

Strawberry Nutrient Management: Potassium Rates and AI-Driven Diagnosis

Shinsuke Agehara

Associate Professor, Horticultural Crop Physiology

Gulf Coast Research and Education Center, University of Florida, IFAS

813-419-6583 | sagehara@ufl.edu



Presentation outline

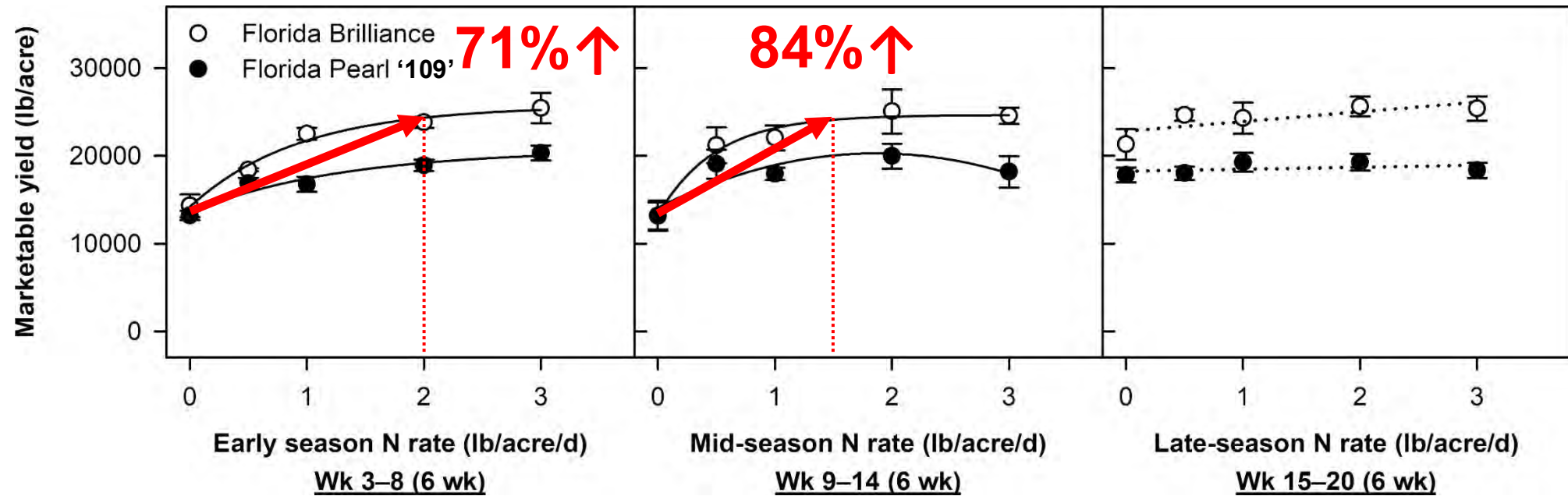
Potassium rate



**AI-Driven
Plant Diagnosis**

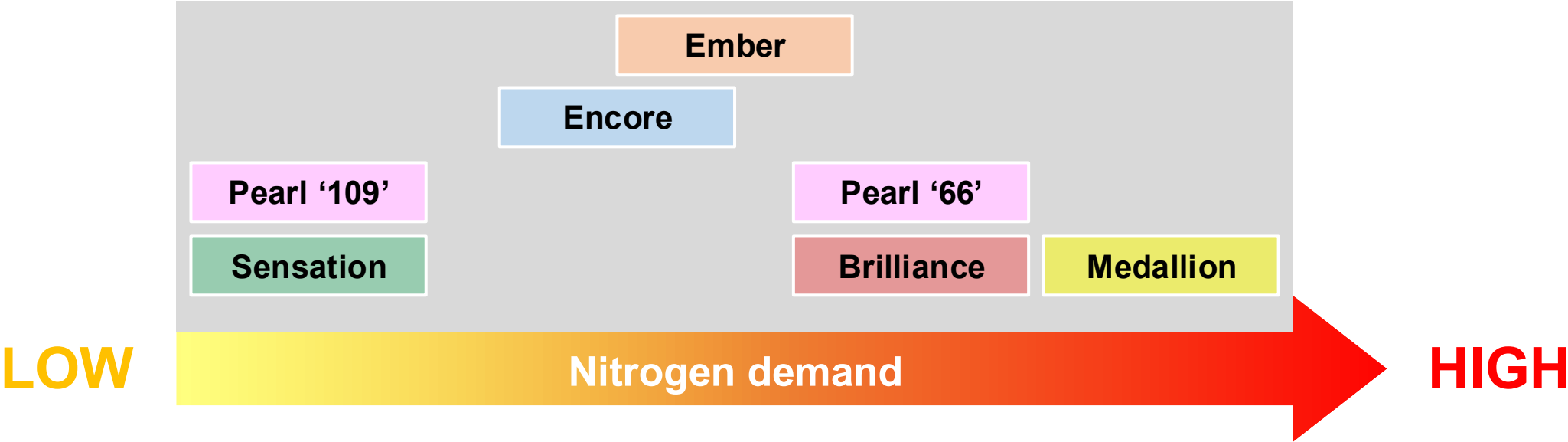


Yield responses to N



Nitrogen recommendation

	Preplant (lb/acre)	Daily N application rate (lb/acre/d)							Total (lb/acre)
		Oct	Nov	Dec	Jan	Feb	Mar	Apr	
Old	0–40	0.6	0.6	0.6	0.6	0.75	0.75	0.6	150
New (2023)	0	1.5–2.0	1.0–2.0	1.0–1.5	0.75–1.0	0.5–1.0	0.5–0.75	0.5–0.75	175



- 1

Mehlich-3 potassium	K ₂ O recommendation
Low (<35 ppm)	120–150 lb/acre
Medium (35–60 ppm)	100 lb/acre
High (>60 ppm)	0 lb/acre

- 3**

If soil test K is low (<35 ppm)...

[illegible]

Functions of potassium in plants

- K is present as K^+ ion in plants.
- Physiological roles
 - Stomatal opening and closure
 - Cell expansion and turgor maintenance
 - Sugar transport from source (leaves) to sink (fruit)
 - Activation of enzymes in sugar metabolism and photosynthesis

Potassium rate trials

Year

- 2024–25
- 2025–26

Cultivar

- Brilliance
- Medallion
- Ember
- Encor

Background soil K

- 19–28 ppm (low)

Trial 1 Early season (6 wk)

#	Daily K ₂ O application rate (lb/acre/d)					Total K ₂ O (lb/acre)
	Oct	Nov	Dec	Jan	Feb	
1	Sprinkler	0.0	1.0	1.0	1.0	84
2		1.0	1.0	1.0	1.0	126
3		2.0	1.0	1.0	1.0	168
4		3.0	1.0	1.0	1.0	210
5		4.0	1.0	1.0	1.0	252

Trial 2 Mid-season (6 wk)

#	Daily K ₂ O application rate (lb/acre/d)					Total K ₂ O (lb/acre)
	Oct	Nov	Dec	Jan	Feb	
1	Sprinkler	1.0	0.0	1.0	1.0	84
2		1.0	1.0	1.0	1.0	126
3		1.0	2.0	1.0	1.0	168
4		1.0	3.0	1.0	1.0	210
5		1.0	4.0	1.0	1.0	252

Trial 3 Late season (6 wk)

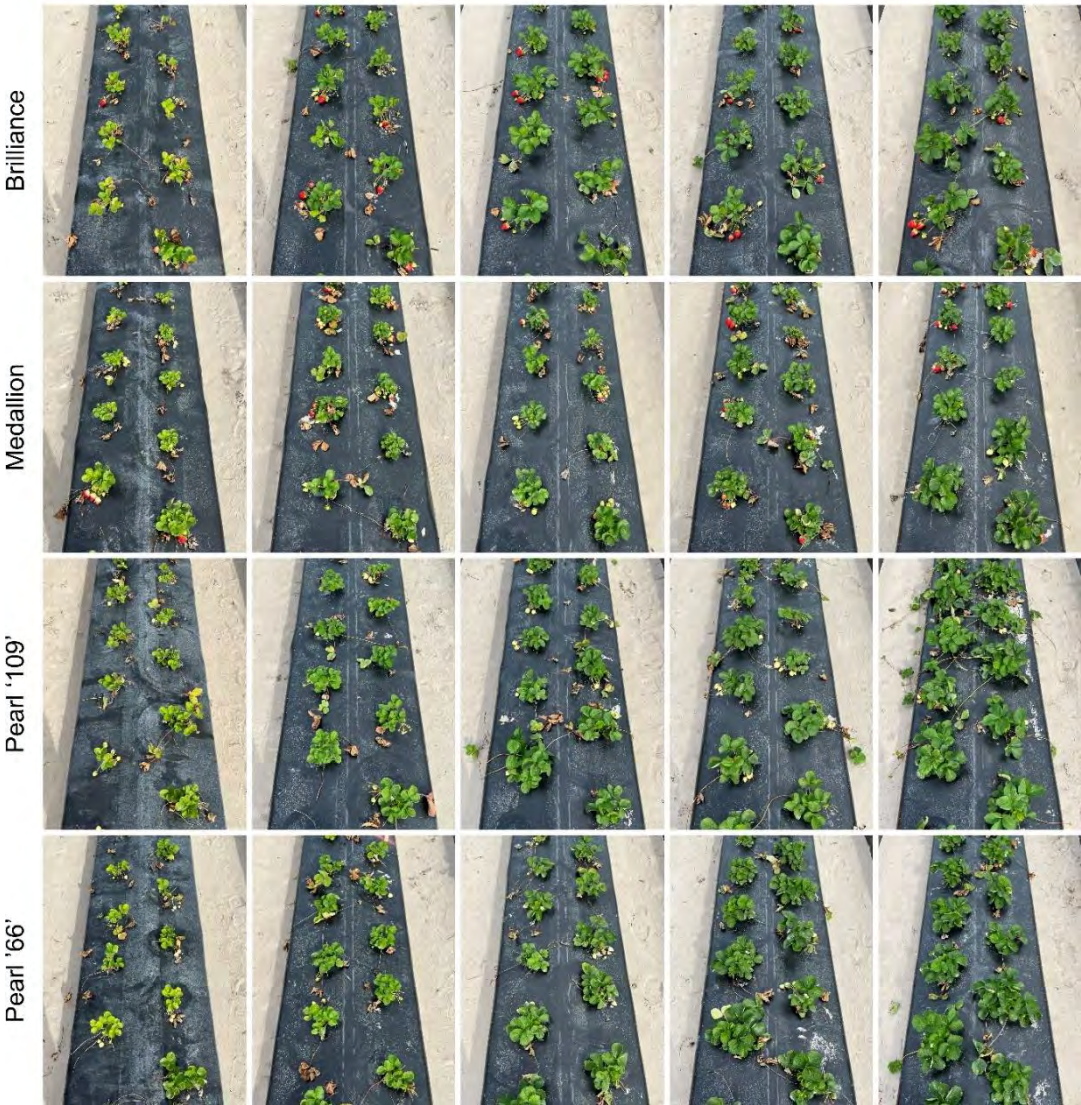
#	Daily K ₂ O application rate (lb/acre/d)					Total K ₂ O (lb/acre)
	Oct	Nov	Dec	Jan	Feb	
1	Sprinkler	1.0	1.0	0.0	1.0	84
2		1.0	1.0	1.0	1.0	126
3		1.0	1.0	2.0	1.0	168
4		1.0	1.0	3.0	1.0	210
5		1.0	1.0	4.0	1.0	252

Nitrogen trial

December 20, 2024 (65 days after planting)

Nitrogen application rate (lb/acre/d)

0 0.5 1 2 3

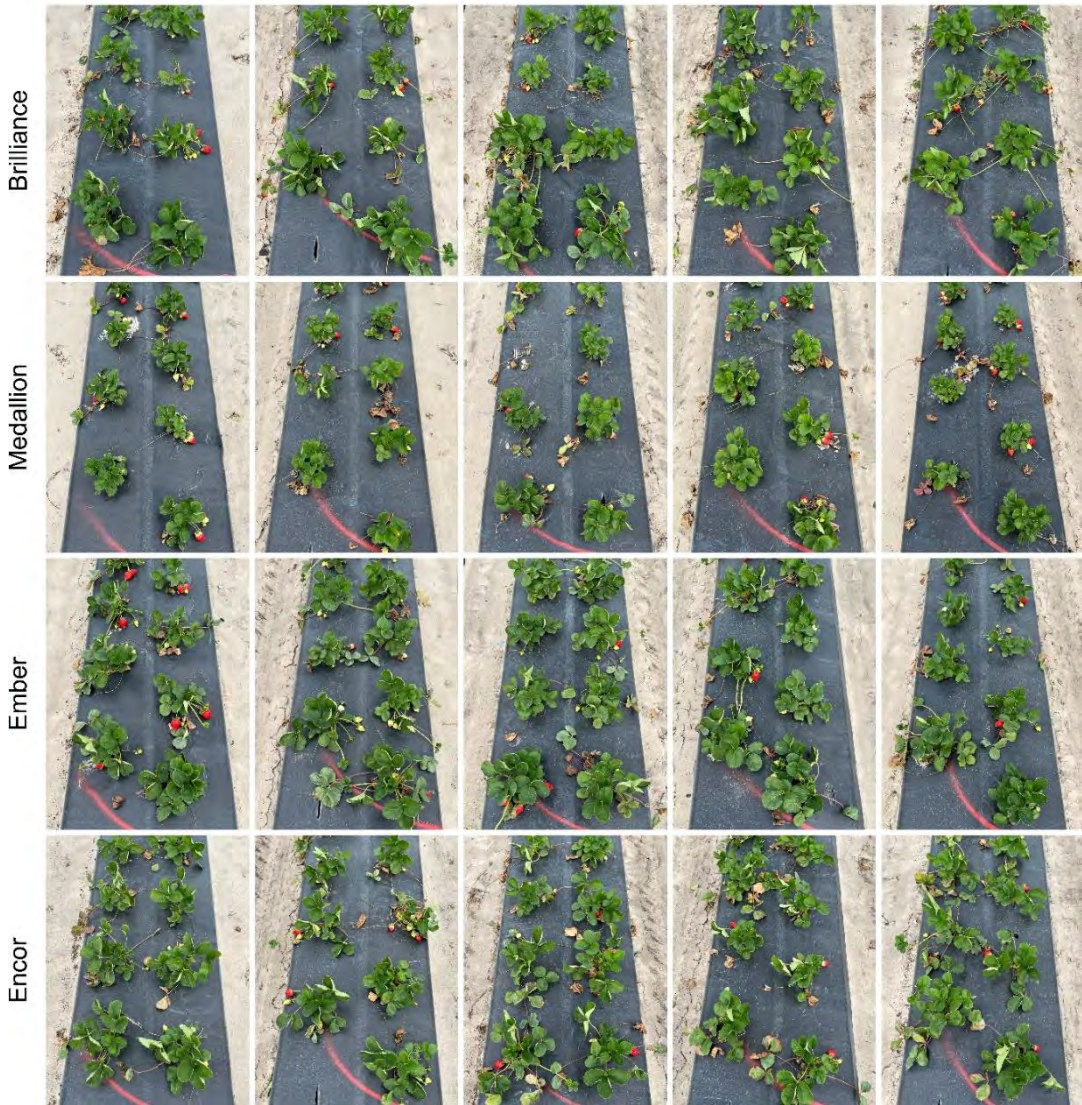


Potassium trial

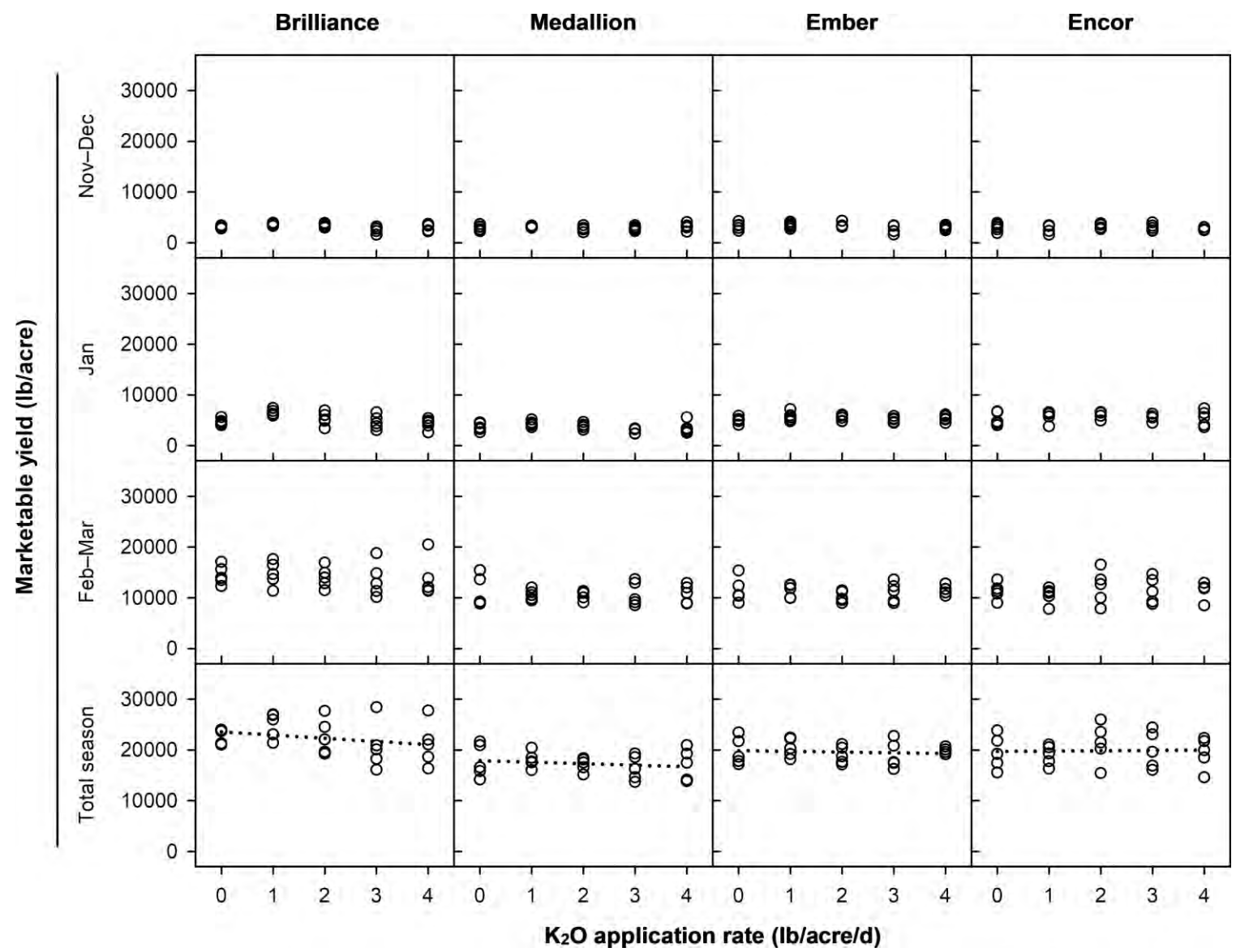
December 14, 2025 (60 days after planting)

Potash (K_2O) application rate (lb/acre/d)

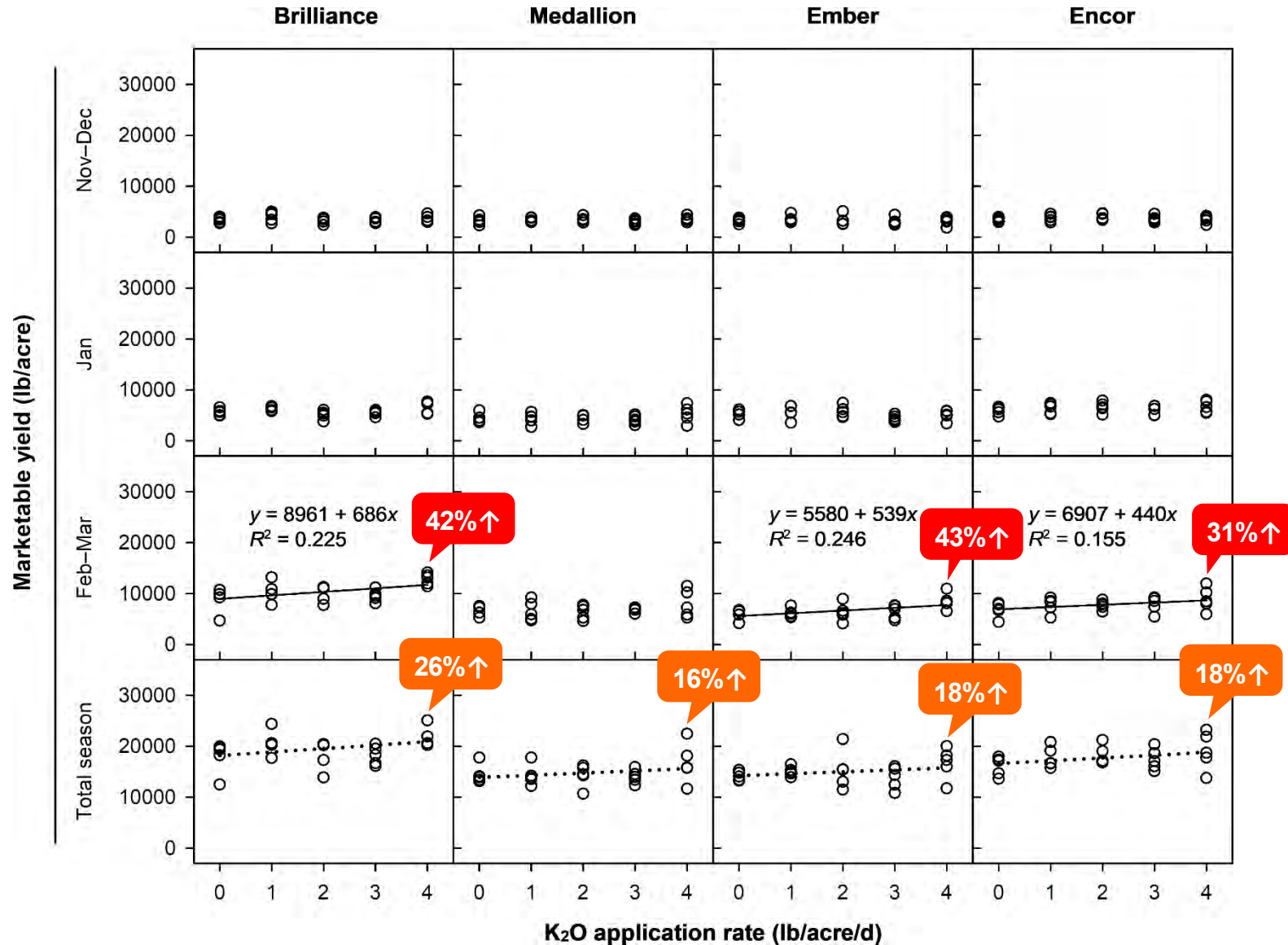
0 1 2 3 4



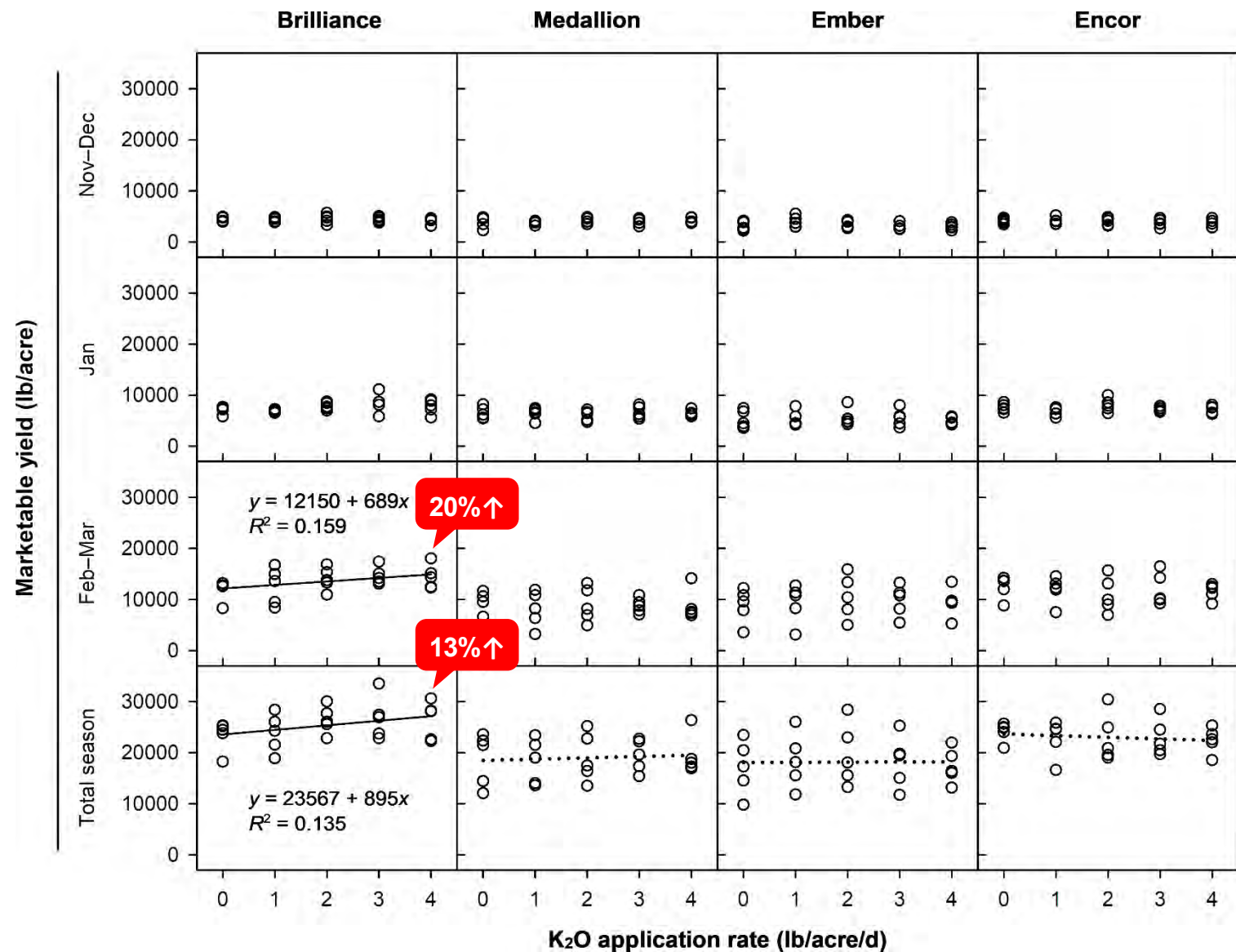
Year 1: Yield responses to early-season K



Year 1: Yield responses to mid-season K

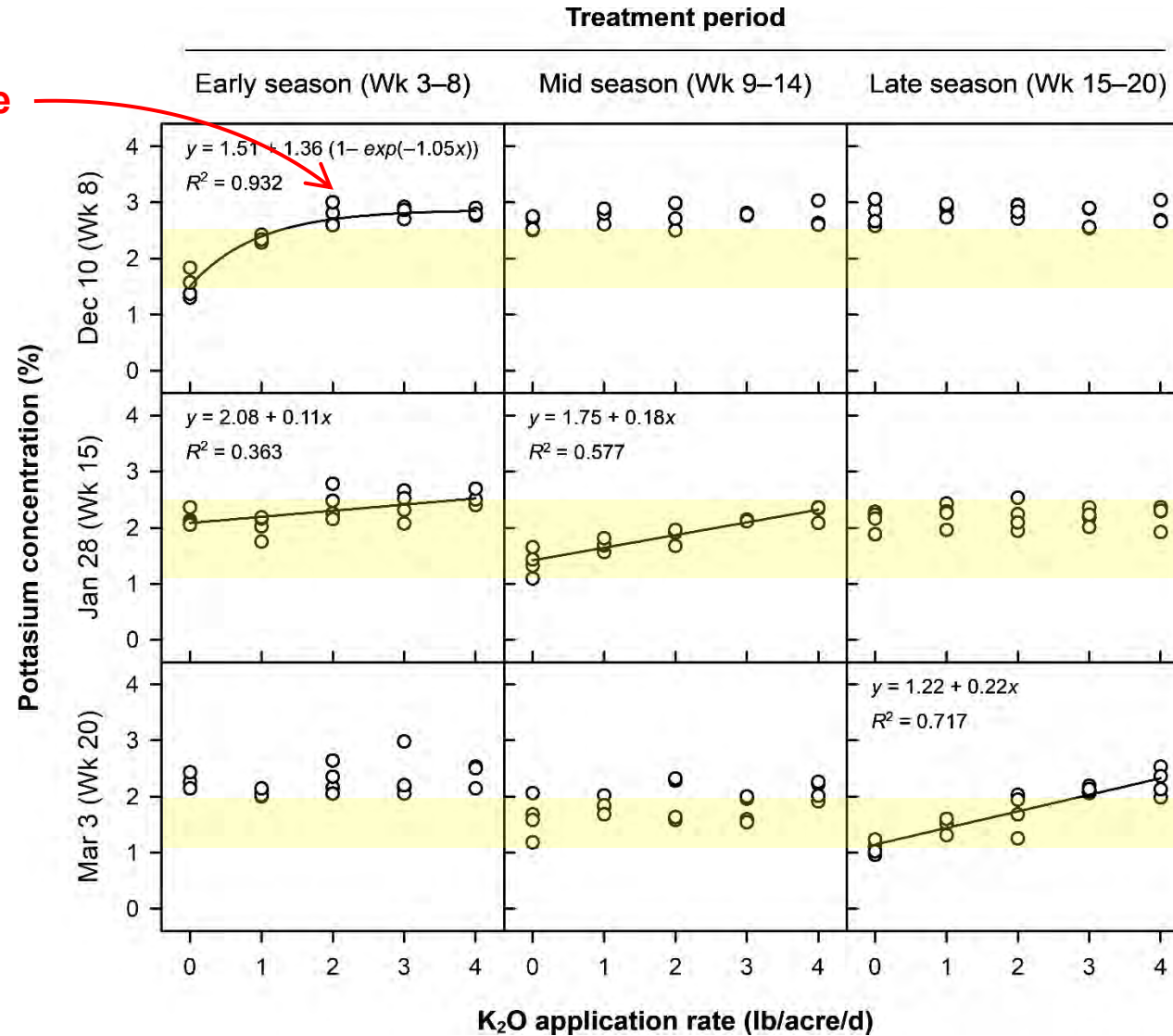


Year 1: Yield responses to late-season K

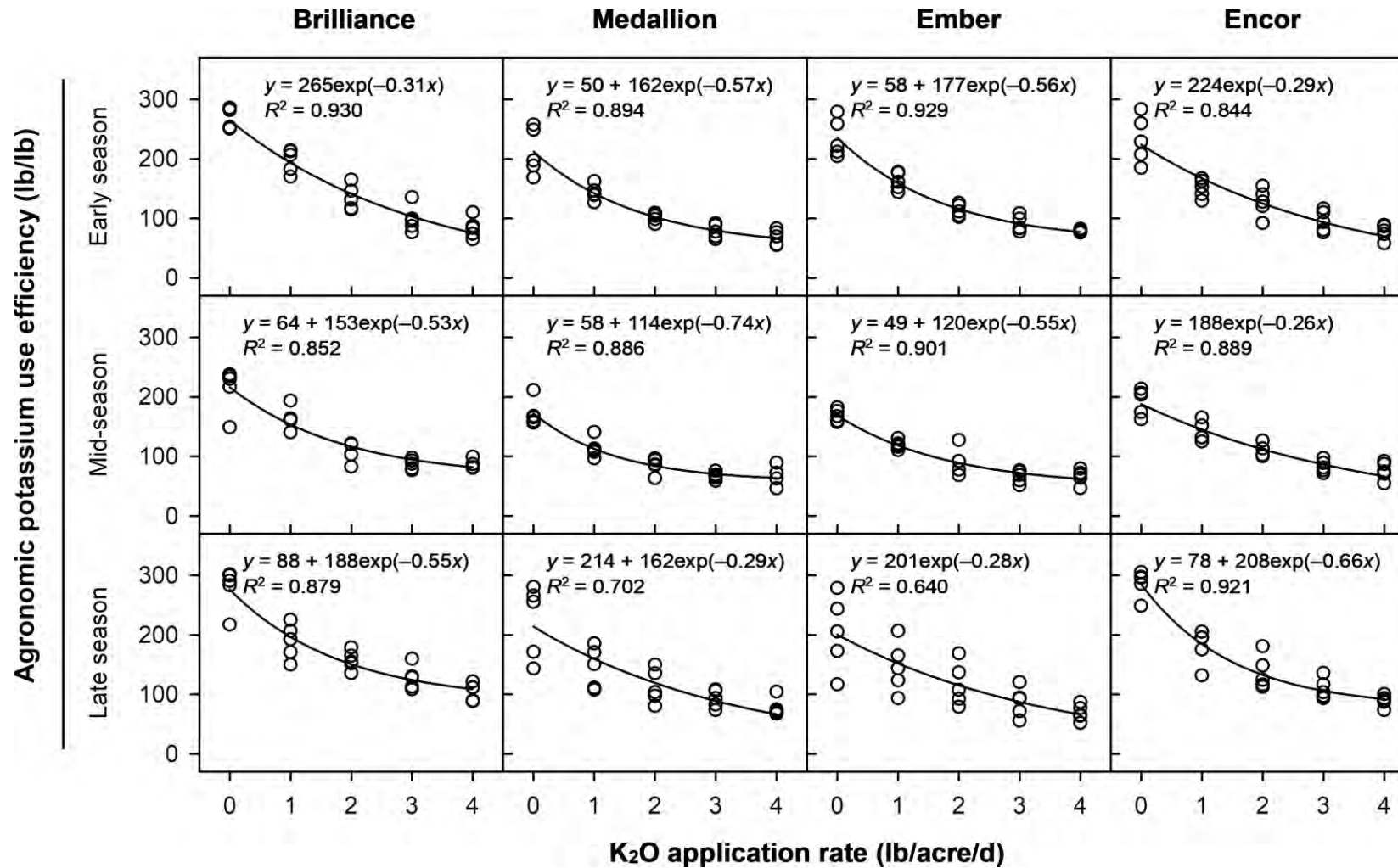


Year 1: Leaf K concentration (Brilliance)

Luxury K uptake

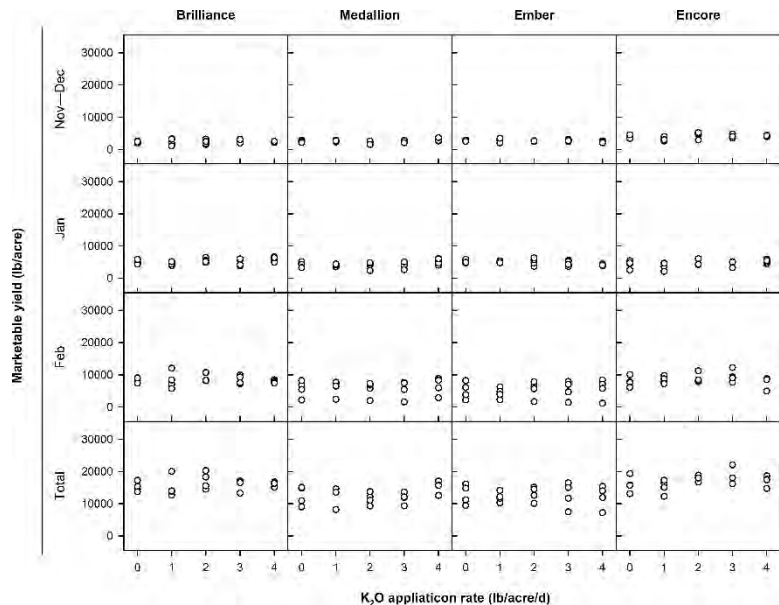
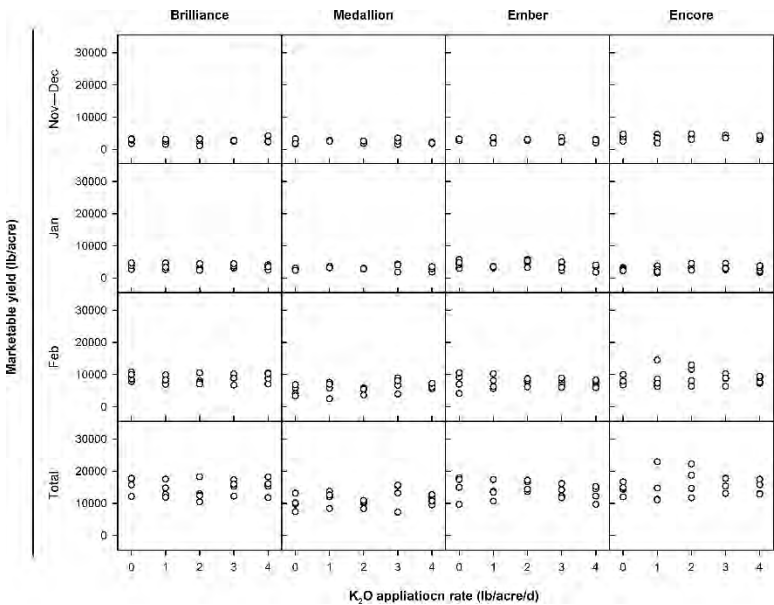
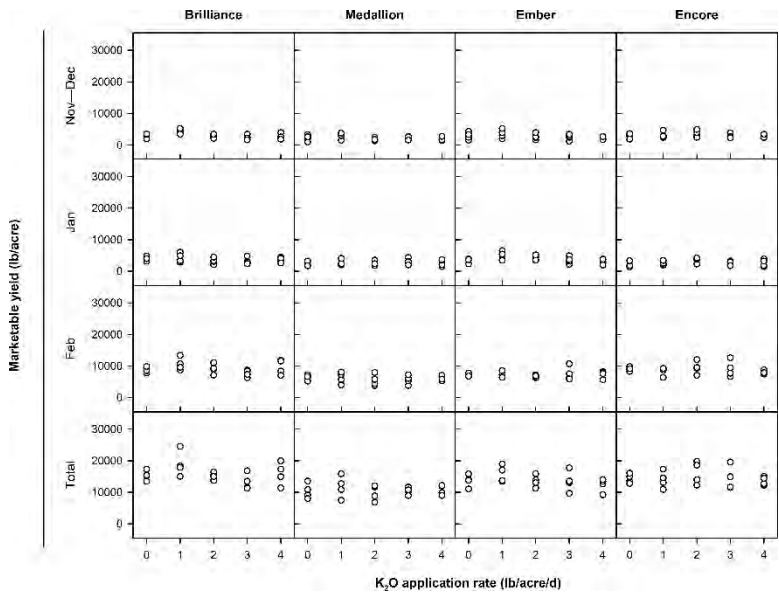


Year 1: Agronomic K use efficiency



Year 2: Yield

No significant response to K rates in all trials



Summary

- Plants are more responsive to K applied during **mid- and late-growth stages**.
 - Vegetative growth = Sink K demand ↓
 - Fruit load ↑ = Sink K demand ↑
- Yield response to K rate: **Brilliance > Ember = Encor > Medallion**.
- K fertilization had **minimal effects on fruit size and Brix**.
- Yield gains by K fertilization are much smaller than N = **low agronomic K use efficiency**.
- **Adjust fertilizer formulation (N:K₂O ratio)** to match seasonal N and K demand.
 - Early season = 5 N : 1–2 K₂O
 - Mid to late season = 5 N : 6–8 K₂O



Presentation outline

Potassium rate



AI-Driven
Plant Diagnosis



Nutrient deficiency symptoms

Nutrient deficiency

Healthy



Nitrogen



Phosphorus



Potassium



Salinity stress



Calcium



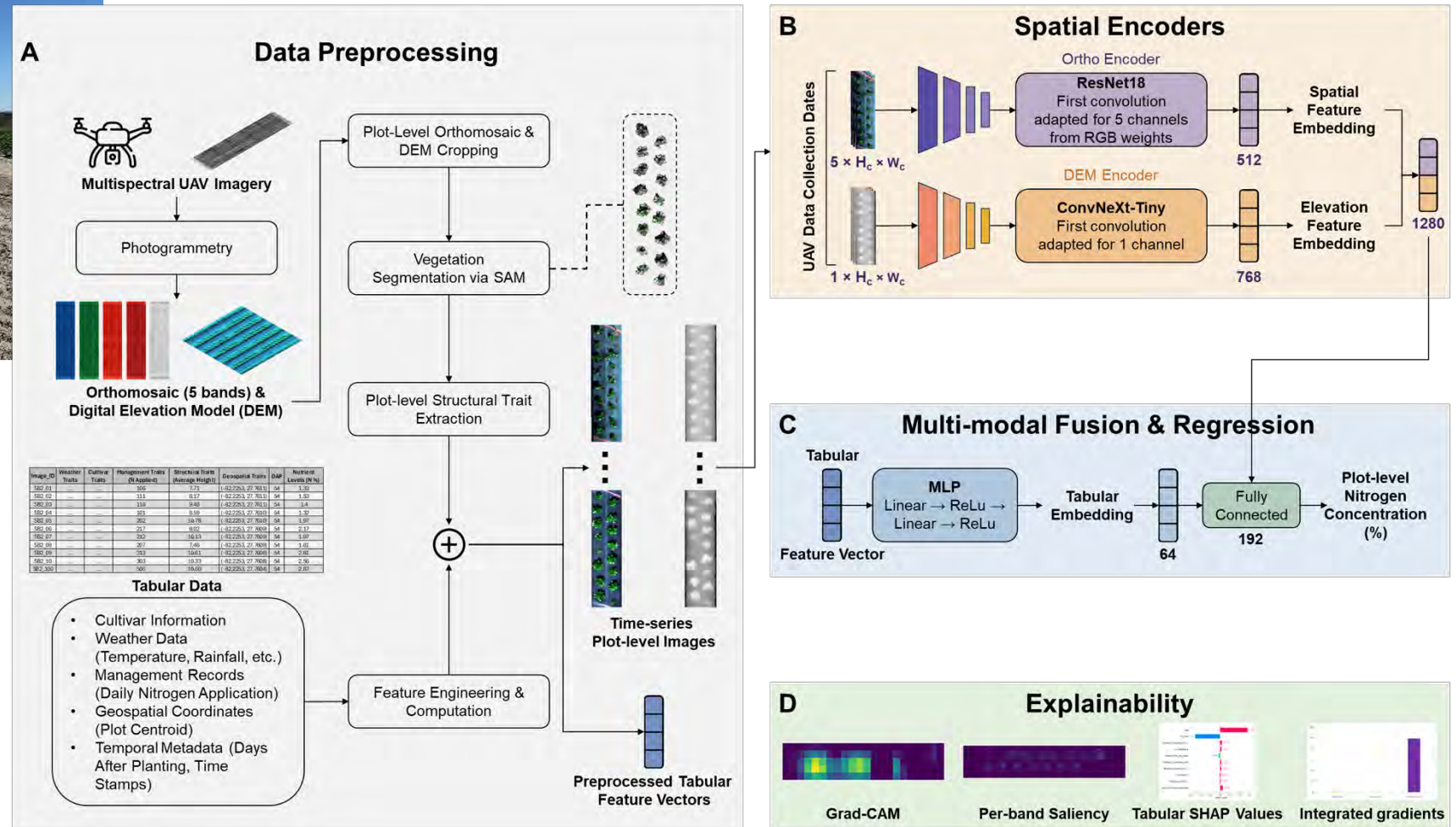
Zinc



Boron

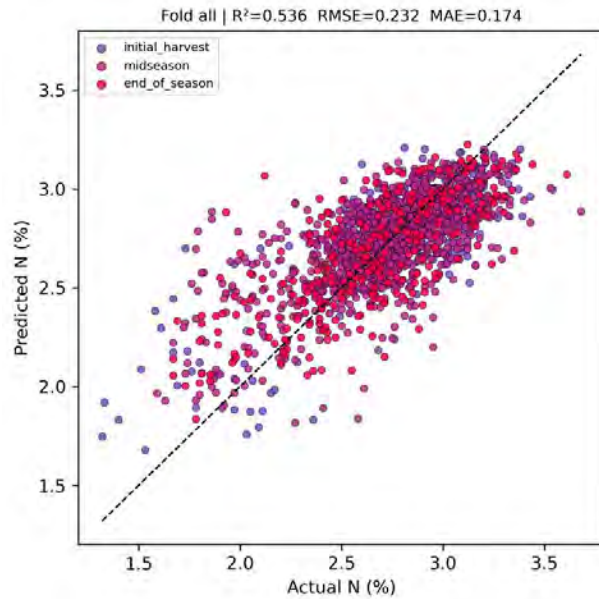
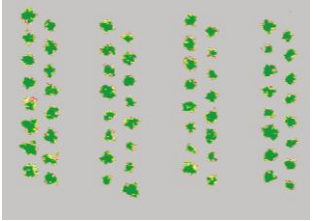


AI-based nutrient diagnosis pipeline



Kalara Dissanayake
PhD student

Leaf N estimation using spectral data



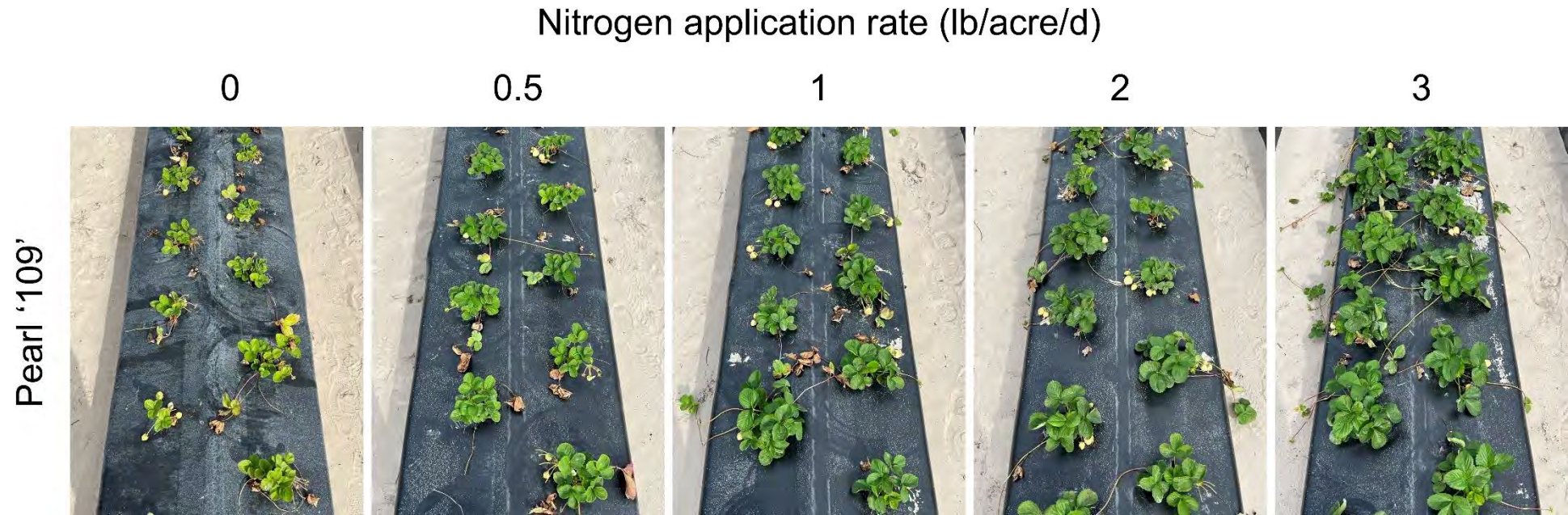
Season	RMSE	R^2
Early	0.24	0.63
Mid	0.23	0.45
Late	0.29	0.70



RMSE → Average prediction error = 0.23–0.29% N

R^2 → Variability in lab values explained by the model = 45–70%

Spectral vs. morphological symptoms



Chlorosis & Size reduction

Size reduction only

Within-canopy spectral patterns

Phosphorus deficiency

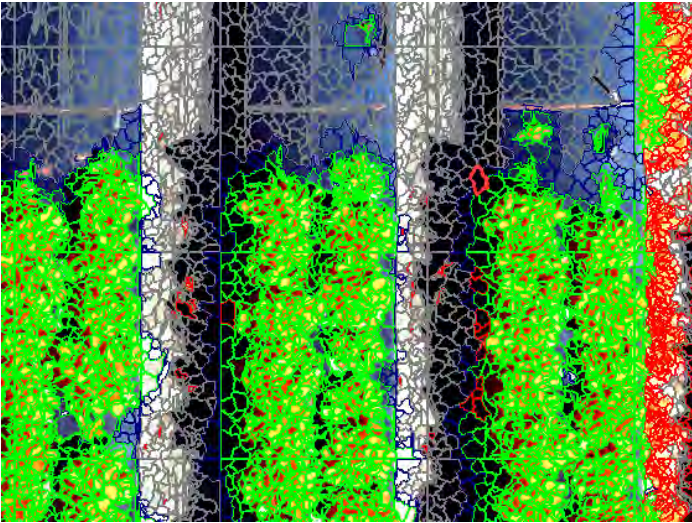


Zinc deficiency



Multi-scale data integration

Field level



Plant level



Metadata

- Weather
- Cultivar
- Days after planting
- Preplant soil test data
- Fertilization records

Multi-scale data integration to maximize the diagnosis accuracy

Acknowledgements



Funding

- Florida Strawberry Research and Education Foundation

Collaborators

- Cecilia Nunes (Associate Professor, University of South Florida)
- Kevin Wang (Assistant Professor, GCREC)
- Dante Pinochet (Professor, Austral University of Chile)

Horticultural Crop Physiology Lab

- Bill Wang (Biological Scientist II)
- Chris DelCastillo (Ag Assistant III)
- Adam Watson (Lab/Field Assistant)
- Lillian Pride (PhD Student)
- Junaid Lone (PhD Student)
- Yasmeen Saleem (PhD Student)
- Farhad Ghasemi (PhD Student)
- Kalara Dissanayake (PhD Student)
- Avninder Kaur (PhD Student)
- Álvaro Bautista (MS Student)
- Madelyn Graethouse (MS Student)

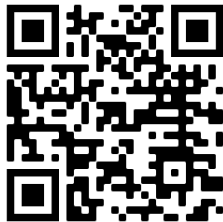
sagehara@ufl.edu
(813) 419-6583



HortLab



Hops



facebook.com/**UFHortLab**
facebook.com/**UFHops**



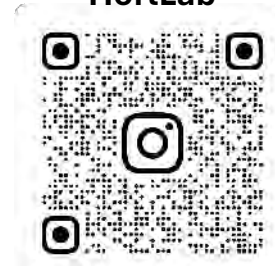
HortLab



youtube.com/**@UFHortLab**



HortLab



instagram.com/**UFHortLab**



HortLab



edis.ifas.ufl.edu/experts/sagehara