



Unmanned Aerial System for Assessment of In-season Nitrogen uptake and Yield in Corn(*Zea mays L.*) Under Various Split N Applications

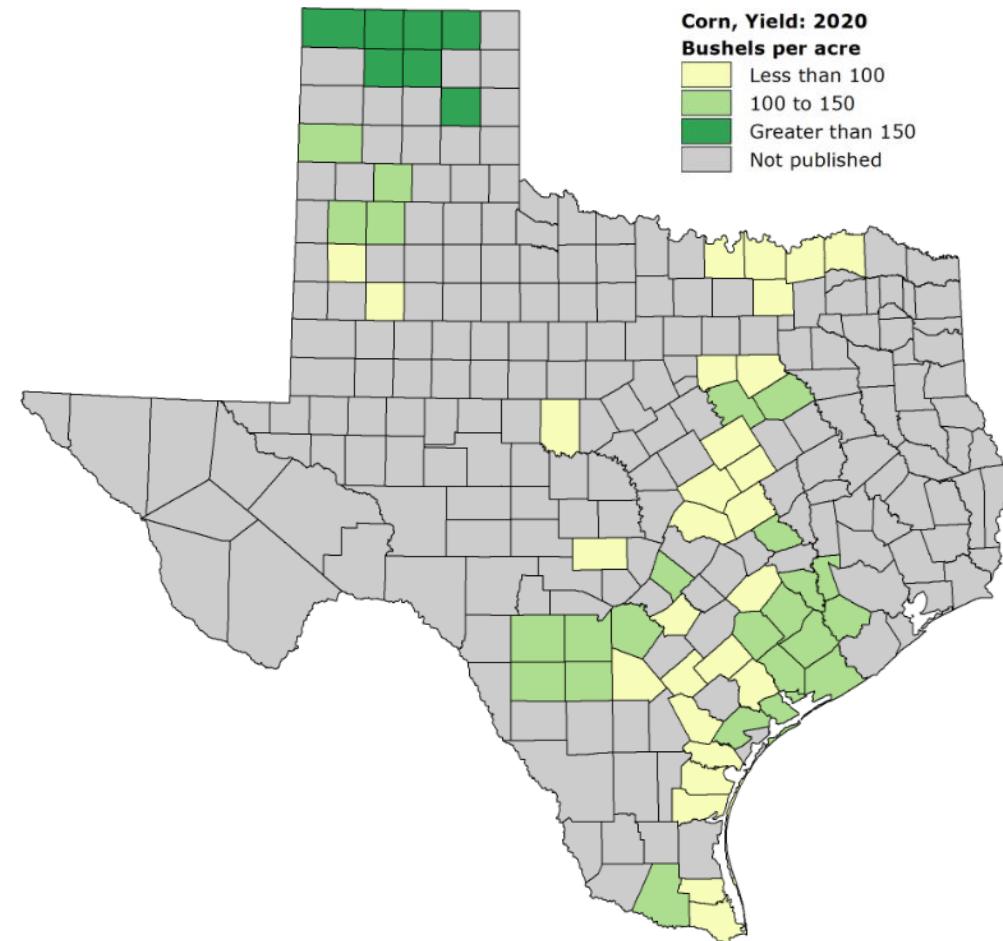
Kisman Bhattacharai, Nithya Rajan, Ronnie Schnell, Holli R. Archer, Pramod Pokharel, Chiranjibi Poudyal

Texas A&M University, College Station, Texas



Corn and Nitrogen (N)

- Corn is a major crop in the Texas High Plains and in Central and South Texas
- More than 2 million acres planted in 2021 ([NASS, 2021](#))
- 11.7 million metric tons harvested ([www.statista.com](#))
- 11 thousand metric tons of Nitrogen application
- the USA is the third largest consumer of Nitrogen(N) fertilizer



Statement of Problem

- Only 30 -50 % of applied N is utilized by crops (Omara et al., 2019)
- Nitrogen management is crucial for sustainable corn production
- Split N application has the potential to reduce N losses and help in N management
- Evaluation of the Split N application is important to derive insights

Index Suggests That Half of Nitrogen Applied to Crops Is Lost

Food production is becoming less efficient at using nitrogen fertilizer, according to a review of global values. Excess nitrogen damages the environment and the climate.

By Jenessa Duncombe 23 August 2021



Ad closed by Google

The power of nitrogen to keep crops healthy can be seen in these corn hybrids. The adult hybrid plants at left, from 2015, keep nitrogen in their leaves longer than the 1958 hybrids at right do. Credit: Tony Vyn

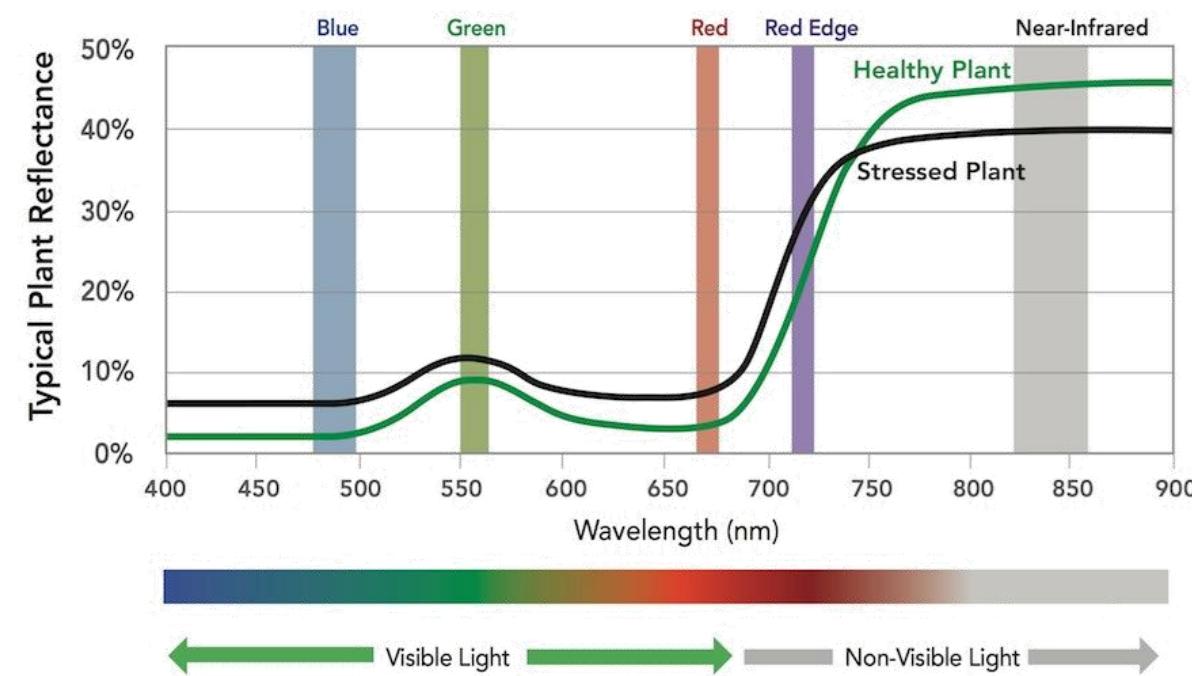
Statement of Problem

- Manual monitoring and assessment for large-scale production fields seems laborious and tedious
- Unmanned Aerial Vehicles (UAVs) can provide a great alternative to the manual assessment
 - Precise data and objective evaluation
 - Large scale assessment in less time



How does UAV help?

- UAV-derived spectral data are sensitive to crop stress ([L Zhang et., al 2019](#))
- UAV based remote sensing can predict yield ([Haghatalab, Atena, et al.2017](#))
- Spectral data can be utilized to study growth parameters



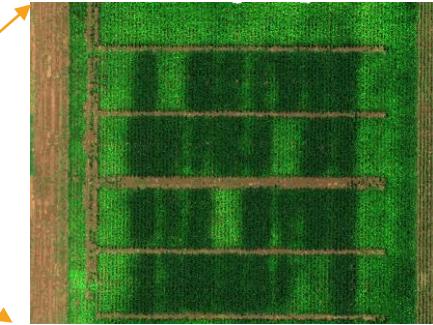
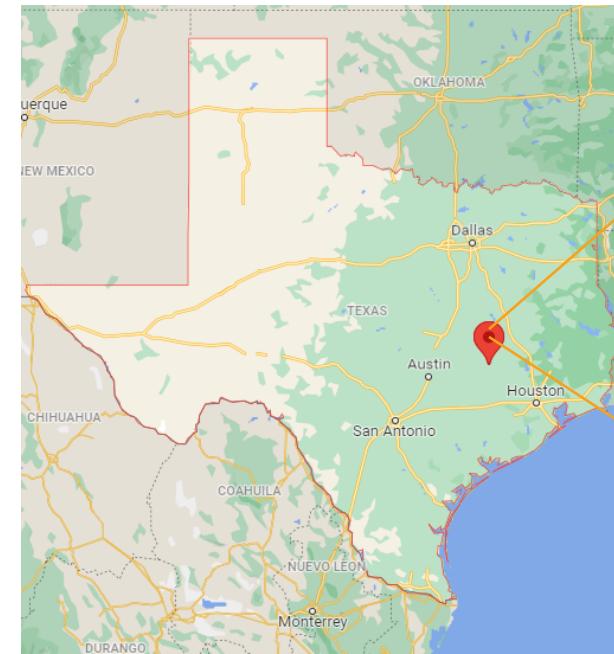
Objectives

1. Determine if remote sensing variables can determine the N-uptake in split N system
2. Evaluate the association of remote sensing variables with yield at different growth stages

Materials and Methods

Study site

- Soil Type: Clay loam(Fine, mixed, thermic, entic, hapluderts)
- Climate: Humid subtropical climate
- Corn growing season: Early March - Late July
- Residual Soil N: 5-7 ppm NO₃-N (0-30 cm depth)
- Two years of study (2021 & 2022)



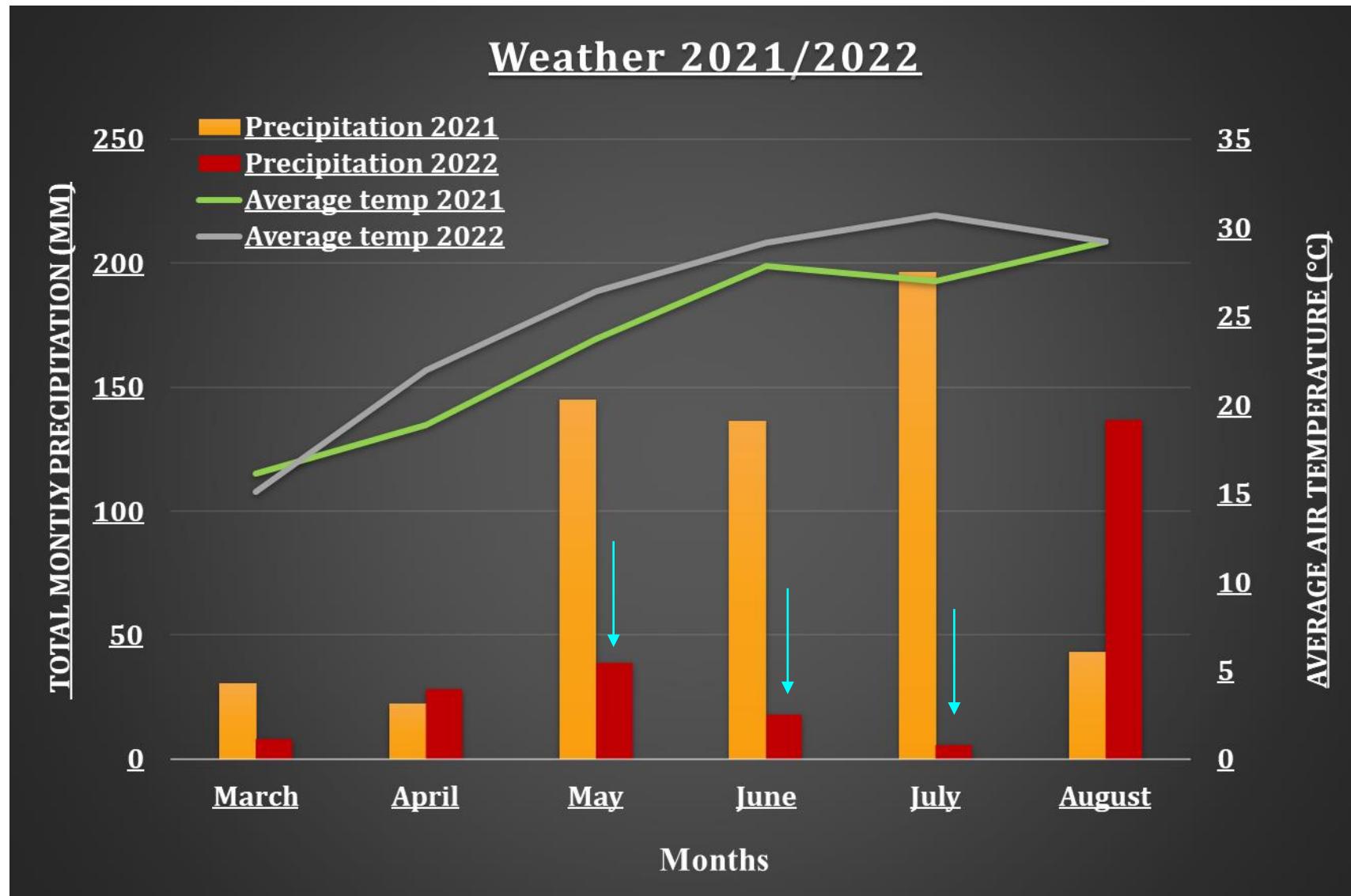
Texas A&M Agri-life Research Farm, College Station ,Texas
(30°32'28.7"N 96°25'41.4"W)

Materials and Methods: Experimental Setup

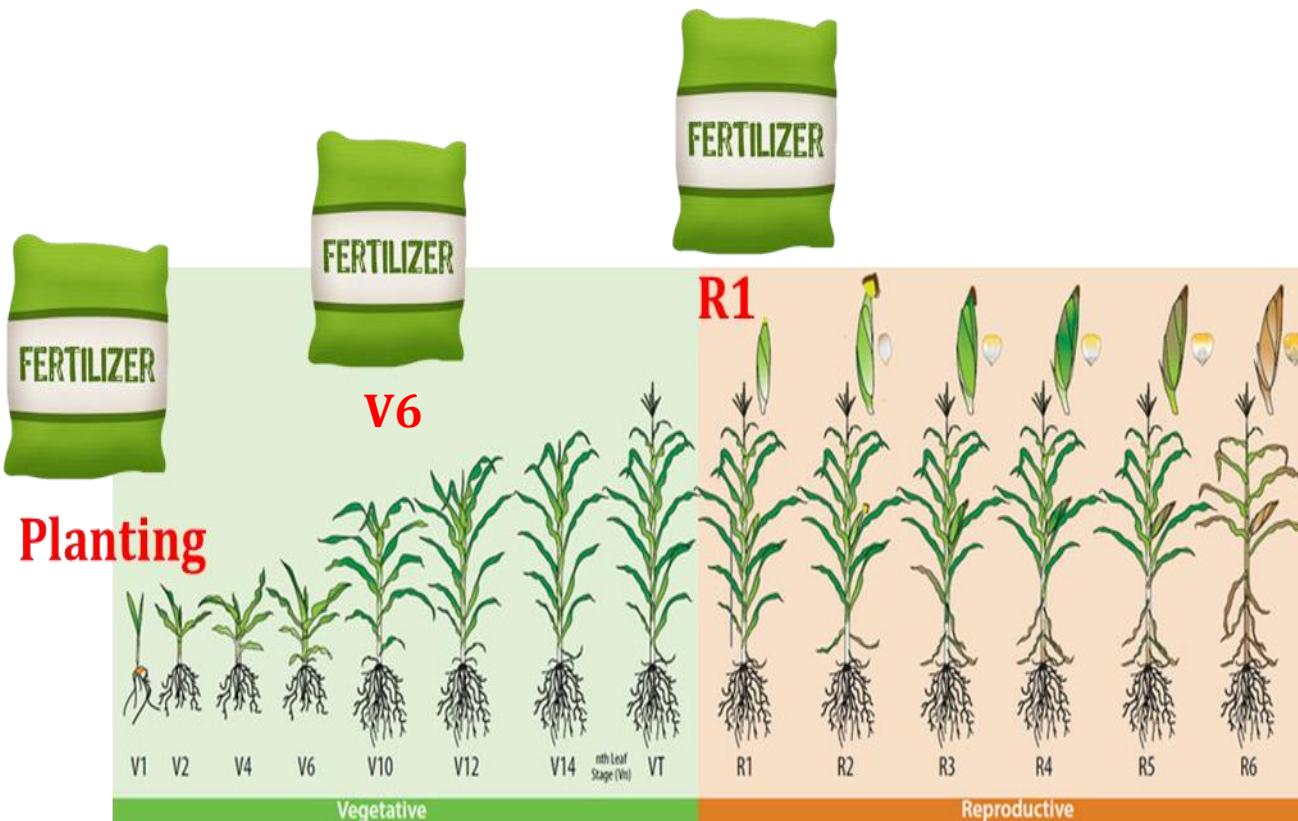
- Each plot: 13m x 6.5m
- RCBD design
- 8 treatments (7 treatments + 1 control) in 2021
- 10 treatments (2 treatments added in 2022)
- 4 replications
- Pioneer 1847 VHYR variety
- Rainfed irrigation management



Materials and Methods: Experimental Setup



Materials and Methods: Experimental Setup



N application rate - 240 kg/hectare

Split-treatment

	<u>First application</u>	<u>Second application</u>
T1-	0	0
T2-	20% (planting)	80% (V6)
T3:	100% (planting)	-----
T4:	90% (planting)	10% (R1)
T5:	80% (planting)	20% (R1)
T6:	70% (planting)	30% (R1)
T7:	60% (planting)	40% (R1)
T8:	50% (planting)	50% (R1)
T9:	50% (planting)	-----
T10:	25% (planting)	-----

Added treatments
for 2022

Unmanned Aerial Vehicle and Sensor

Data collection: UAV imagery



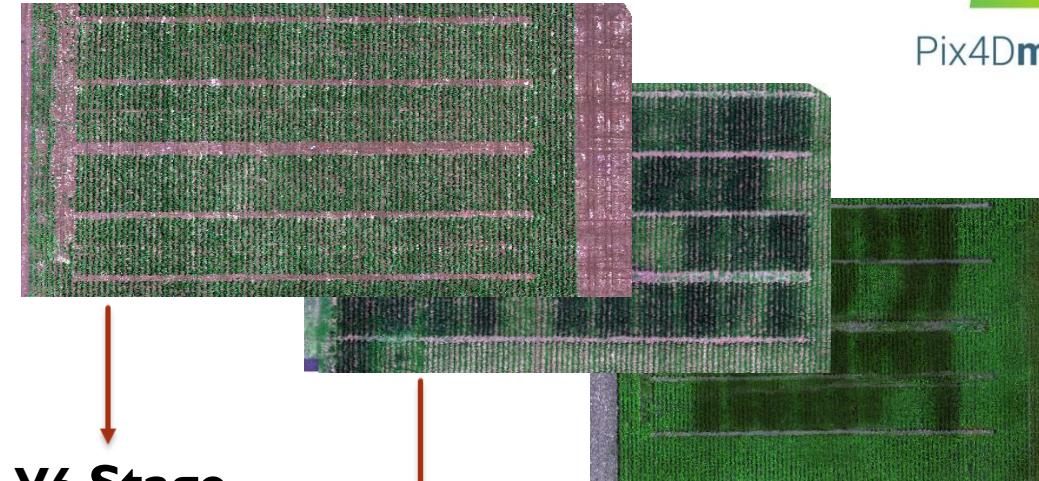
Hylio attached with Sentera 6X camera

- Five bands (Blue, Green, Red, NIR, Red-Edge)
- Altitude 15 m
- Image overlap: 80 %
- Image resolution: 0.8 cm/pixel

Generated ortho-mosaicked imagery

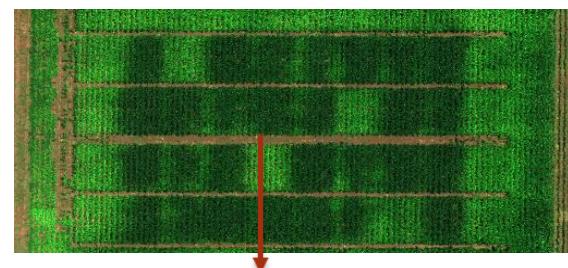


Pix4Dmapper



2021

V14 Stage



Added in 2022

Remote Sensing Data: Vegetation Indices

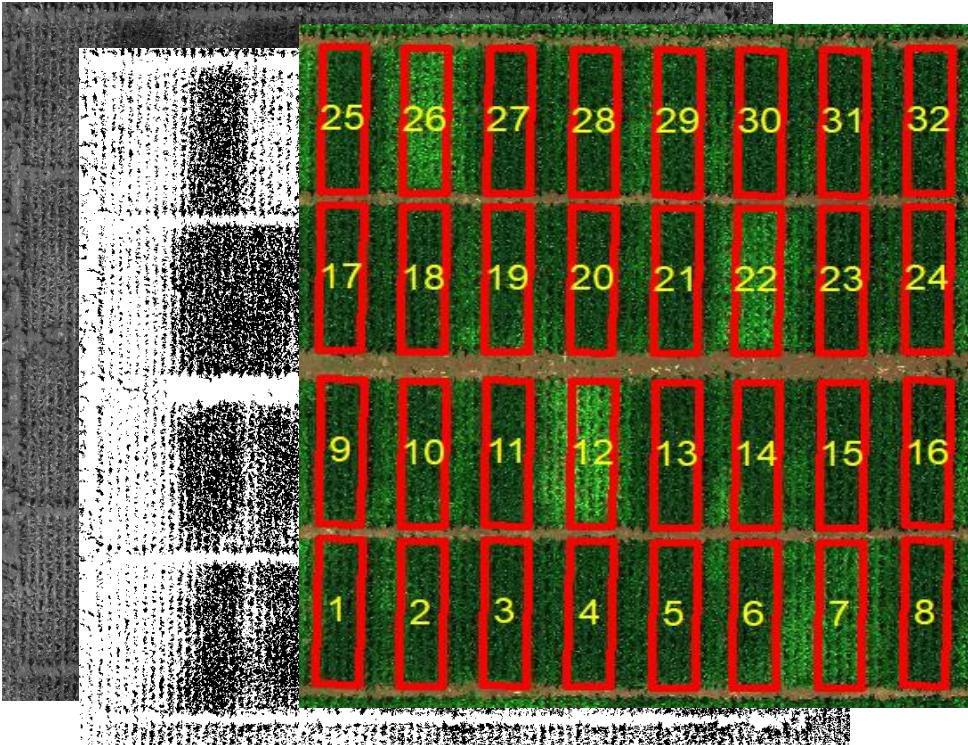
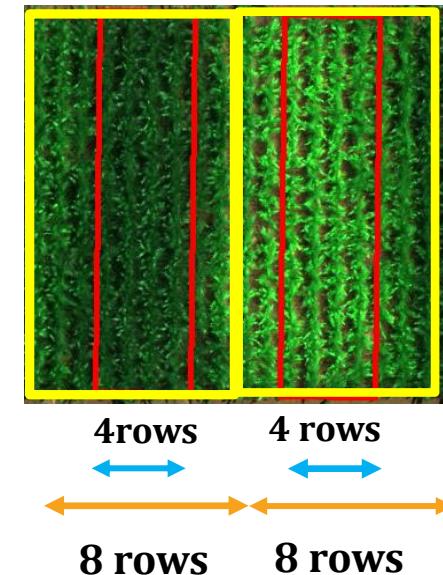


Image masking and NDVI & NDRE

NDVI: $(\text{NIR} - \text{Red}) / (\text{NIR} + \text{Red})$

NDRE: $(\text{NIR} - \text{RedEdge}) / (\text{NIR} + \text{RedEdge})$



To reduce boarder effect

Table

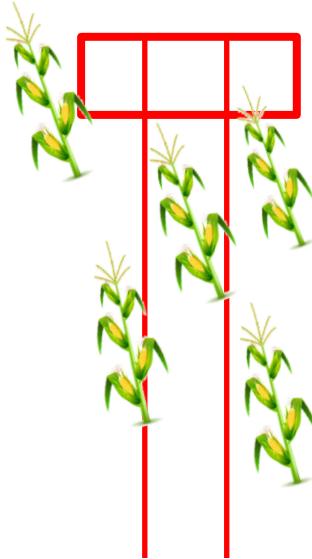
ZonalSt_NDRE_VG

OBJECTID *	Id	COUNT	AREA	MIN	MAX	RANGE	MEAN	STD	SUM
1	1	33638	15.956804	0.270661	0.660765	0.390105	0.474835	0.068703	15972.493561
2	2	33065	15.684991	0.270019	0.631054	0.361035	0.445017	0.07793	14714.502263
3	3	32669	15.497141	0.270047	0.628772	0.358725	0.442522	0.074959	14456.736019
4	4	33077	15.690684	0.27002	0.63098	0.360961	0.436642	0.072073	14442.793723
5	5	33529	15.905098	0.270029	0.637089	0.367059	0.469639	0.075827	15746.516334
6	6	33435	15.860507	0.270032	0.604638	0.334606	0.429837	0.06929	14371.606126
7	7	19446	9.224568	0.270002	0.490798	0.220796	0.339382	0.044706	6599.620406
8	8	19427	9.215555	0.270022	0.477076	0.207054	0.332926	0.043672	6467.759449

VI extraction

Materials and Methods

Data collection : Leaf, Stem and Grain tissue for N analysis



5 representative plants were cut and separated

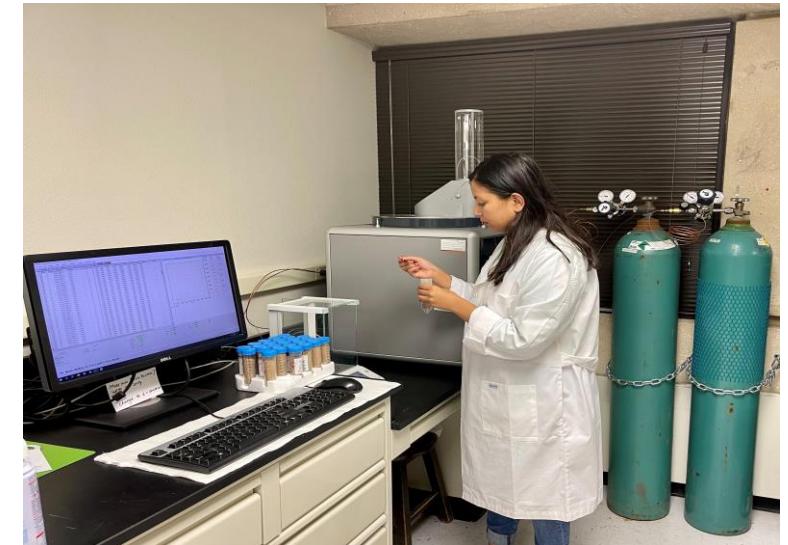
Sample oven dried: 60° C



Ground sample for N analysis

250 mg of sample of N analysis

Total above-ground N uptake
(kg per ha)

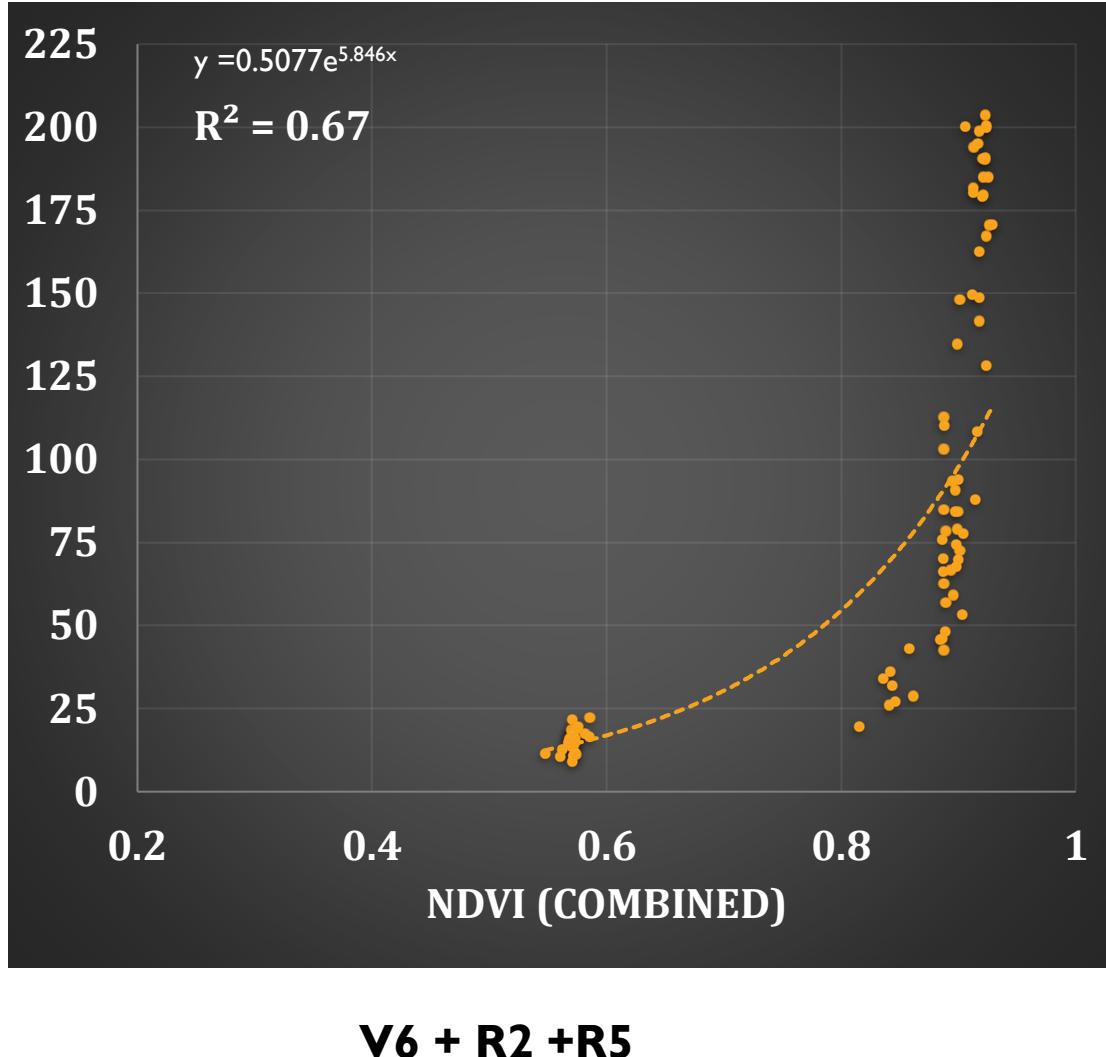


Nitrogen analysis - VARIO MAX
CUBE ELEMENTAR

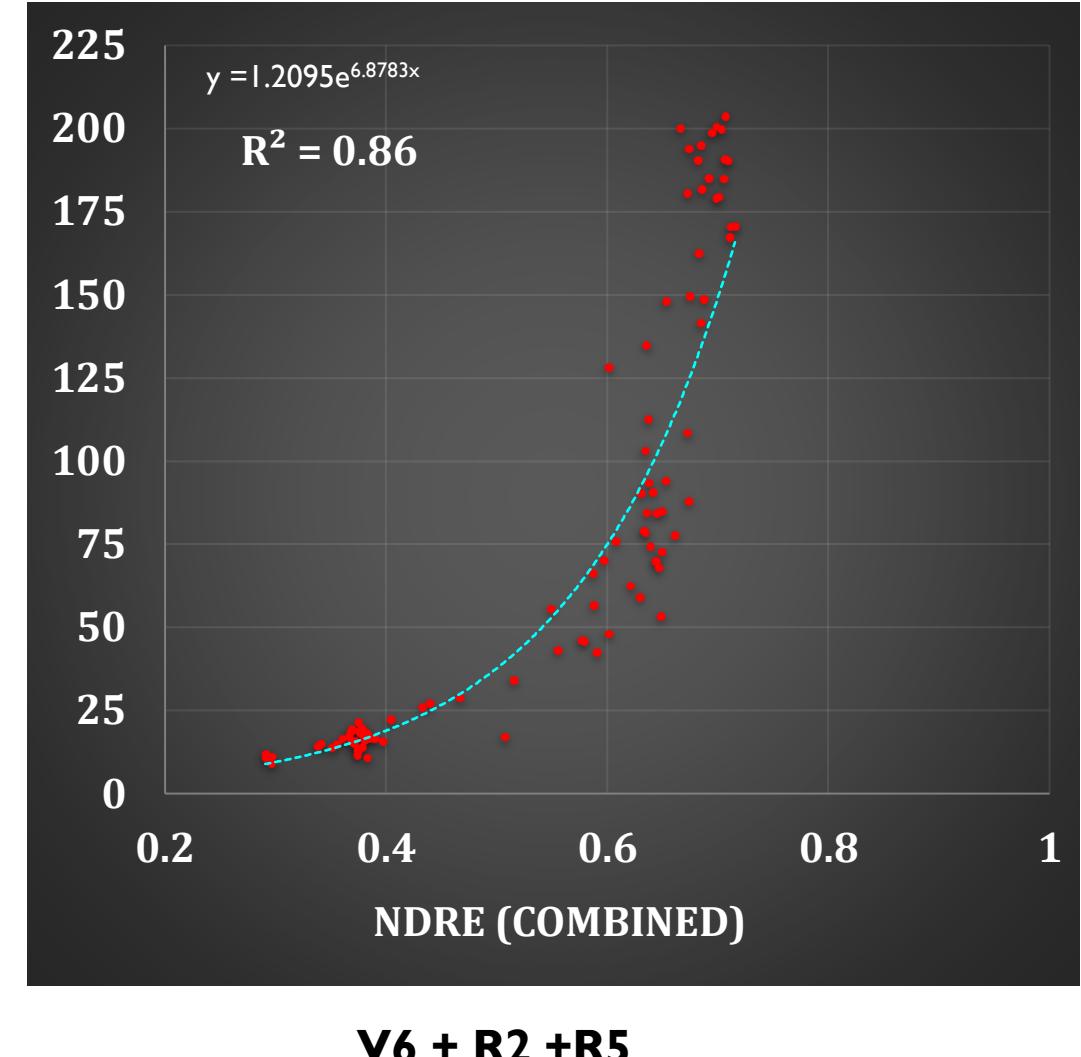
Result and Discussion

Regression Analysis – Plant N uptake & vegetation Indices -2021

Total above-ground plant N uptake (kg ha^{-1})



V6 + R2 +R5

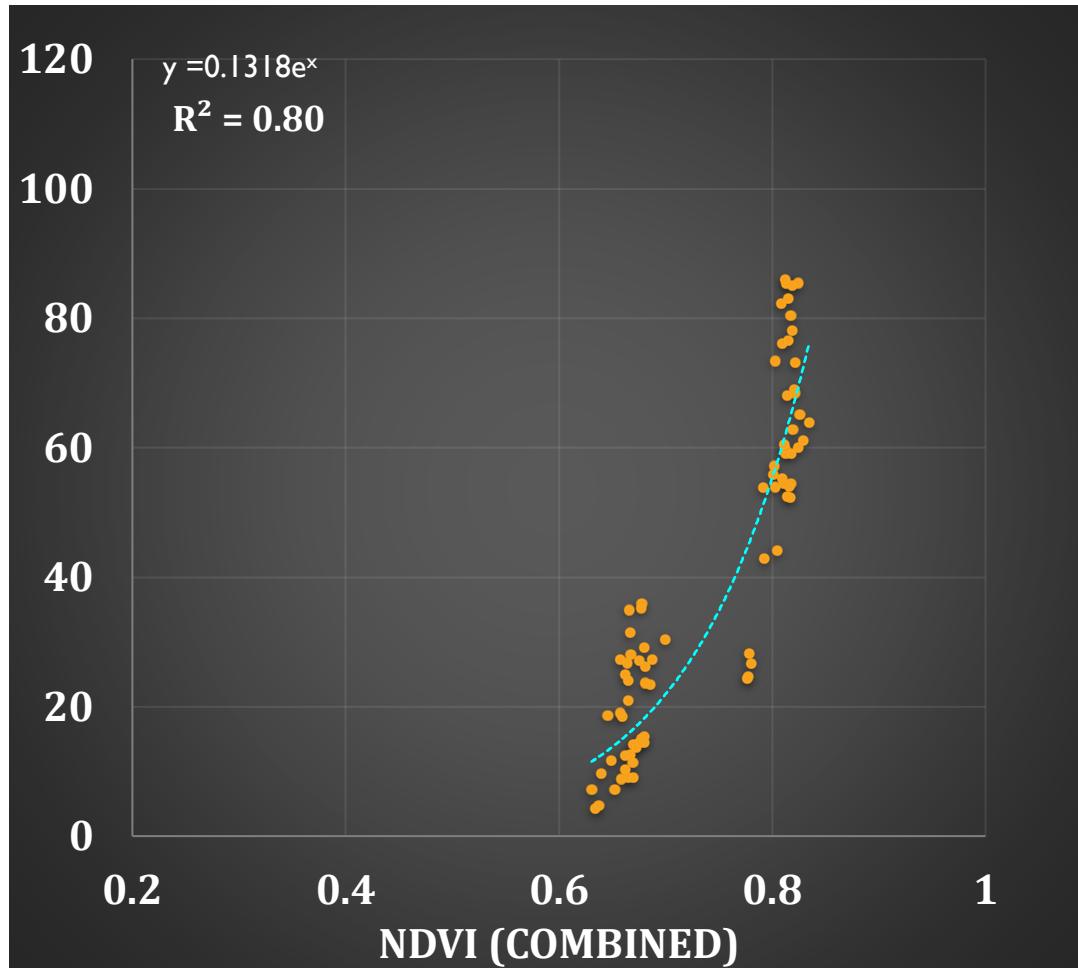


V6 + R2 +R5

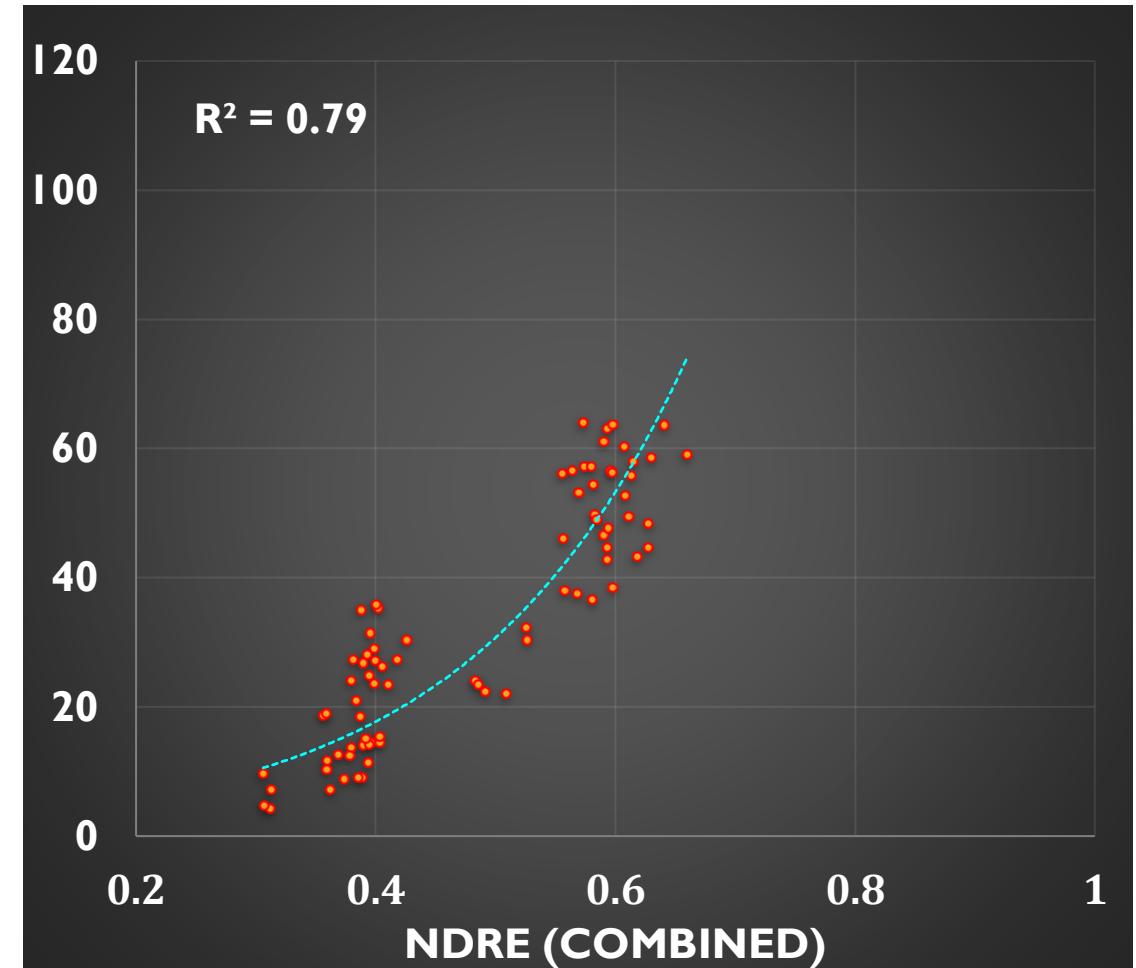
Result and Discussion

Regression - Plant N uptake & vegetation Indices -2022

Total above-ground plant N uptake (kg per ha)

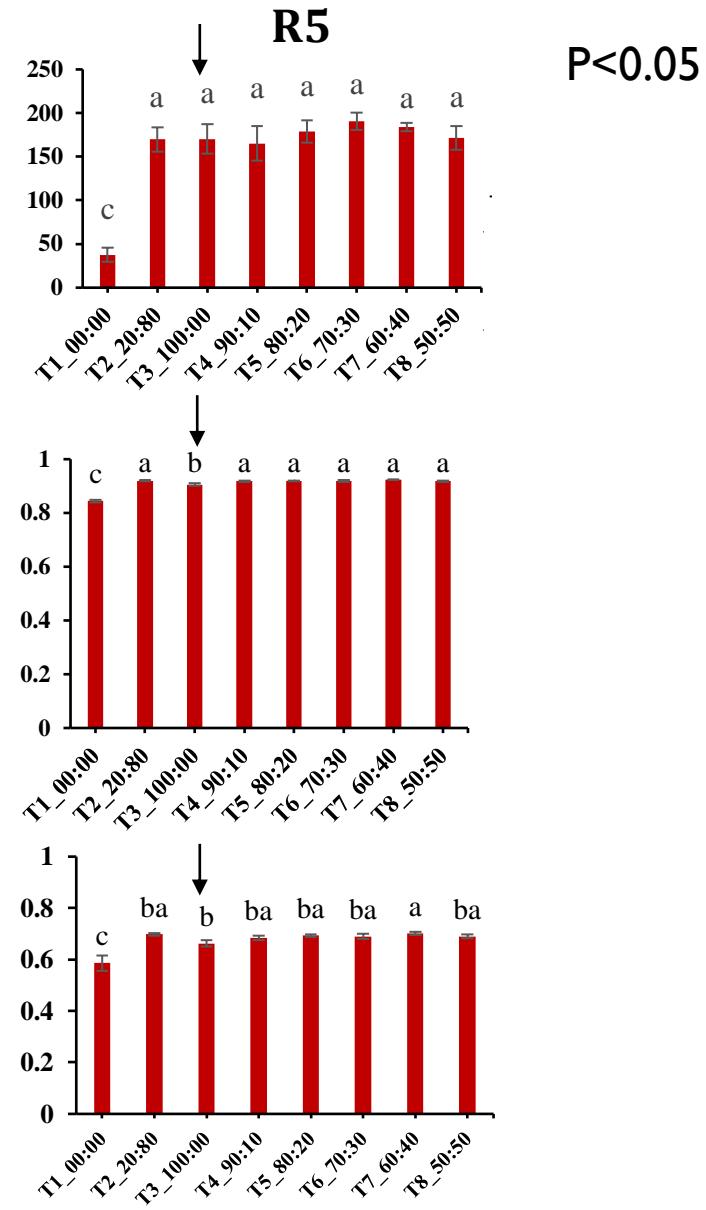
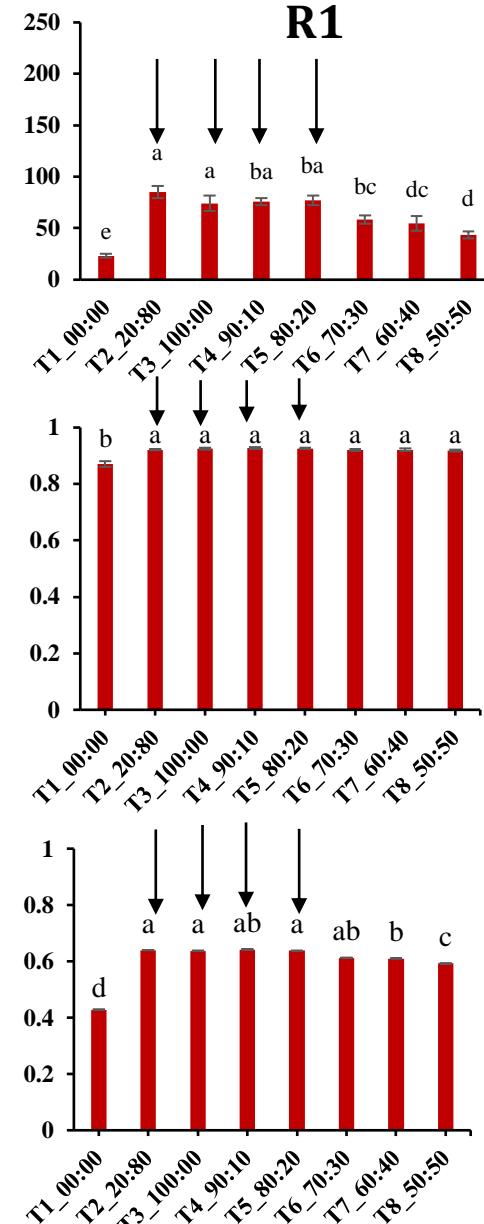
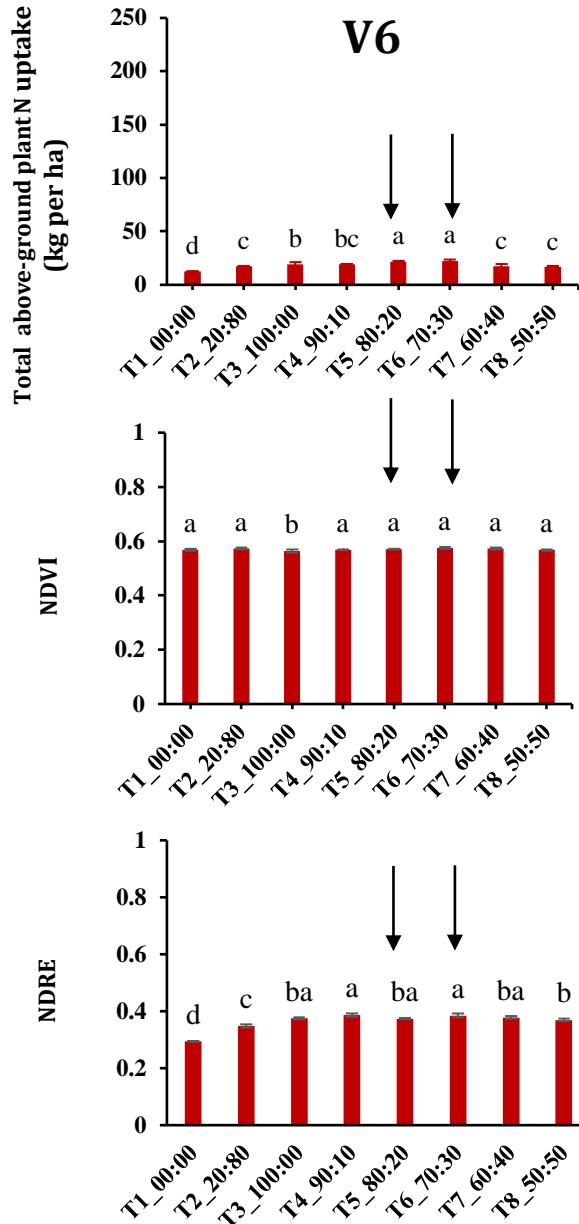


V6+V14

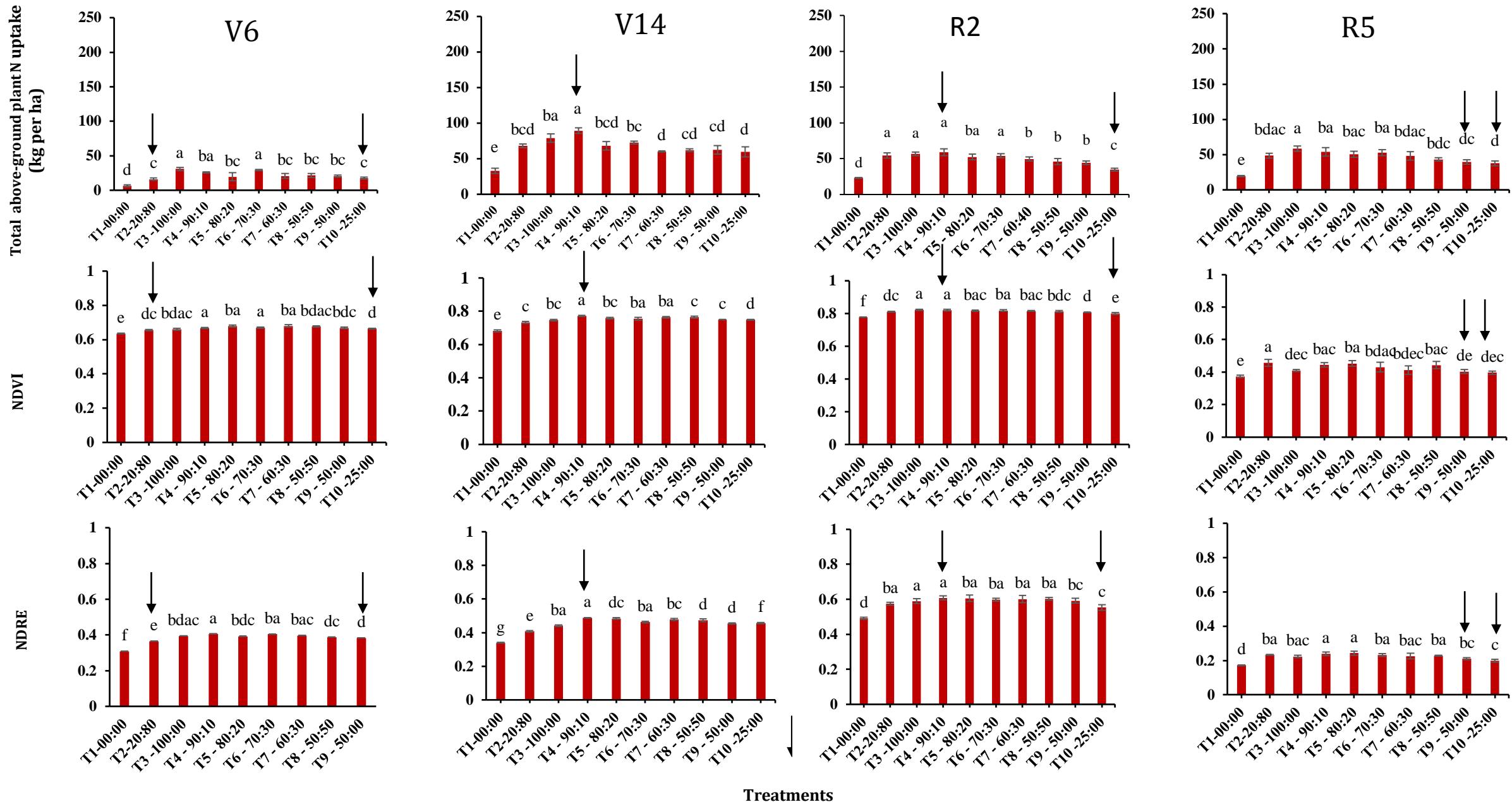


V6+V14

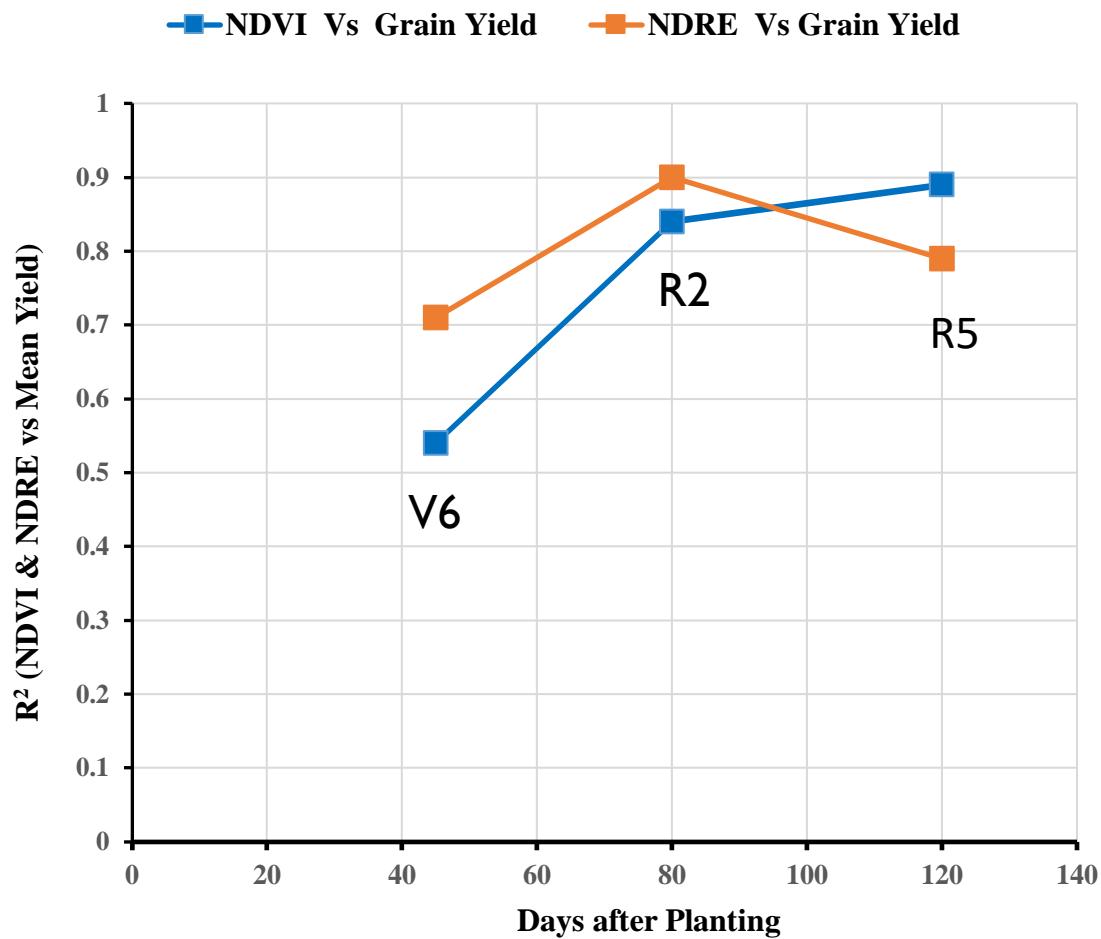
Vegetation Indices response to plant N uptake -2021



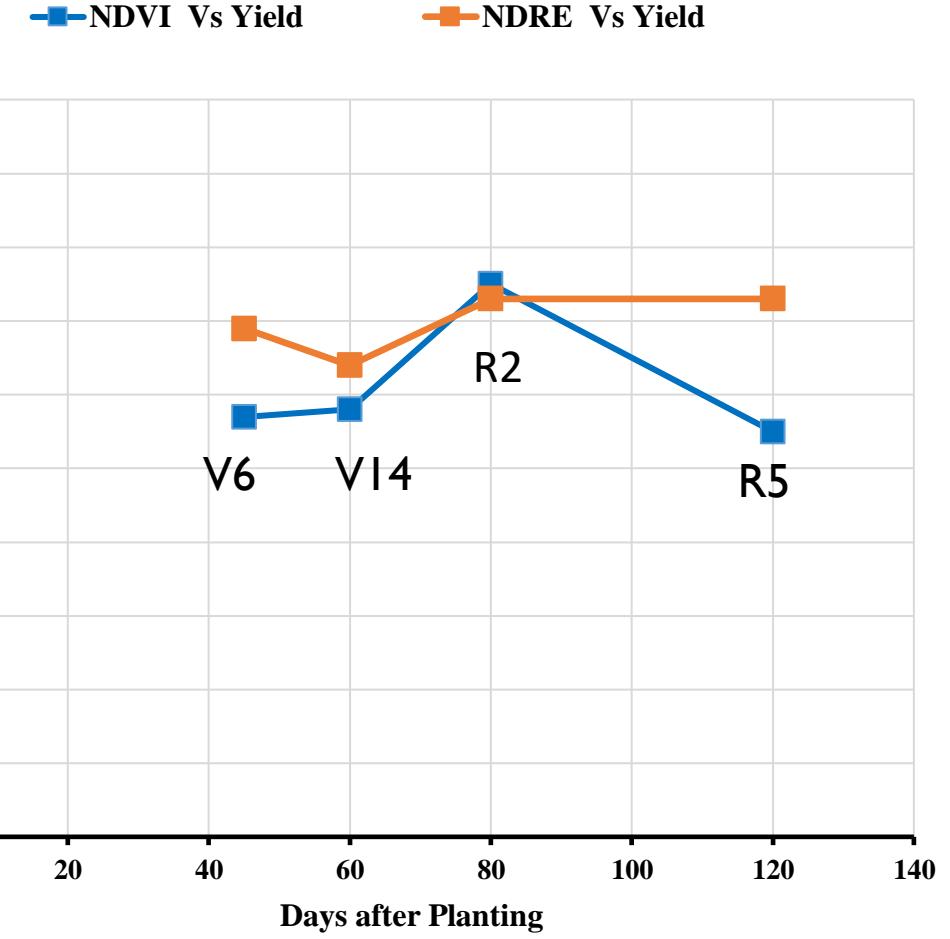
Vegetation Indices response to plant N uptake -2022



Result and Discussion



2021



2022

Conclusions

Non-drought

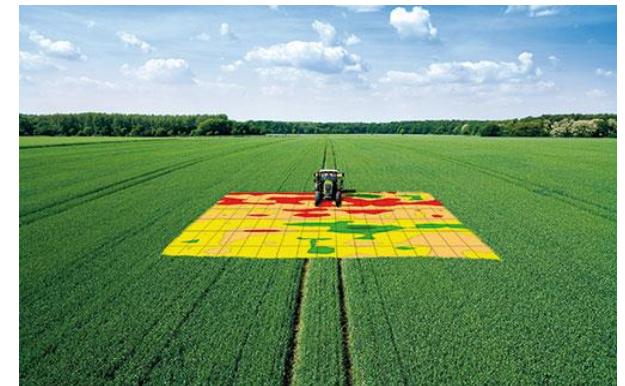
- NDRE correlated better than NDVI to yield before 80 DAP

Drought

- NDVI correlated slightly better compared to NDRE
- NDRE correlated better than NDVI to yield before 80 DAP

Main Takeaway:

- UAV-based remote sensing can be an alternative means of Split N assessment, especially under non-drought conditions



Future Direction

- Validate the vegetation index – N uptake estimation model at the field level
- Multiple remote sensing variables can be used combinedly via Machine Learning Regression techniques for better N-uptake estimation

Acknowledgements



Crop Physiology and Agroecology Lab
Soil and Crop Science Department

Advisor committee
Dr. Nithya Rajan
Dr. Ronnie Schnell
Dr. Holli Archer



Funded by
Texas A&M Traids for
Transformation project(T3)

Acknowledgement

Thank you!!

Questions?