves Sugarbeet with green leaves Harvested Affalf or unharvested Sweet Clover (1.5 plantalsq-ft) Harvested Affalf or unharvested Sweet Clover (2.4 plantalsq-ft)
Select Nearest Wheat Price \$3.00	Harvested Alfalfa or unharvested Sweet Clover (1-2 plants/sq-ft) Harvested Alfalfa or unharvested Sweet Clover (+1 plants/sq-ft)
Select Nearest N Cost (\$0.20	Nitrogen provided by previous crops:
Nitrogen Recommendation Before Credits: 75	Please indicate the previous tilling method used in the field.
Please indicate the amount of nitrates in the soil. (Enter the analysis result in the box.)	Is the field/area conventionally tilled (chisel and/or disc, and/or field cultivator or plow)? Has the field/area been in No-till for 1 to 5 years?

Machine Learning

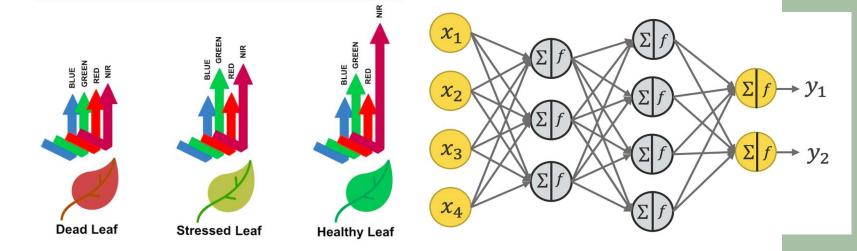
- AI vs ML vs DL
- ML the model will identify new and complex patterns to update itself without human intervention (no feature selection required)
- Example: Vegetative indices to predict yield



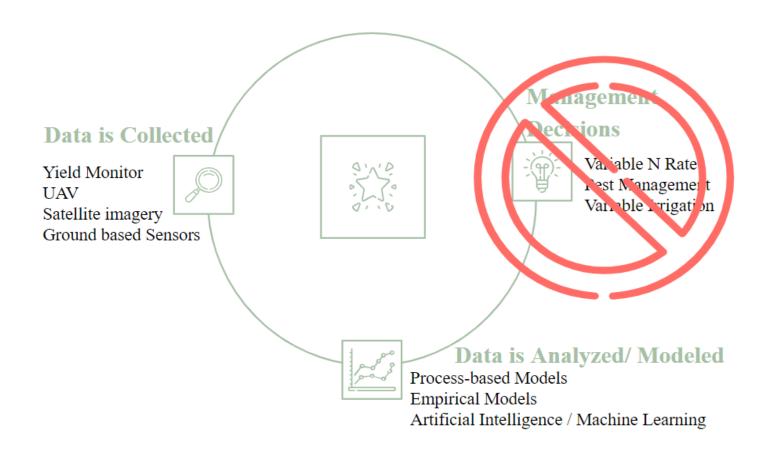


Machine Learning

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Barriers to Adoption of these Management Tools



Barriers to Adoption of these Management Tools

Some Precision Ag Technologies are more prevalent than others suggesting there are barriers to the adoption of some of these tools



More Yield Monitors



Less Variable Rate Technology

Barriers Identified Through Survey

Agricultural retailers have been surveyed about producer attitudes every other year since the mid-1990s to give insights (Erickson & Lowenberg-Deboer, 2022)

#1 Economic Limitations

#2 Trust Concerns

#3 Data Privacy

Workforce Availability



Bruce Erickson and James Lowenberg-DeBoer

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Departments of Agronomy and Agricultural Economics

Purdue University

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https://ag.purdue.edu/digitalag/_media/ croplife-report-2022.pdf

Financial Barrier



Financial Barriers

Crop prices have been observed to dictate how much of a barrier

Return on Investment

Smaller farms have a more difficult time making a return on investment which deters farmers from taking these economic chances

Ashworth et al., (2018) discussed that the cost of the equipment per hectare decreases with increasing farm size

Two Farmers Want to Use Variable Rate Application

Farmer 1:

Farm size: 100 ha



Farmer 2:

Farm size: 1000 ha



Two Farmers Want to Use Variable Rate Application

Mapping Software: \$3,000



Applicator: \$100,000





Total: \$103,000

Farmer 1:

Farm size: 100 ha

Cost /ha: \$1,030



Farmer 2:

Farm size: 1000 ha

Cost /ha: \$103



Two Farmers Want to Use Variable Rate Application

Mapping Software: \$3,000

SMS Basic
SMS Advanced
\$985/3:260 Yearly Maintenance Fee

\$2895/3:275 Yearly Maintenance Fee

Applicator: \$100,000





Total: \$103,000

Farmer 1:

Farm size: 100 ha

Cost /ha: \$1,030



Farmer 2:

Farm size: 1000 ha

Cost /ha: \$103



Considerations: Potential Solutions

Collecting data (yield, soil, remote sensing imagery, etc) will provide **no economic impact until a decision is made** that drives an increase in revenue.

- Studies have observed significant advantages in which PA does save 20-30% in input costs (Winstead & Fulton, 2010)
- A farmer with 100 ha may not use PA technology the same way as one with 1,000 ha.
- University extensions can guide you to the right farm technology.

Industry-Driven Solutions

 Industry paying farmers and creating a market for crops produced with PA technology

Trust Concerns



Farmer Opinions on UAVs (Miles, 2019)

ADOPTION OF UNMANNED AERIAL SYSTEMS BY FARMERS IN TEXAS

A Thesis

by

MISTY VIDRINE MILES

Submitted to the Office of Graduate and Professional Studies of Texas A&M University in partial fulfillment of the requirements for the degree of

MASTER OF SCIENCE

High-Tech Enthusiasts: Embrace UAVs for advanced data and innovation.

Purposeful/Selective Users: Value UAVs but use them strategically.

Conventional/Selective Skeptics: Prefer traditional methods, showing cautious adoption.

Key Takeaway:

Trust and Expectations: Varying levels of trust in technology influence adoption, driven by differing expectations of UAV benefits.

Sticking Points for DSS in Agriculture

#1 Model Explainability/Communication

#2 | Clear Decision-Making Responsibility

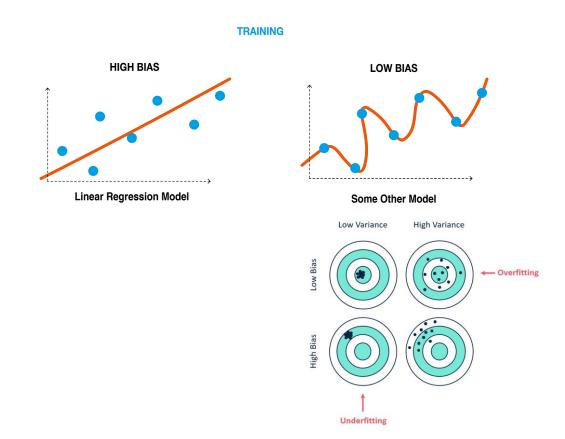
#3 | Fairness in Model Development

 $\begin{array}{c|c} x_1 & \Sigma f \\ \hline x_2 & \Sigma f \\ \hline x_3 & \Sigma f \\ \hline \end{array} \begin{array}{c} \Sigma f \\ \hline \Sigma f \\ \hline \end{array} \begin{array}{c} \Sigma f \\ \hline \end{array} \begin{array}{c} \Sigma f \\ \hline \end{array} \begin{array}{c} Y_1 \\ \hline \end{array}$

Deep Learning Model

Bias from Training Models with Large Farm Data

A lack of small farm data could create a data bias for future PA tools using ML and AI models



Examples of Cultivating Trust

- Building a tool that is not just easy to use but also **useful**
 - Vite.net: Vineyard Management using in field sensors
 - Identified specific issues for vineyards
 - Convey information easily and clearly
 - Pigs2Win: Swine Production DSS
 - Worked closely with farmers during design process, used excel, became familiar quickly





Data Ownership/ Privacy



Legal Issues with Data

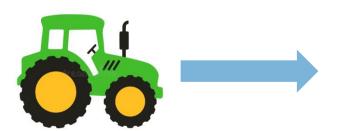
Bushel of corn is **tangible** and **excludable**



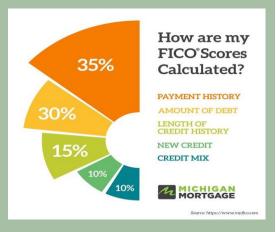
Farm Data:

Not Tangible

Not Excludable







- Data ownership is similar to a credit score
- Who is profiting and what are they doing with it?
- ML motivates third parties to acquire and aggregate data

Data Ownership

Open Source Data Sharing

Individual Contracts

Trade Secrets

End User License Agreements

Landlord vs Tenant

Ag Data Transparent Certified Companies

ambrook Leaf

https://www.agdatatransparent.com/

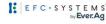
Leaf

Unified API for Agriculture

Ambrook, Inc. Ambrook















LogiAg

LogiAg System



Agi3 Ltd.

Agi3 Ltd.



Aleop Solution Inc.

Solution Aleop Inc.



AG Growth International Ag Growth International







BECK'S











Ecosystem Services Marke

Consortium

Eco-Harvest by ESMC

Combyne Ag

(COMBYNE





AgSights

bioTrack Plus









Ever.Ag Corporation My Dairy Dashboard



John Deere John Deere Operations Center



Simplot

Simplot Grower Solutions

prospera

A valmont V COMPANY





Agriculture

GrainFox









Conservis IIC







Prospera 365, Insights, Scheduling, AgSense, Variable Rate Irrigation

Palliser Insurance Co. Ltd. Palliser Direct Ltd.

FarmLink Marketing Solutions GrainFox

CarbonTerra

Global Ag Risk Solutions Global Ag Risk Solutions

Farm Management Software

GROWMARK FS Advanced Information Services

Traction

Traction Ag Farm Management Software

Farm Credit Canada Financement agricole Canada

WorkForce Availability



Knowledge, Skills, and Abilities in Precision Ag

- Long-term adoption requires skilled labor that is proficient in PA A 2015 survey (Erickson et al., 2018) Knowledge, skills, and abilities in the precision agriculture workforce
- Employees in PA meets or exceed education expectations

- However -

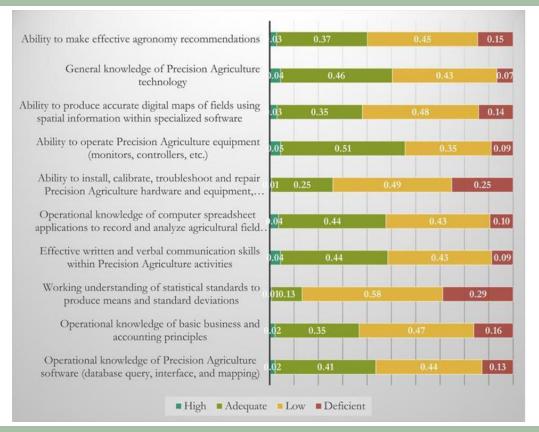
There is a shortage of employees capable of practical field applied skills:

Data-intensive thinking

Understanding Statistical Standards

Effectively design field trials for data

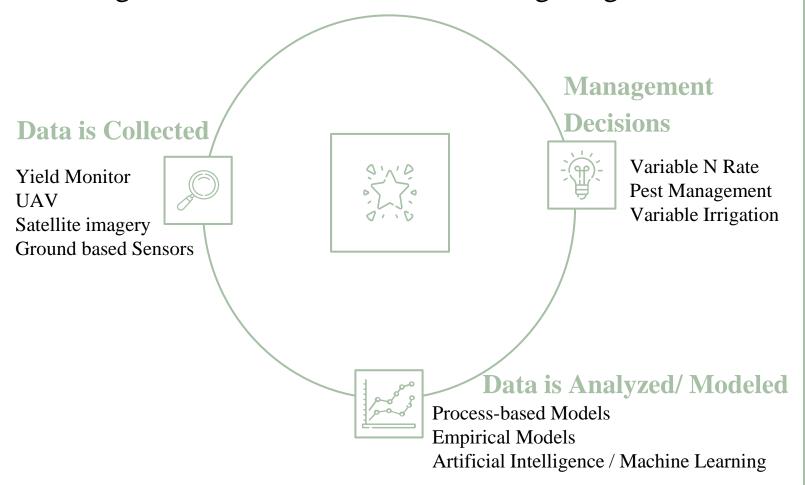
Knowledge, Skills, and Abilities in Precision Ag

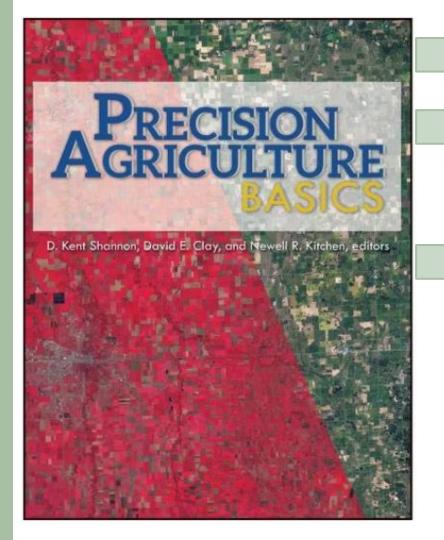


- Great Plains IDEA
 - Interactive Distance Education Alliance (IDEA) offers fully online programs and coursework
- Ag IDEA
 - A Tri-Societies affiliate offering courses in agriculture disciplines to close the gap in knowledge
 - Offering Precision Agriculture Majors and Minors at Land Grant Universities
 - I encourage you to seek out classes involving: agriculture machinery and management, agronomy, and data science



Lowering Barriers Will Push Precision Ag Progress Forward





Revision of PA Basics Book

We are looking for students to review the updated and new chapters

Certificate from ASA-CSA-SSSA to put on your resume

Questions?

Contact information: skye.brugler@sdstate.edu

Office: Raven 105, please stop by anytime!