BACKGROUND

Variety selection is an important consideration for growing high quality oats. Oats are bred for numerous traits including disease resistance, maturity, height, yield and lodging. Oats are grown for various end uses, including food grade milling, cover crop seed, livestock feed (either as grain, hay, or oatlage), straw bedding or as a nurse crop for alfalfa. Matching the variety characteristics with the end market is an important step for successful production. For instance, farmers looking to grow oats for livestock feed may be interested in selecting oats that have the highest yield possible, whereas farmers looking to sell oats into a food-grade market should select a variety that has high test weight, because the grain must reach a certain heft to avoid dockage.

To examine oat variety performance across Iowa, PFI has conducted variety trials every year since 2015. Occasionally one variety will perform the best, in terms of yield, across all trial locations. But more commonly the top yielding variety differs across sites. That fact that some varieties perform well in certain locations but not in other locations is called a genotype-by-environment interaction and it is commonplace in plant breeding, especially among small grains.

The best way to get reliable variety performance given genotype-by-environment interactions is to set up hyper-localized variety trials. Usually, however, that is not feasible. So instead, PFI, in conjunction with plant breeders, has been working since 2018 on developing models to predict small grain performance in localized environments. These genotype-by-environment models power a decision tool that takes into account existing small grains variety data and uses that information to predict performance of varieties based on a grower’s ZIP code.

In a Nutshell:

• Variety selection is a key component of growing high quality oats, but information about which varieties are best suited to perform in a particular location is often lacking.
• Cooperators compared the performance of their traditional oat variety with a variety predicted to do well from a genotype-by-environment model.

Key Findings:

• At three of the five farms there was no difference in yield performance between the two oat varieties. At one farm (Ingels) the model-selected oat variety had a higher yield than the traditional variety, but at another farm (Fehr), the opposite occurred – the traditional variety outperformed the model-selected variety.
• At two of the five farms there was no difference in test weight between the traditional and model-selected oat varieties. At two farms (Fehr and Rosmann) the model-selected variety had a higher test weight than the traditional variety, whereas at another farm (Wedemeier) the traditional variety had a higher test weight.
• Coming up with models that can reliably estimate variety performance across farm locations is challenging and requires a combination of historical data, predictions and on-the-ground calibration. These results indicate that our current models predicting oat variety selection can continue to be refined.

Cooperators
Darren Fehr - Mallard
Chad Ingels - Randalia
Ron Rosmann - Harlan
Tony Thompson - Elkhart
Scott Wedemeier - Maynard

Collaborators
Lucia Gutierrez, University of Wisconsin-Madison
Paul Sandro Garcia, University of Wisconsin-Madison
Madhav Bhatta, University of Wisconsin-Madison

Funding
SARE
Walton Family Foundation

Oat Selector Tool Variety Trial

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While models can be extremely useful, they always benefit from calibration and refinement. To understand how effectively this variety selection model was at predicting oat performance, five cooperators conducted a trial comparing the oat variety they would typically use ("traditional" variety) alongside a variety predicted to perform well in the model ("model-selected" variety).

**METHODS**

All farms planted their traditional variety alongside the model-selected variety in randomized replicated strip trials (Figure 1A). Information about each variety used in these trials can be found in the Appendix (Table A1). The varieties trialed at each farm as well as oat production practices can be found in Table 1.

Yields were collected by the farmer at each site and samples of grain from each strip replicate were mailed to the University of Wisconsin-Madison where they were analyzed for test weight.

**Data analysis**

Data were analyzed in R version 4.0.2. A two-way ANOVA with farm and variety type (traditional vs model-selected) was fit for our two response variables: yield and test weight. Means separations are reported using Tukey's Least Significant Difference (LSD) determined at the 95% confidence level.

**RESULTS AND DISCUSSION**

For both our response variables – yield and test weight – there was a significant interaction between our predictor variables – farm and variety source. This indicates that the direction of our response (whether the model-selected variety outperformed traditional variety) varied across sites.

**Yield**

Yields ranged from 42 to 137 bu/ac across sites (Figure 1). Thompson's oats had the lowest yields, but they also were the only oats that did not receive any fertilizer (Table 1). At Fehr's farm 41M (the traditional variety) had higher yield than Sumo (the model-selected variety). At Ingels' farm Saddle (the model-selected variety) had higher yield than Goliath (the traditional variety). At Rosmann's, Thompson's, and Wedemeier's farms there was no statistically significant differences between oat yields.

**Test weight**

Test weight ranged from 34.6 to 46.6 lb/bu across farms (Figure 2). The majority of farms had both varieties meet the threshold for food grade milling – 38 lb/bu. Sumo oats (the model-selected variety) at Fehr's farm had the highest test weight. At Fehr's and Rosmann's farms, Sumo and MN-Pearl, the model-selected varieties, had a higher test weight than the traditional varieties, 41M and Deon, respectively. At Wedemeier's farm, Reins, the traditional variety, had a higher test weight than the model-selected variety, Hayden. At Ingels' and Thompson's farms there was no statistically significant difference between oat test weights.

**CONCLUSIONS AND NEXT STEPS**

The mixed results from these trials demonstrate that selecting a variety based on a model may not always result in higher yield or test weight. While models can be useful, they also need calibration to be accurate. These on-farm trials provide valuable information that will refine efforts to leverage small grain variety information into models that can assist farmers in decision making.

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**TABLE 1. Oat varieties and management information for each farm**

<table>
<thead>
<tr>
<th></th>
<th>FEHR</th>
<th>INGELS</th>
<th>ROSMANN</th>
<th>THOMPSON</th>
<th>WEDEMEIER</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Oat Varieties</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Traditional</td>
<td>41M</td>
<td>Goliath</td>
<td>Deon</td>
<td>Esker2020</td>
<td>Reins</td>
</tr>
<tr>
<td>Model-selected</td>
<td>Sumo</td>
<td>Saddle</td>
<td>MN-Pearl</td>
<td>Deon</td>
<td>Hayden</td>
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<tr>
<td><strong>Management</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Previous crop</td>
<td>Soybeans</td>
<td>Corn</td>
<td>Soybeans</td>
<td>Corn</td>
<td>Corn</td>
</tr>
<tr>
<td>Reps</td>
<td>3</td>
<td>4</td>
<td>4</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>Plot size (ft)</td>
<td>40 x 1100</td>
<td>12 x 987</td>
<td>13.5 x 1163</td>
<td>10 x 450</td>
<td>45 x 1194</td>
</tr>
<tr>
<td>Planting date</td>
<td>Apr. 2</td>
<td>Apr. 17</td>
<td>Apr. 1</td>
<td>Mar. 8</td>
<td>Apr. 10</td>
</tr>
<tr>
<td>Underseeding</td>
<td>none</td>
<td>Alfalfa pasture mix</td>
<td>Alfalfa, red clover and orchard grasses</td>
<td>none</td>
<td>Red clover</td>
</tr>
<tr>
<td>Seeding rate (seeds/ac)*</td>
<td>1,500,000 - 1,749,999</td>
<td>1,000,000 - 1,249,999</td>
<td>1,749,999 - 2,000,000</td>
<td>1,000,000 - 1,249,999</td>
<td>1,500,000 - 1,749,999</td>
</tr>
<tr>
<td>Tillage</td>
<td>Apr. 2</td>
<td>Field cultivation</td>
<td>none</td>
<td>Mar. 21</td>
<td>Disked twice</td>
</tr>
<tr>
<td>Fertilizer</td>
<td>Nov. 1, 2019</td>
<td>1 ton/ac chicken litter with light incorporation</td>
<td>Apr. 20</td>
<td>200 lb/ac potash 100 lb/ac AMS</td>
<td>Mar. 21</td>
</tr>
<tr>
<td>Harvest date</td>
<td>July 22</td>
<td>July 22 - swathed</td>
<td>July 25 - picked up</td>
<td>July 21 - swathed</td>
<td>July 22 - picked up</td>
</tr>
</tbody>
</table>

*Seeding rate converted from bu/ac using estimates of 32 lb/bu test weight and 15,000 seeds/lb seed weight.
FIGURE 1. Yield comparisons between farmer-chosen oat varieties (traditional) and the variety predicted to do well by the model (model-selected). Bar height represents average yield, while points represent the yield from individual replicates. By farm, letters above bars indicate whether differences in yield within a farm were greater than the least significant difference (LSD), indicating statistically significant differences at the 95% confidence level.

FIGURE 2. Test weight comparisons between farm-chosen oat varieties (traditional) and the variety predicted to do well by the model (model-selected). Bar height represents average test weight, while points represent the test weight from individual replicates. By farm, letters above bars indicate whether differences in test weight within a farm were greater than the least significant difference (LSD), indicating statistically significant differences at the 95% confidence level. The dashed line at 38 lb/bu indicates the “gold standard” of test weight. Anything over a 38 lb/bu test weight is eligible for food grade milling without price dockage.
APPENDIX - TRIAL DESIGN AND WEATHER CONDITIONS

Table A1. Origin, PVP, and disease ratings for varieties used across farms.

<table>
<thead>
<tr>
<th>Variety</th>
<th>Origin*</th>
<th>Year Released</th>
<th>PVPb</th>
<th>Maturity</th>
<th>Disease Ratingsc</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>41M</td>
<td>SK</td>
<td>2016</td>
<td>Yesf</td>
<td>Medium</td>
<td>R</td>
<td>S</td>
</tr>
<tr>
<td>Deon</td>
<td>MN</td>
<td>2014</td>
<td>Yes</td>
<td>Late</td>
<td>MR</td>
<td>MS</td>
</tr>
<tr>
<td>Esker2020</td>
<td>WI</td>
<td>2020</td>
<td>Yes</td>
<td>Mid-Late</td>
<td>MS</td>
<td>MS</td>
</tr>
<tr>
<td>Goliath</td>
<td>SD</td>
<td>2013</td>
<td>Yes</td>
<td>Late</td>
<td>MS</td>
<td>R</td>
</tr>
<tr>
<td>Hayden</td>
<td>SD</td>
<td>2015</td>
<td>Yes</td>
<td>Mid-Late</td>
<td>MS</td>
<td>MS</td>
</tr>
<tr>
<td>MN-Pearl</td>
<td>MN</td>
<td>2018</td>
<td>Yes</td>
<td>Mid-Late</td>
<td>MS</td>
<td>MR</td>
</tr>
<tr>
<td>Reins</td>
<td>IL</td>
<td>2016</td>
<td>Yes</td>
<td>Early</td>
<td>MR</td>
<td>MR</td>
</tr>
<tr>
<td>Saddle</td>
<td>SD</td>
<td>2018</td>
<td>Pending</td>
<td>Early</td>
<td>MR</td>
<td>MR</td>
</tr>
<tr>
<td>Sumo</td>
<td>SD</td>
<td>2017</td>
<td>Yes</td>
<td>Early</td>
<td>MR</td>
<td>MR</td>
</tr>
</tbody>
</table>

* Origin: IL-University of Illinois, MN-University of Minnesota; SD-South Dakota State University; SK-Saskatoon, Saskatchewan, Canada; WI-University of Wisconsin.

b PVP = Plant Variety Protection. The PVP Act provides a certificate to the developer of a variety granting exclusive rights for reproducing and marketing the seed.

c Disease Ratings: S = susceptible; MS = moderately susceptible; MR = moderately resistant; R = resistant.

d Disease: BYDV = Barley Yellow Dwarf Virus.

e Also known as Ore3541M

 Origin: IL-University of Illinois, MN-University of Minnesota; SD-South Dakota State University; SK-Saskatoon, Saskatchewan, Canada; WI-University of Wisconsin.

 FIGURE A1. Sample experimental design used by the cooperators. Each farm had 3-4 replicates of their traditional variety and the variety predicted by the model. This design allows for statistical analysis of the results.
PFI’s Cooperators’ Program helps farmers find practical answers and make informed decisions through on-farm research projects.

The Cooperators’ Program began in 1987 with farmers looking to save money through more judicious use of inputs.

If you are interested in conducting an on-farm trial contact Stefan Gailans @ 515-232-5661 or stefan@practicalfarmers.org.

FIGURE A2. Monthly average temperature (top) and total rainfall (bottom) for February - August. Bars represent the 2020 growing year, while lines represent long term averages (1950-2019) for each site. Data taken from the nearest weather stations to each farm: Fehr – Pocahontas, Ingels – Fayette, Rosmann – Harlan, Thompson – Ankeny, Wedemeier – Oelwein.

REFERENCES


