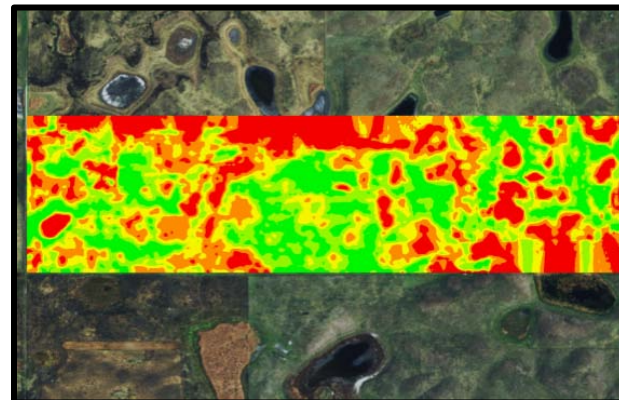
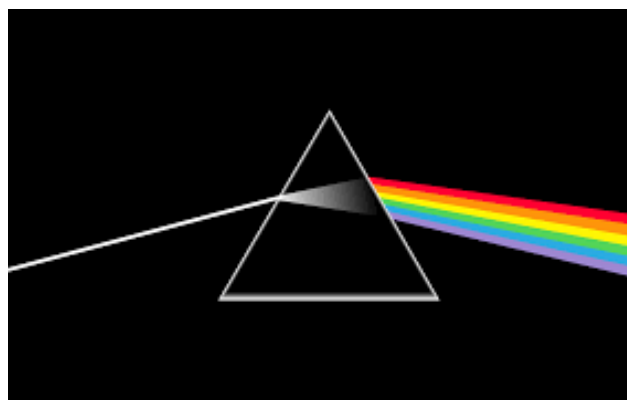


Integrated Remote Sensing Tools for Timely Predictions of Alfalfa Nutritive Value



Reagan L. Noland

M. Scott Wells

Craig C. Sheaffer

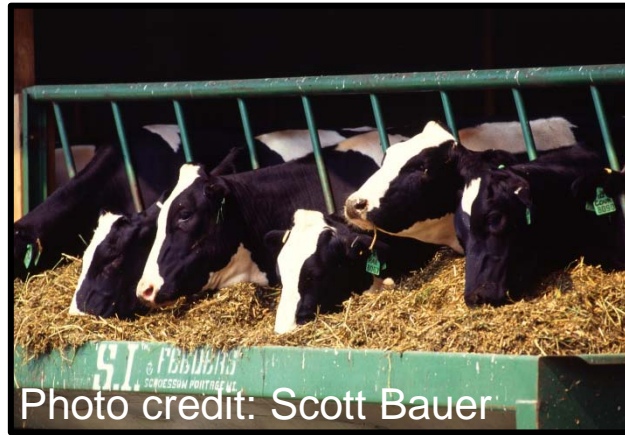
North American Alfalfa
Improvement Conference
July 14, 2016



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Introduction: Importance of Alfalfa

- Environmental value – ecological services
- Economic value – livestock feeding
 - Yield
 - **Nutritive value**



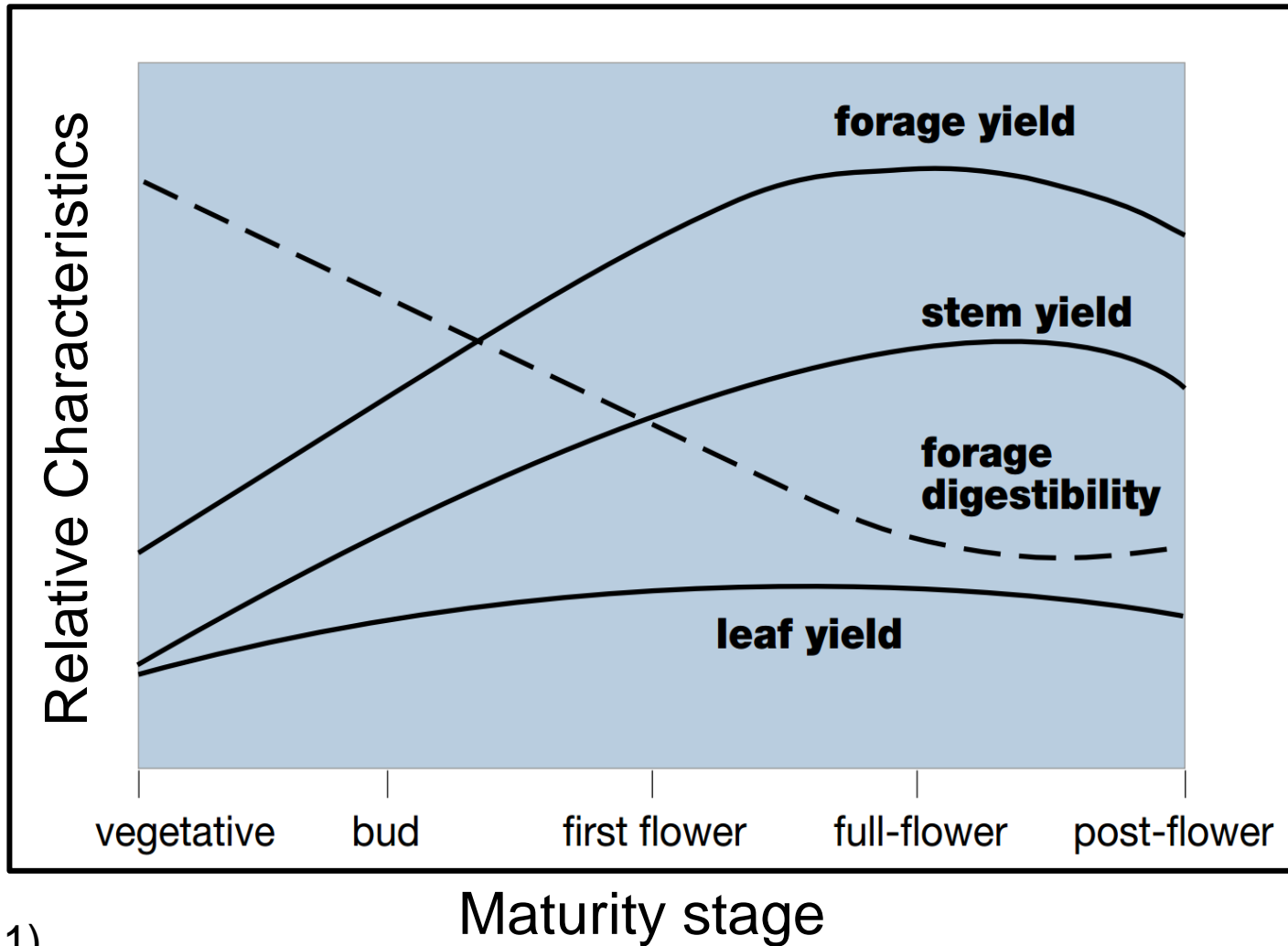
Introduction: Current Hay Prices

Hay Grade	Bale type	----- Price (\$/ton) -----		
		Average	Minimum	Maximum
Prime (> 151 RFV/RFQ)	Small Square	234	125	300
	Large Square	176	150	210
	Large Round	No reported sales		
Grade 1 (125 to 150 RFV/RFQ)	Small Square	115	105	150
	Large Square	139	110	163
	Large Round	91	67	135
Grade 2 (103 to 124 RFV/RFQ)	Small Square	No reported sales		
	Large Square	101	80	123
	Large Round	66	30	120
Grade 3 (87 to 102 RFV/RFQ)	Small Square	No reported sales		
	Large Square	70	65	75
	Large Round	67	45	100

(UW-Extension, 2016)



Nutritive Value vs. Maturity



(ASA, 2011)



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Quantifying Alfalfa Maturity

Mean Stage by Weight (MSW) and Mean Stage by Count (MSC)



Vegetative
growth stages 0-2



Bud
growth stages 3-4



Flower
growth stages 5-6



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Quantifying Alfalfa Maturity

Mean Stage by Weight (MSW) and Mean Stage by Count (MSC)



Vegetative
growth stages 0-2

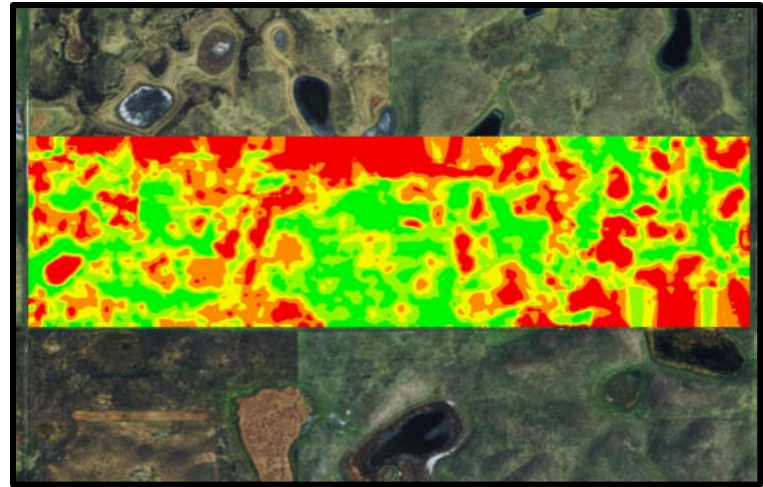
Bud
growth stages 3-4

Flower
growth stages 5-6



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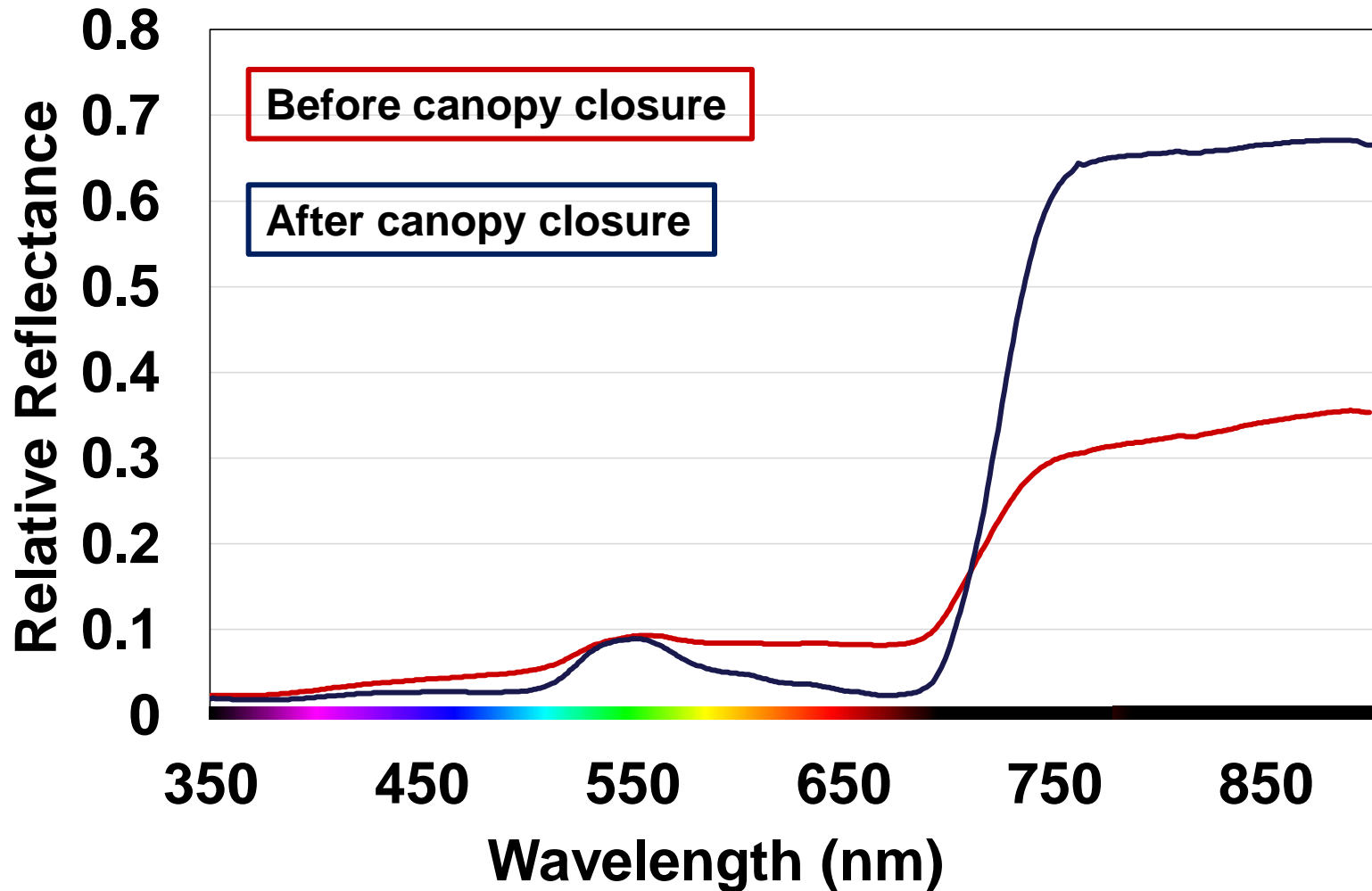
Introduction: Remote Sensing



- Quick, non-destructive assessment
- Information at the field scale
- Optimize timing of harvest
 - (as well as other field operations)



Introduction: Canopy Reflectance



Introduction: Remote Sensing

Canopy Visible and Near-infrared Reflectance Data to Estimate Alfalfa Nutritive Attributes Before Harvest

Patrick J. Starks,^{*} Michael A. Brown, Kenneth E. Turner, and Bradley C. Venuto

ABSTRACT

Canopy reflectance (i.e., remotely sensed) data

P.J. Starks and K.E. Turner, USDA-ARS Grazinglands Research Lab., 7207 West Cheyenne Street, El Reno, OK 73036; M.A. Brown, USDA-ARS Grazinglands Research Lab., Retired; B.C. Venuto, USDA-ARS

Estimation of Biomass and Canopy Height in Bermudagrass, Alfalfa, and Wheat Using Ultrasonic, Laser, and Spectral Sensors

Jeremy Joshua Pittman ^{1,2,*}, Daryl Brian Arnall ², Sindy M. Interrante ¹, Corey A. Moffet ¹ and Twain J. Butler ¹

Application of local binary patterns in digital images to estimate botanical composition in mixed alfalfa-grass fields



Keenan C. McRoberts ^{a,*}, Brent M. Benson ^b, Erika L. Mudrak ^c, David Parsons ^d, Debbie J.R. Cherney ^e

^a Department of Animal Science, Cornell University, 149 Morrison Hall, Ithaca, NY 14853, USA

^b 203 Solutions LLC, 414 Water Street, Unit 1808, Baltimore, MD 21202, USA

^c Cornell Statistical Consulting Unit, Cornell University, B09 Savage Hall, Ithaca, NY 14853, USA

^d School of Land and Food, University of Tasmania, Private Bag 98, Hobart, Tasmania, Australia

^e Department of Animal Science, Cornell University, 329 Morrison Hall, Ithaca, NY 14853, USA

Starks et al., 2016
McRoberts et al., 2016
Pittman et al., 2015



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Objectives

- Explore potential to use known vegetative indices to predict alfalfa maturity and nutritive status
- Develop new predictive models from spectral data



Methods: Design

- A Randomized Complete Block Design was superimposed on a uniform stand of alfalfa at Rosemount, MN.
 - 2014: 3rd cutting (8 replications)
 - 2015: 1st and 3rd cutting (12 replications)
- Treatments: 10 varying stages of alfalfa maturity.



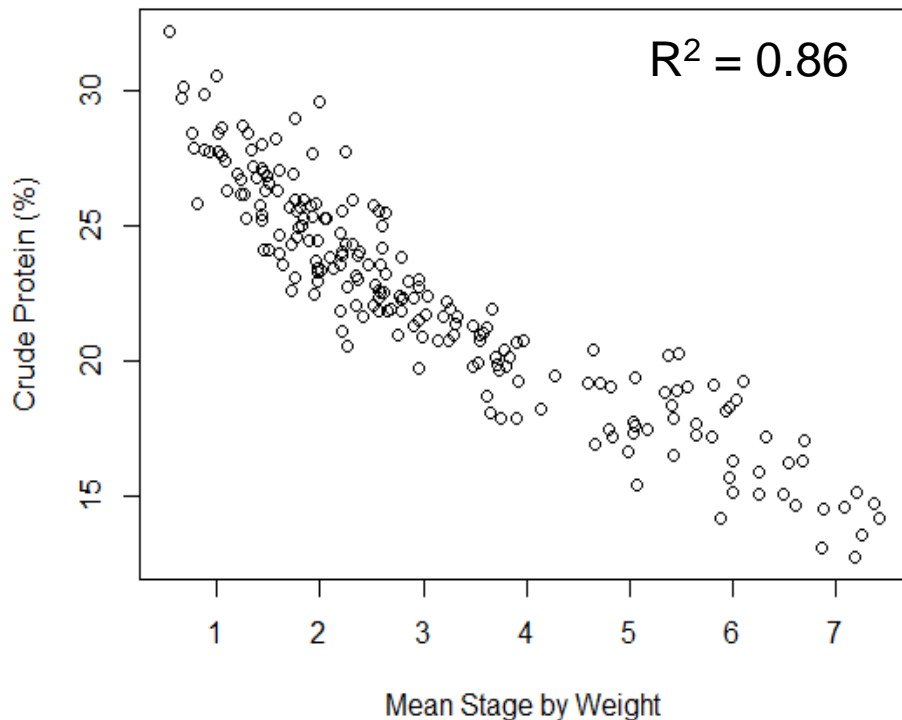
Methods: Data Collection

- Collect canopy reflectance data prior to destructive sampling
 - FieldSpec 4 (ASD Inc.) measured raw reflectance (350-2500 nm)
- Harvest all plots for yield, nutritive status, and maturity assessment
- Nutritive analysis performed with a Perten NIRS system
- Select wavebands correlated to response variables based on AIC (Akaike Information Criterion)
- Fit linear models to the selected predictors.

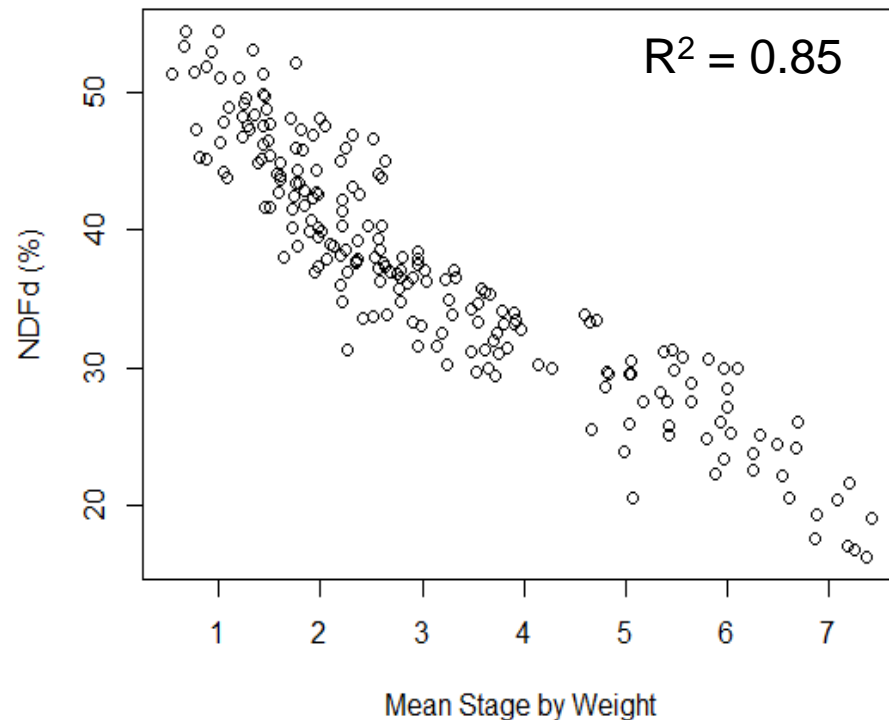


Results: Growth Staging Still Works

MSW vs. CP



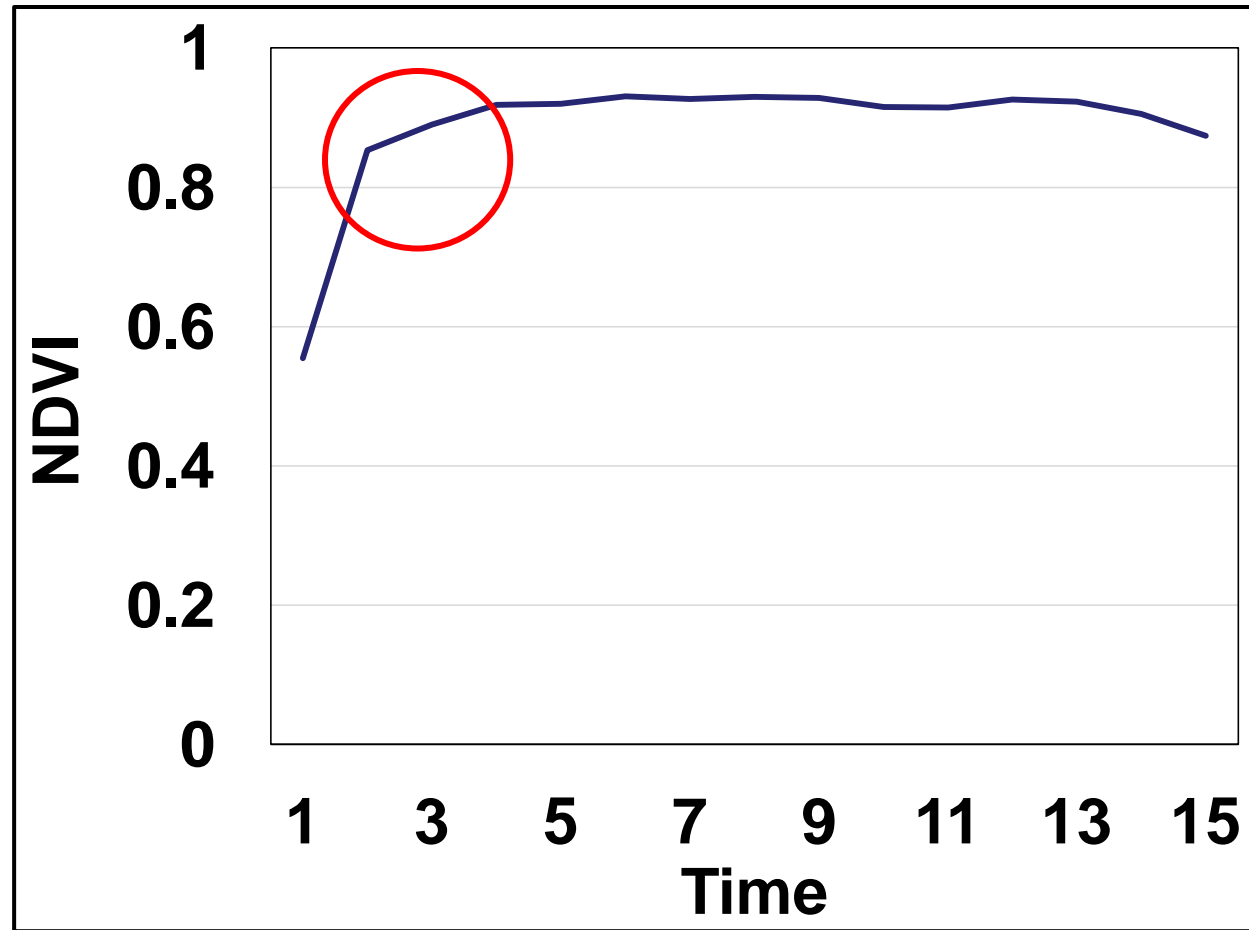
MSW vs. NDFd



Results: Known Indices

NDVI : Normalized Difference Vegetative Index

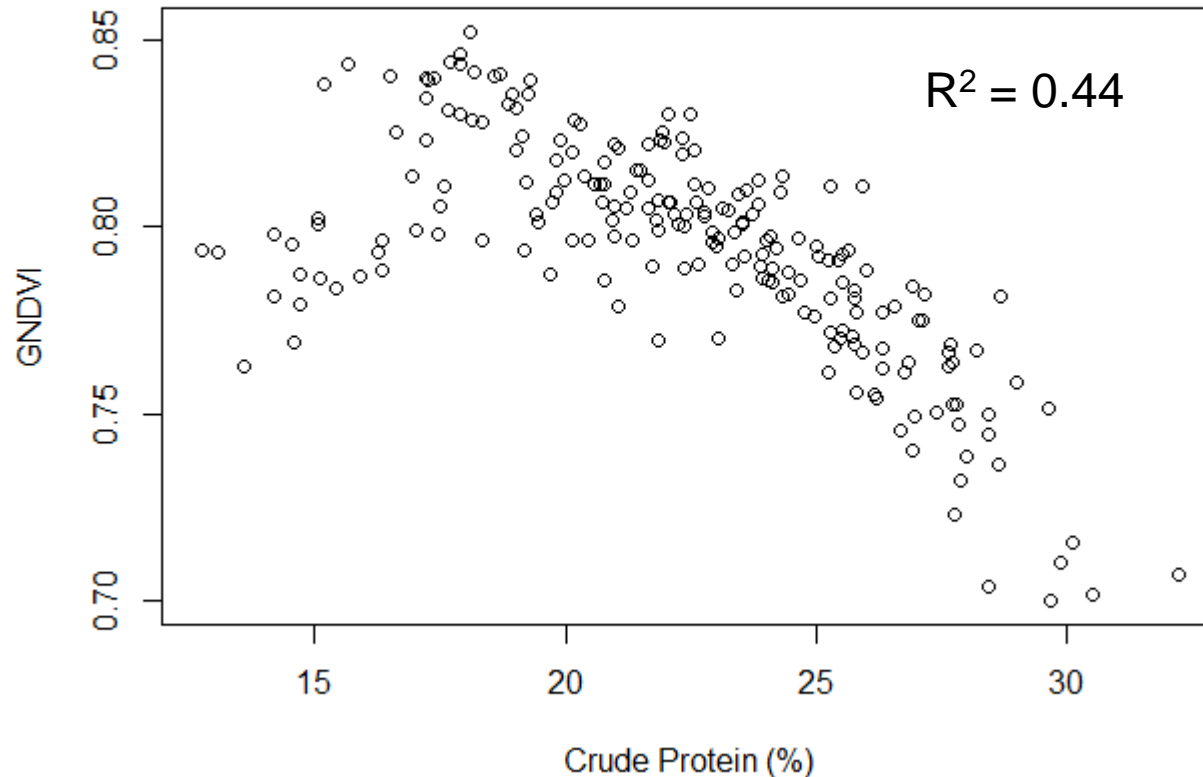
- Common spectral index used in agriculture
- Saturates with canopy closure in alfalfa.



Results: Known Indices

GNDVI : Green Normalized Difference Vegetative Index

Crude Protein vs. Green Normalized Difference Vegetation Index

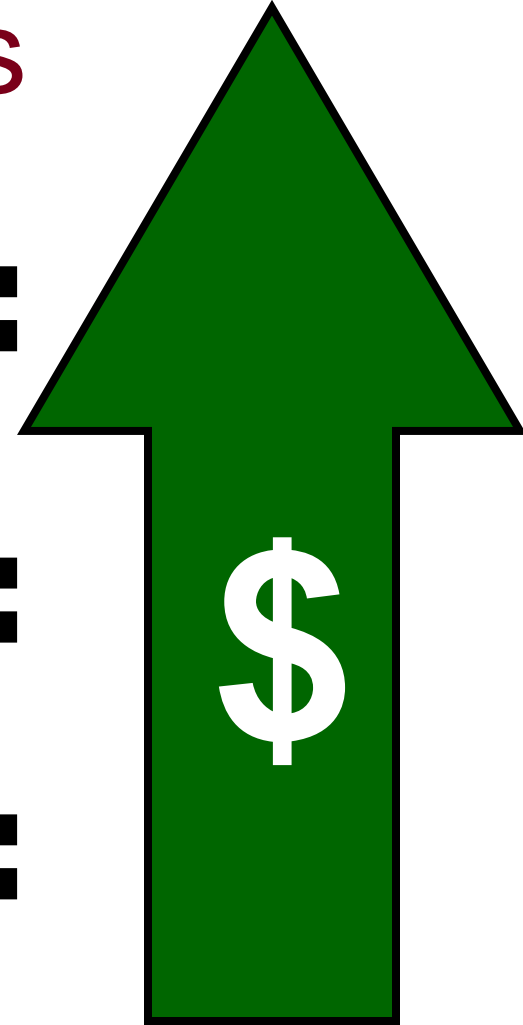


*Best correlation between a published index and crude protein



Limiting economic factors for spectral sensors

- Spectral Range
- Spectral resolution
- Number of bands



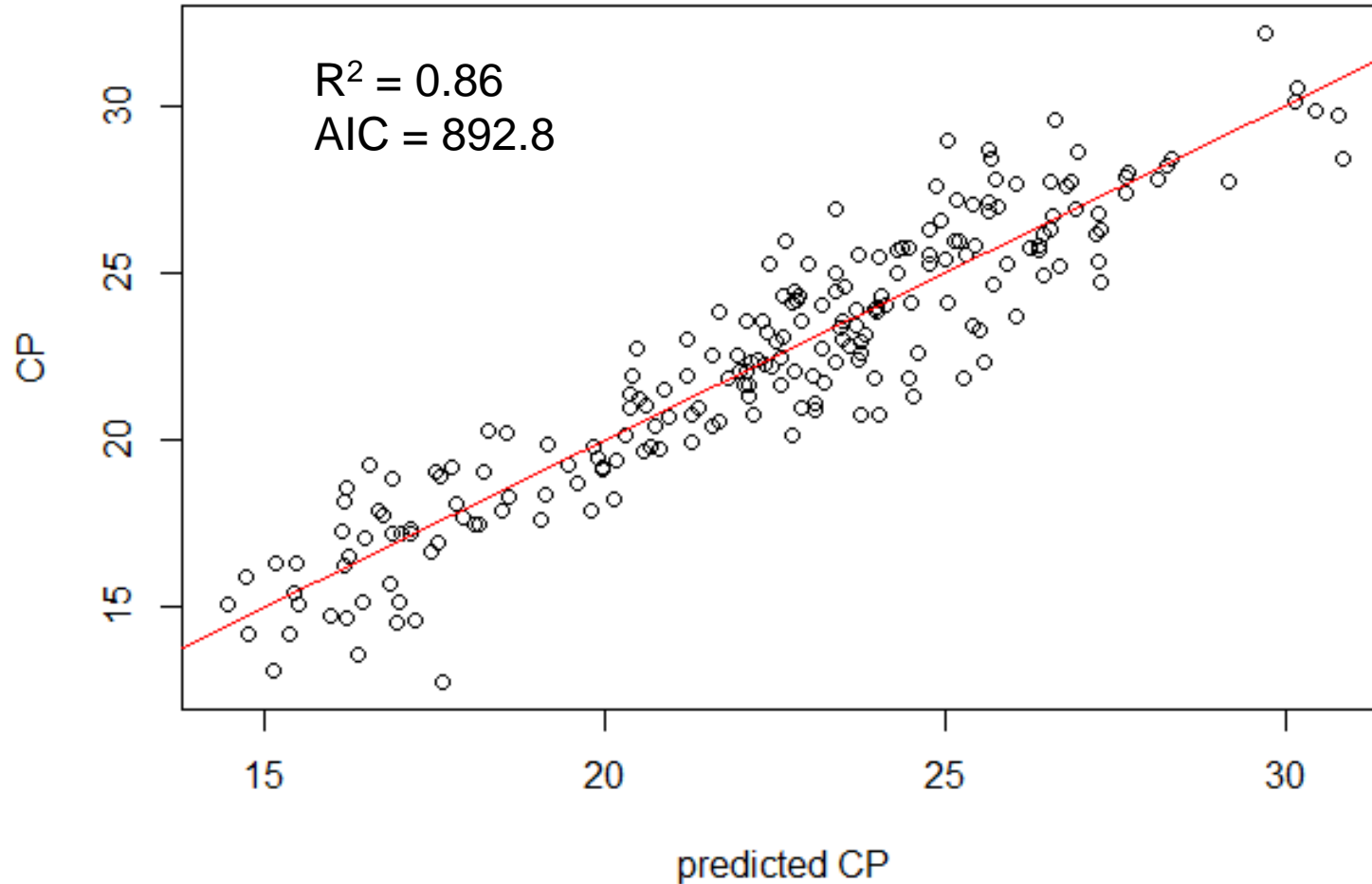
Results: New models

- From the full range of spectral data
- Identified 8 wavebands that best predicted crude protein and minimized AIC
- Checked effects of adding environmental covariates
 - *Growing Degree Units (GDUs) since cut*
- Reduced model to improve utility
 - *Lower spectral range (VIS/NIR)*
 - *Lower resolution (10 nm bands)*



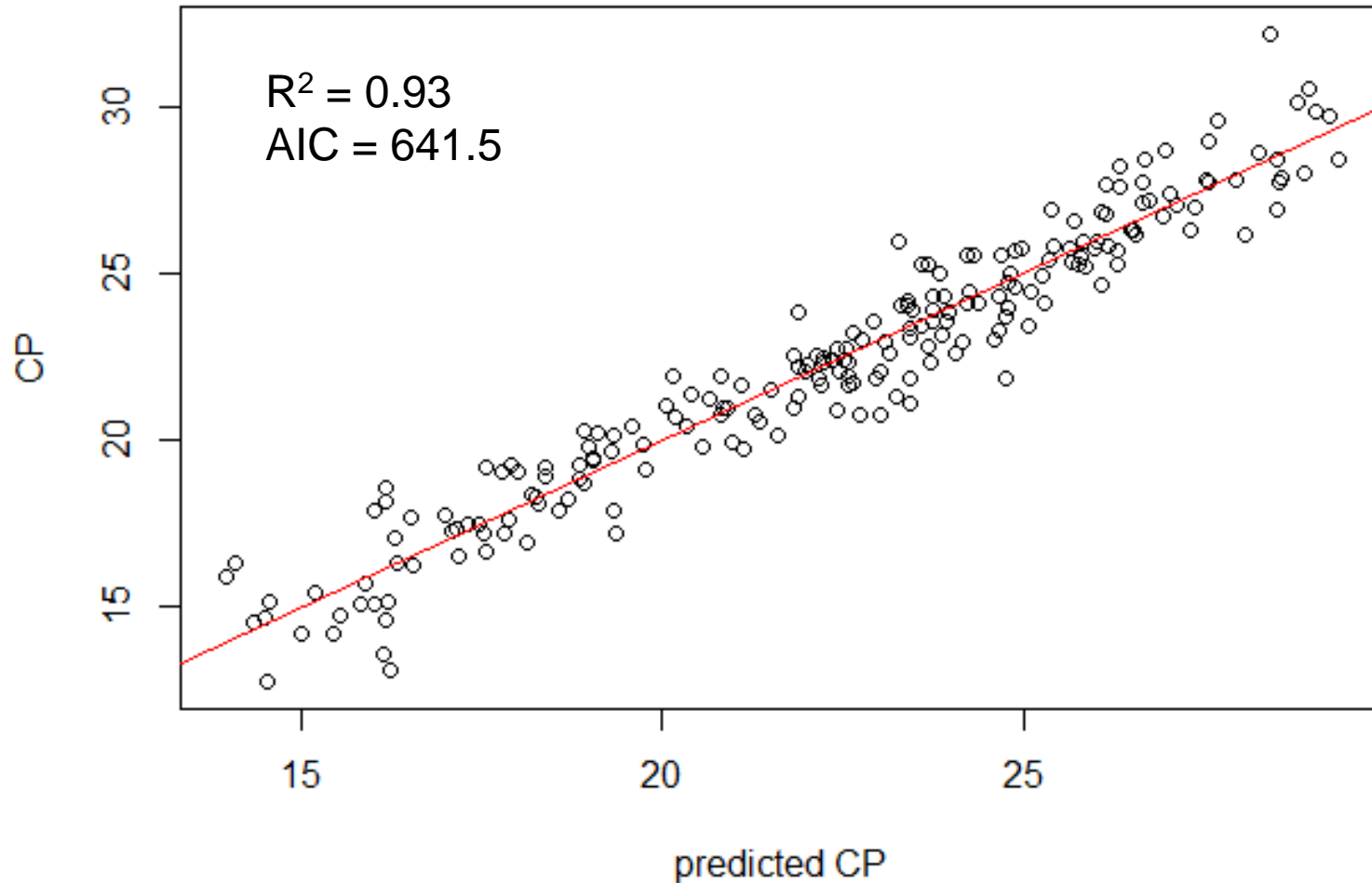
Full Model: 8 bands from 350-2500 nm

CP estimated by 8 wavebands



Full Model with GDU covariate

CP estimated by 8 wavebands with GDU covariate

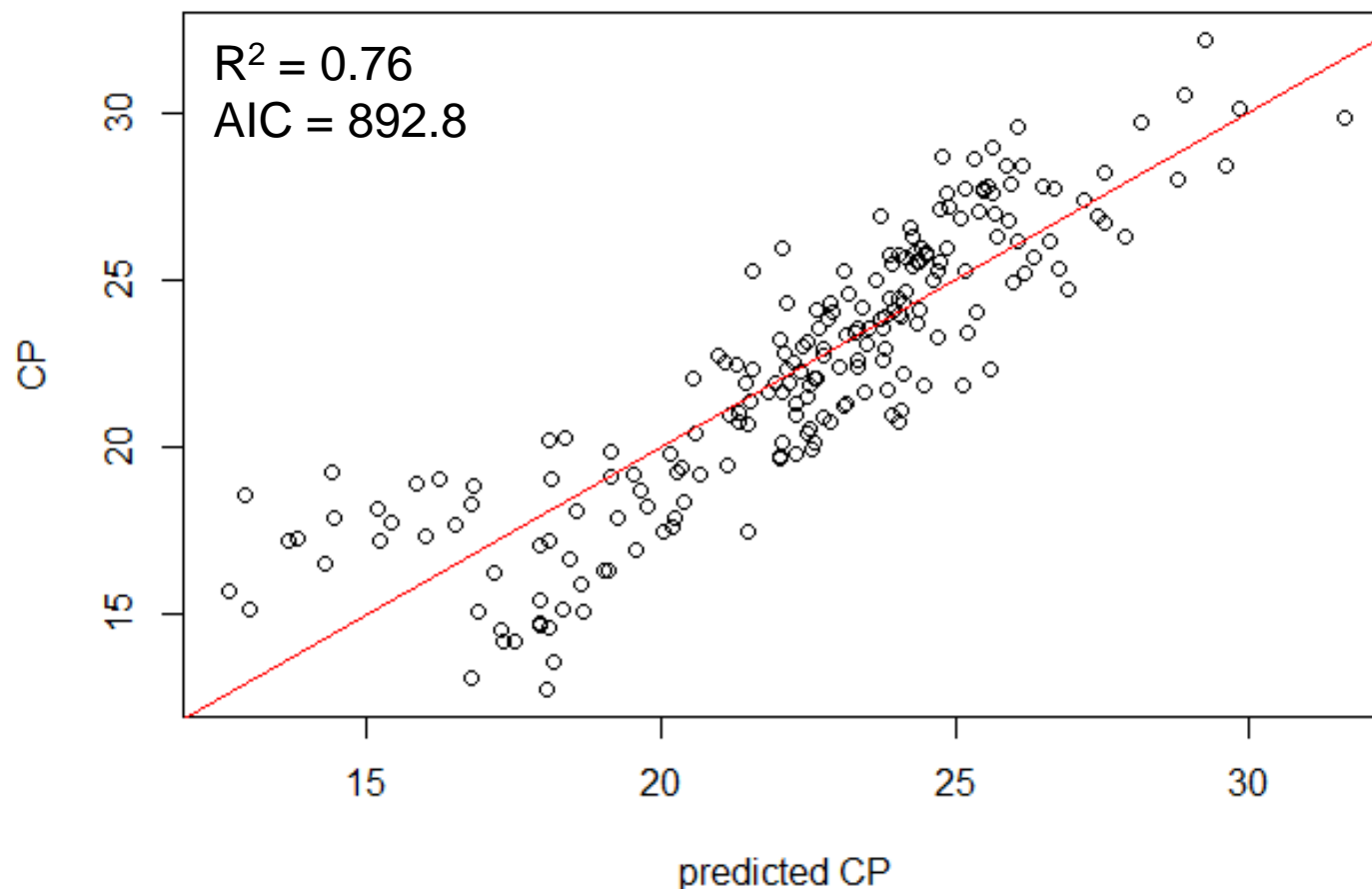


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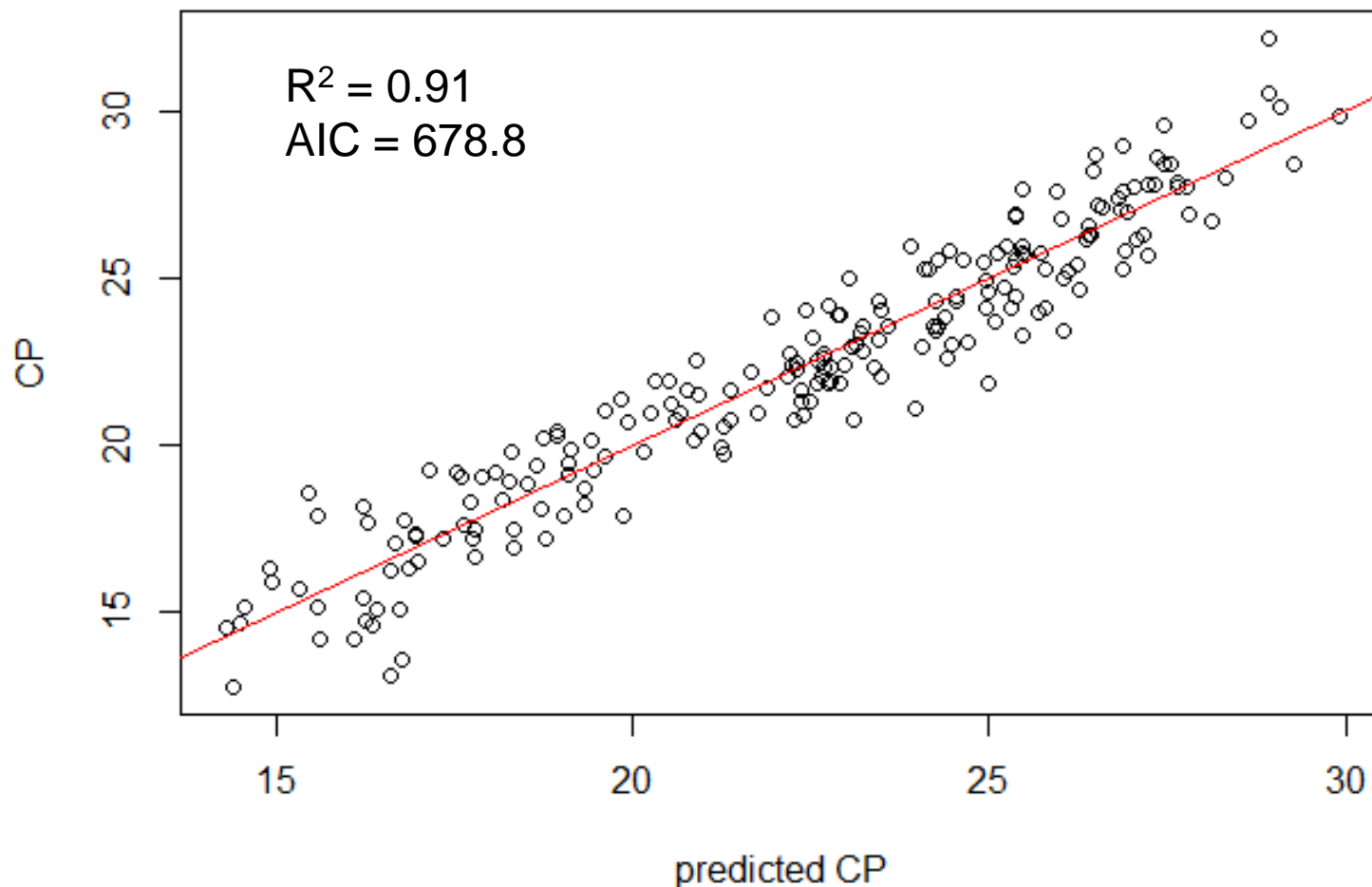
Reduced Model: 3 bands from 350-1100 nm

CP estimated by 3 wavebands



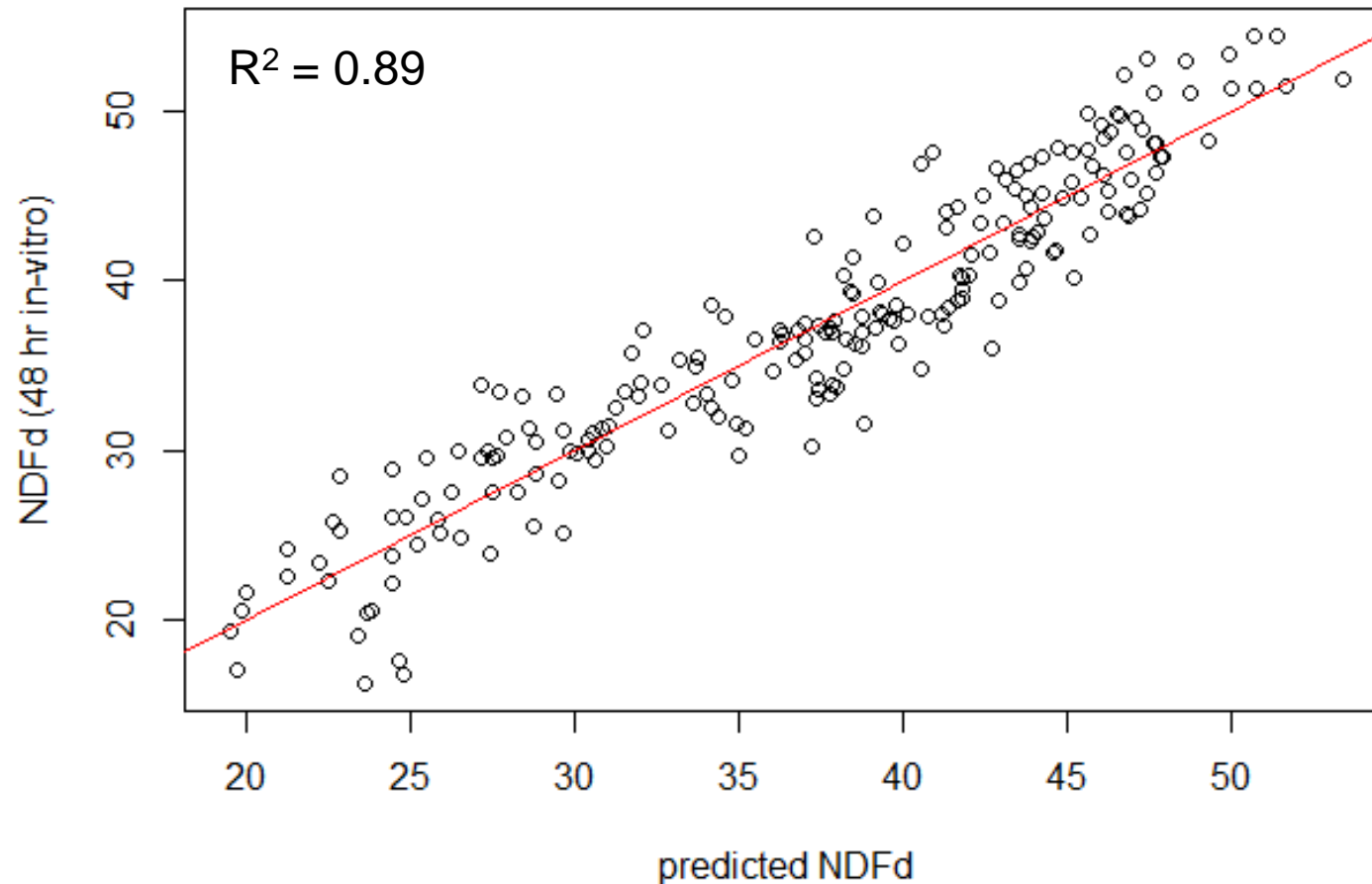
Reduced model with GDU covariate

CP estimated by 3 wavebands with GDU



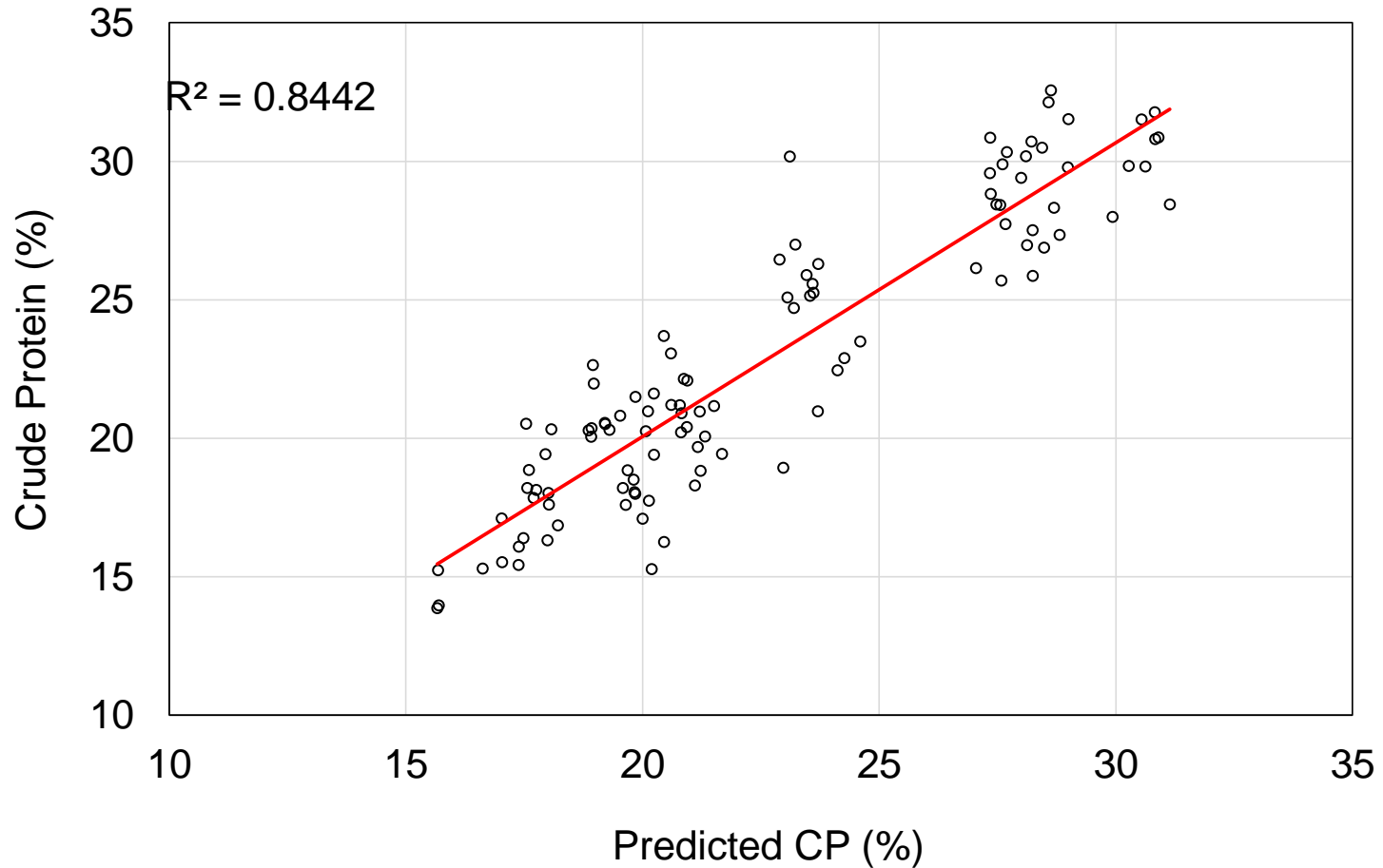
Same 3 bands applied to NDFd

NDFd estimated by GDUs and 3 wavebands

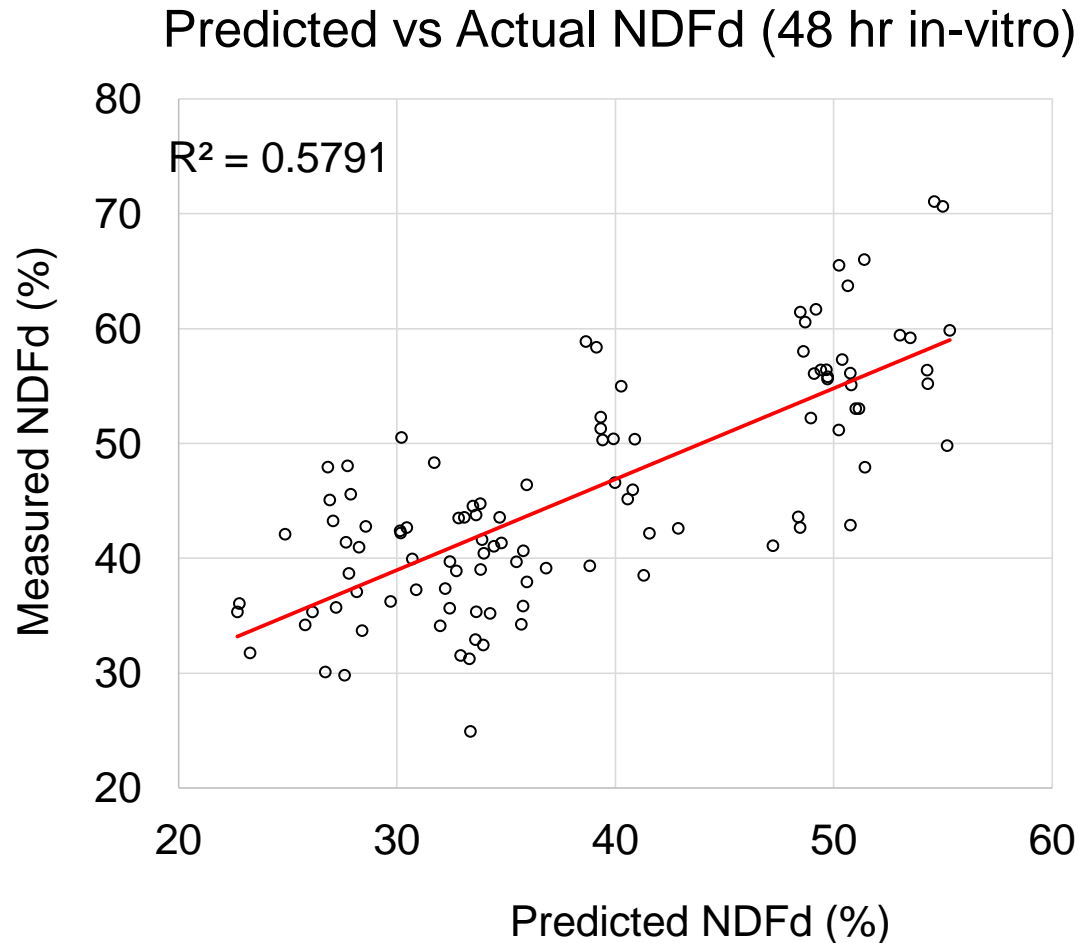


2015 model applied to 2014 data

Predicted vs Actual Crude Protein (2014)



2015 model applied to 2014 data



Conclusions

- Canopy reflectance, integrated with climate information, can inform predictions of alfalfa nutritive value.
- New models using 3 wavebands in the VIS/NIR regions with GDUs as covariate maintained strong predictability and near-optimum model fit.
- The accuracy of passive reflectance measurements is affected by light conditions. Active sensors developed from these results would avoid this issue.



Acknowledgements

- Dr. M. Scott Wells
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- Eric Ristau
- Joshua Larson
- Farm Intelligence
- Farm Nutrients



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Questions?



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