



# Development of an Open-Source Web-Based Tool for Alfalfa Yield and Quality Prediction

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## RATIONALE & OBJECTIVES

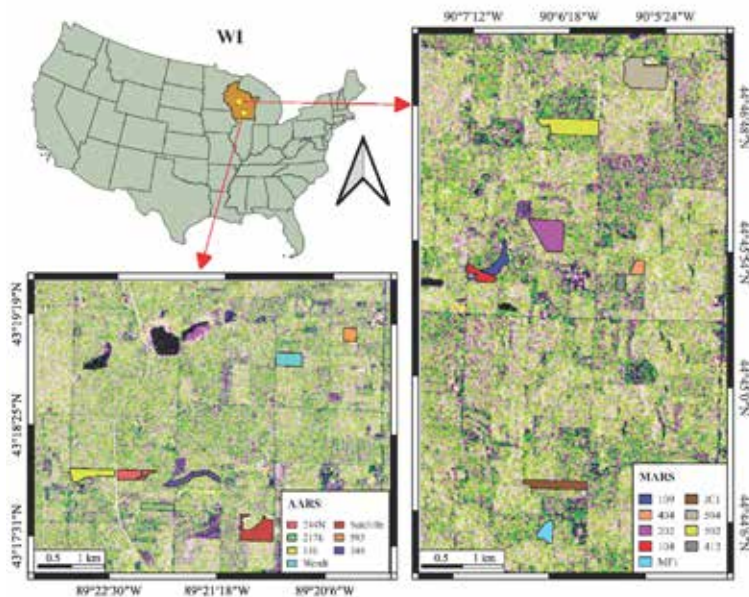
- Develop machine learning models to estimate in-season alfalfa yield and quality using multi-source data and validate the results on Wisconsin Alfalfa Yield and Persistence database.
- Create an open-source web-based tool to provide end-users with in-season alfalfa yield and quality maps to assess their spatial variability for precision management.

## STUDY DESCRIPTION

### Locations:

Arlington Agricultural Research Station (7 sites),  
Marshfield Agricultural Research Station (9 sites).

Figure 1. The distribution of sampling sites.



### Data:

The features extracted from Sentinel-1 Ground Range Detected (GRD) products, environmental factors from Daymet V4 product and cutting order.

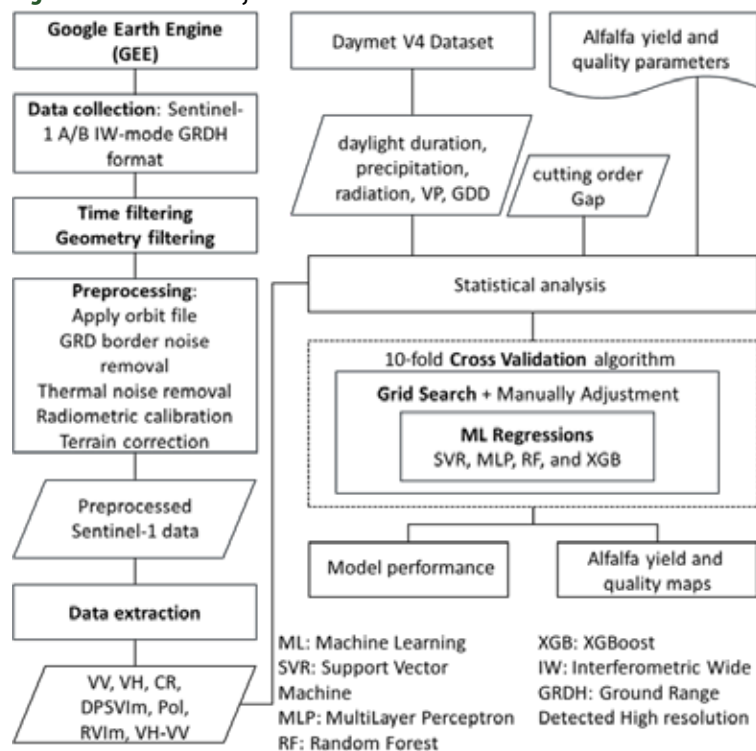
### Machine Learning Models:

Support Vector Regression (SVR), Multilayer Perceptron (MLP), Random Forest (RF), and Extreme Gradient Boosting (XGB).

### Yield and Quality:

Dry matter yield (DMY), crude protein (CP), neutral detergent fiber (NDF), NDF digestibility (NDFD), and acid detergent fiber (ADF).

Figure 2. Workflow of Objectives 1.



### Development environment:

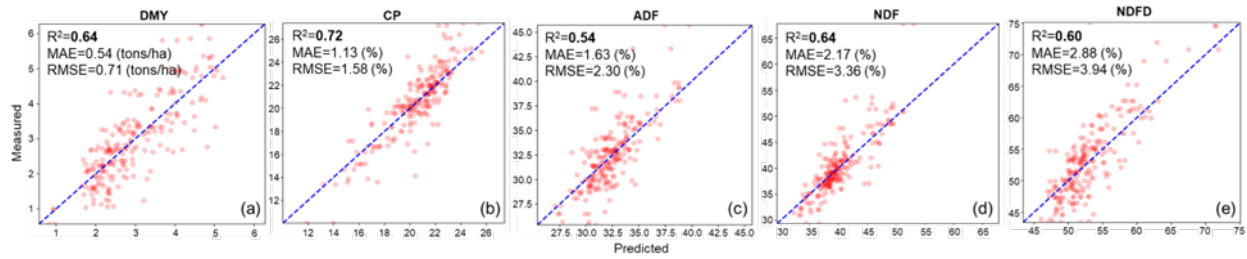
HTML+ JavaScript+ Python (Flask)+ PythonAnywhere service.

## RESULTS

### Objectives 1:

- Estimated and measured alfalfa yield and quality traits (Figure 3).
- The XGB model consistently performed the best for all alfalfa yield and traits.
- The DMY estimates highly agreed with the measured values with an average  $R^2$  of 0.64 and RMSE of 0.71 tons/ha. The best result for estimating CP was with an average  $R^2$  of 0.72. The performance in estimating alfalfa fiber indicators was achieved with the highest average  $R^2$  of 0.54, 0.64, and 0.60, respectively.

**Figure 3.** Scatterplots of the estimated and measured alfalfa yield and quality traits, where a - e are for DMY, CP, ADF, NDF, and NDFD, respectively. Estimated values in the plots for individual samples and evaluation metrics were averaged over the 20 times cross validation process.

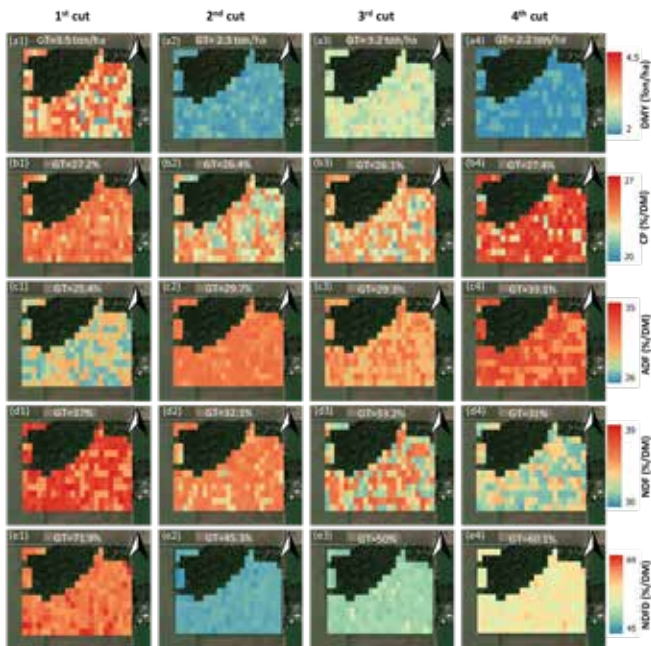


**Table 1.**  $R^2$  of the four machine learning models in estimating alfalfa yield and quality traits.

Input	Traits	SVR	MLP	RF	XGB
With Environmental factors	DMY	0.53	0.53	0.59	<b>0.64</b>
	CP	0.23	0.24	0.59	<b>0.72</b>
	ADF	0.27	0.26	0.35	<b>0.54</b>
	NDF	0.23	0.24	0.49	<b>0.64</b>
	NDFD	0.23	0.27	0.40	<b>0.60</b>
Without Environmental factors	DMY	0.40	<b>0.47</b>	<b>0.47</b>	0.45
	CP	<b>0.19</b>	0.18	0.13	0.11
	ADF	<b>0.15</b>	<b>0.15</b>	0.07	0.07
	NDF	0.01	<b>0.04</b>	-0.00	-0.10
	NDFD	0.12	0.12	<b>0.18</b>	0.14

- Performance for the fiber indicators and CP was not ideal only with the SAR features with the best  $R^2$  all  $<0.2$ .
- Performance was significantly higher when environmental factors were added, especially for the quality traits.

**Figure 4.** Visualization of estimated alfalfa DMY and quality traits of Sutcliffe in 2020.



- The average values of each map corresponded well with their field-level ground data.
- The estimated maps also demonstrated the spatial variations of each field.

## Objectives 2:

- The webtool: <https://alfalfabse.pythonanywhere.com>.
- The webtool can present the saved yield maps of the two study locations in the database.

**Figure 5.** Some saved yield map of fields in location A.



- The users can also define an area and the date that they wish to show the yield for.

**Figure 6.** Yield map of a newly defined area.



## CONCLUSIONS/SUGGESTIONS

- The potential of using Sentinel-1 radar features and environmental factors in estimating alfalfa yield and quality was proved in this study.
- Field maps that were generated for alfalfa DMY and quality using the model estimates depicted in-field variations, providing a valuable reference for decision-making of field management practices and alfalfa harvest time.