Assessing Multi-season Table Beet Root Yield from Unmanned Aerial Systems

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Introduction

- Table beets have gained popularity due to their health benefits, including cardiovascular improvements and anti-inflammatory properties.
- Remote sensing is essential in precision agriculture for estimating crop yields, but subterranean crops like beets present unique challenges.
- Most yield estimation studies rely on specific growth stages or growing seasons, making models less robust to changes across stages and seasons.

Objective:

- Develop a robust model to predict table beet root yield using multi-season UAS and meteorological data.
- Compare the predictive performance of multispectral systems with hyperspectral + LiDAR systems to assess their relative utility in yield estimation.

Data Collection



Study location: Geneva, New York, USA, at Cornell AgriTech.



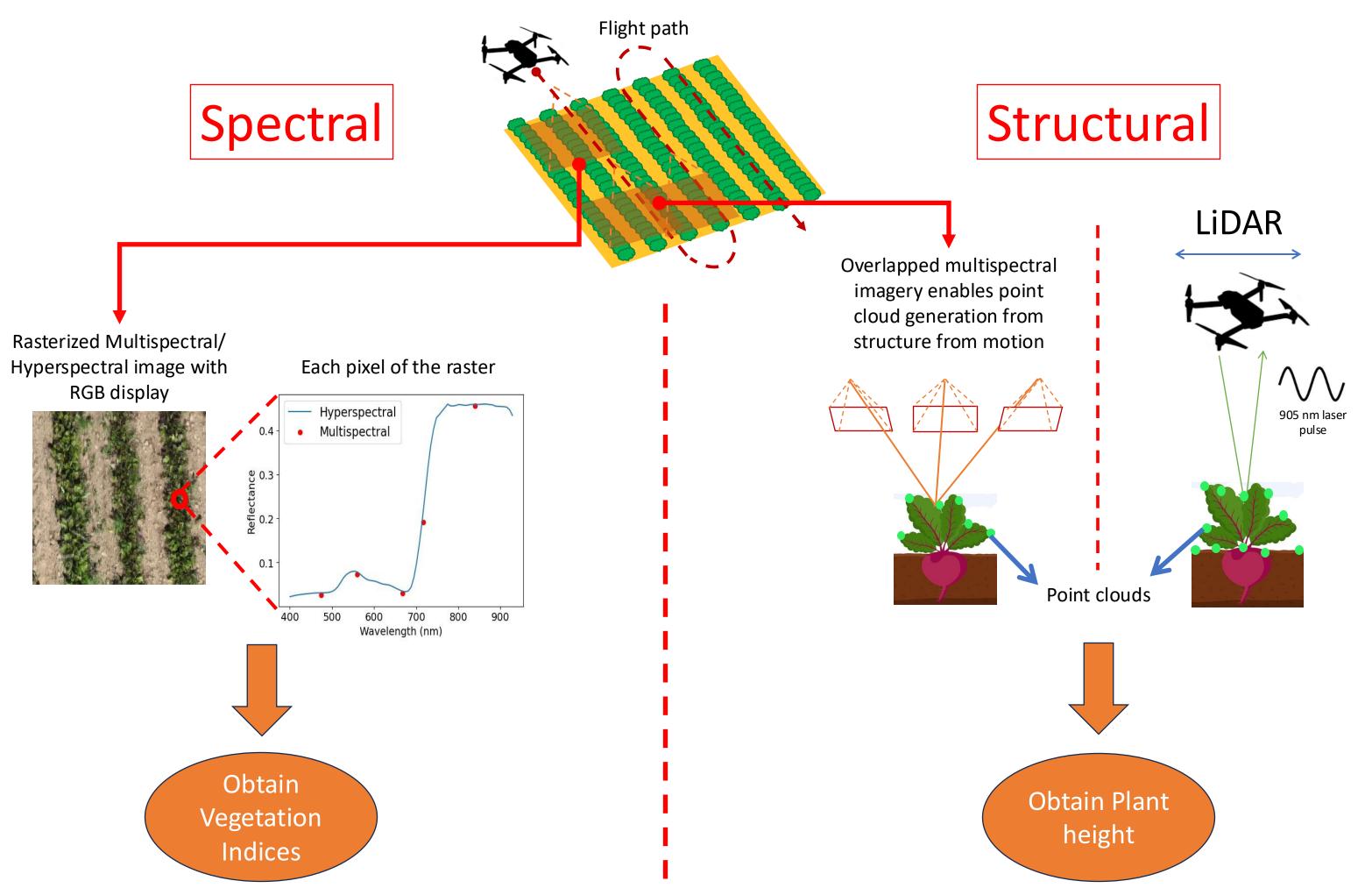
Timeline



College of Science **Chester F. Carlson Center** for Imaging Science

Input Features from UAS

Extracting features from each imaging systems



Modelling Root Yield

EVAP

- Growing Degree Days (GDD) and evapotranspiration (EVAP), obtained from local weather stations, were used as meteorological inputs.
- Several vegetation indices were derived both multispectral hyperspectral imagery, while the mean plant height for each plot was calculated using Structure-from-Motion with multispectral data and LiDAR.
- Gaussian Process Regression (GPR) was applied to predict harvest root yield.

Structural -GDD and

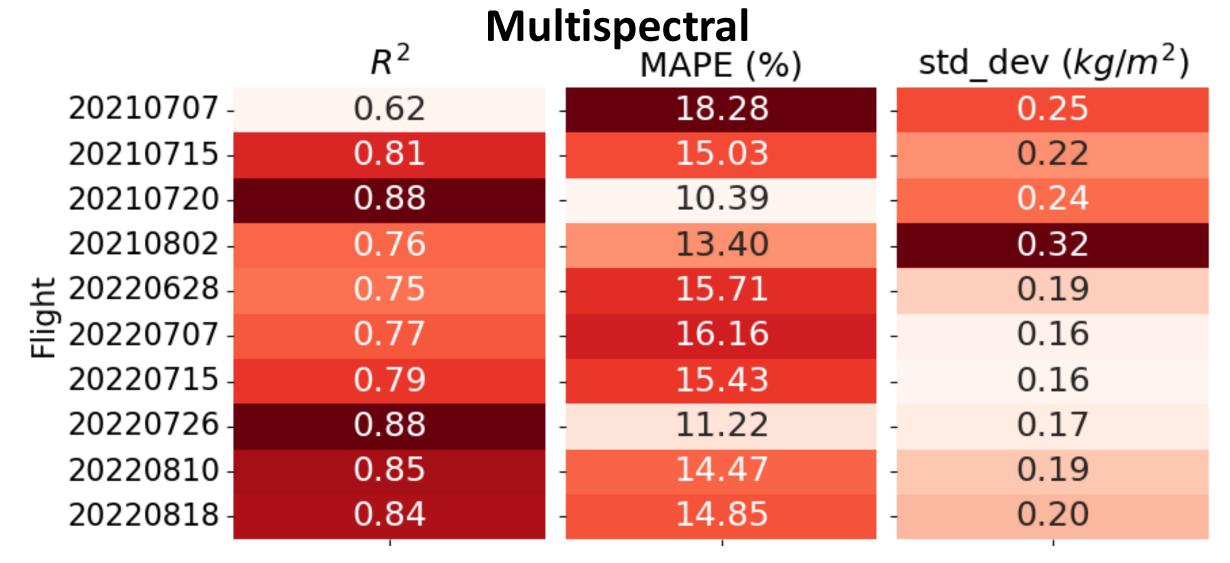
Spectral data

from UAS

Single ML

Model Performance by Flight

The average coefficient of determination (R^2) , mean absolute percent error (MAPE) and standard deviation for model prediction within each flight date for each system are shown below.



Hyperspectral + LiDAR

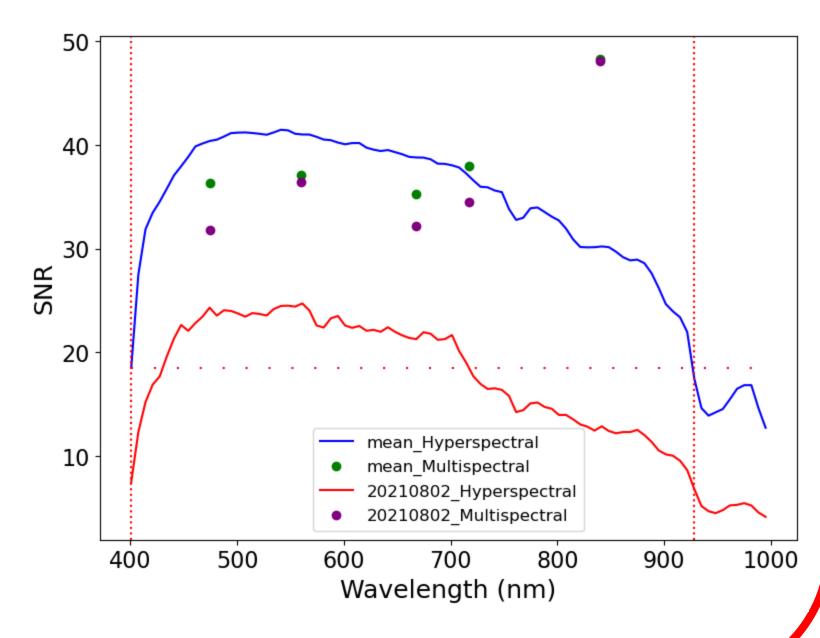
	R^2	MAPE (%)	std_dev (<i>kg/m</i> ²)
20210707 -	0.51	19.62	0.26
20210715 -	0.81	15.00	- 0.23
20210720 -	0.85	10.41	0.25
20210802 -	0.84	7.64	0.37
<u>뉟</u> 20220628 -	0.75	16.64	- 0.18
은 20220707 -	0.76	16.97	- 0.15
20220715 -	0.79	15.38	- 0.16
20220726 -	0.85	13.02	- 0.16
20220810 -	0.86	- 14.27	- 0.20
20220818 -	0.84	14.47	- 0.23
	1		

Discussion

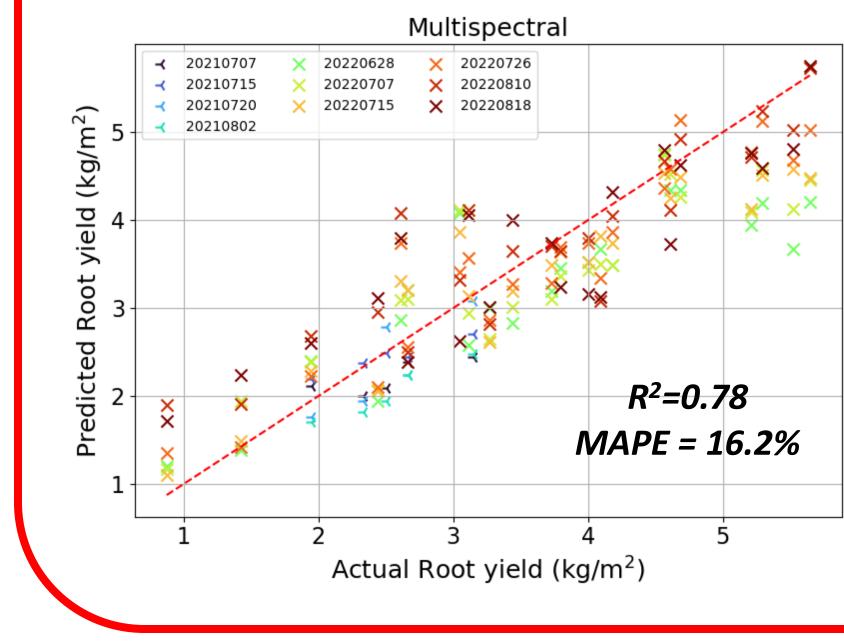
- Earlier flight dates exhibited lower prediction accuracy.
- 2021 data had higher uncertainty due to fewer instances; 2021/08/02 flight had elevated uncertainty across both systems.
- Five-band multispectral sensor demonstrated comparable performance to hyperspectral + LiDAR system.

Yield at

harvest



Overall Model Performance



Hyperspectral + LiDAR $R^2 = 0.78$ MAPE = 15.4%Actual Root yield (kg/m²)

Acknowledgments

We credit Nina Raqueno and Tim Bauch of the RIT drone team, who captured the UAS imagery. This research is principally supported by Love Beets USA and New York Farm Viability Institute (NYFVI), in addition to National Science Foundation (NSF), Partnership for Innovation (PFI) award no. 1827551.

Reterences

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