

What's New in Corn Silage Harvest Technology

2025 World Dairy Expo

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**AGRI
TECHNICA** ^{DLG}
THE WORLD'S NO. 1

9. - 15.
NOVEMBER
HANNOVER
GERMANY

20
25



CLAAS | CEMOS AUTO CHOPPING



Fendt | Katana 850



NEW HOLLAND | ForageCam™ System

CLAAS

Hall 13 | Stand C18

Subject area: 07.3 Machinery and equipment for chopping, mowing, conditioning and baling of mowed material

CEMOS AUTO CHOPPING - automatically controlled grain processing based on AI-supported CPS real-time analysis

Fendt - AGCO GmbH

Hall 20 | Stand B26

Subject area: 07. Harvesting technology

Fendt - ForageQualityCam

NEW HOLLAND

Hall 3 | Stand B21

Subject area: 07.3 Machinery and equipment for chopping, mowing, conditioning and baling of mowed material

ForageCam™ system



These systems employ **artificial intelligence to distinguish grain from plant material, classify grain particles by size, and calculate the CSPS**, which is then displayed to the operator. With this immediate feedback, operators can continuously adjust cracker settings to match harvesting conditions, ensuring optimal silage quality. Beyond improving feed efficiency, this technology also enhances documentation for future management decisions, while helping **increase throughput, lower wear, and reduce fuel consumption**.





What is CSPS?

J. Dairy Sci. 88:4414–4425

© American Dairy Science Association, 2005.

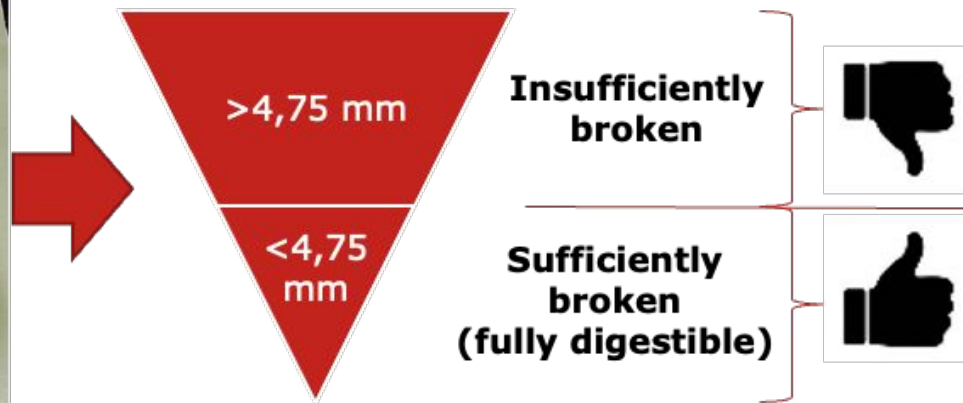
Chemical and Physical Characteristics of Corn Silages and Their Effects on In Vitro Disappearance

G. Ferreira^{1,2} and D. R. Mertens³

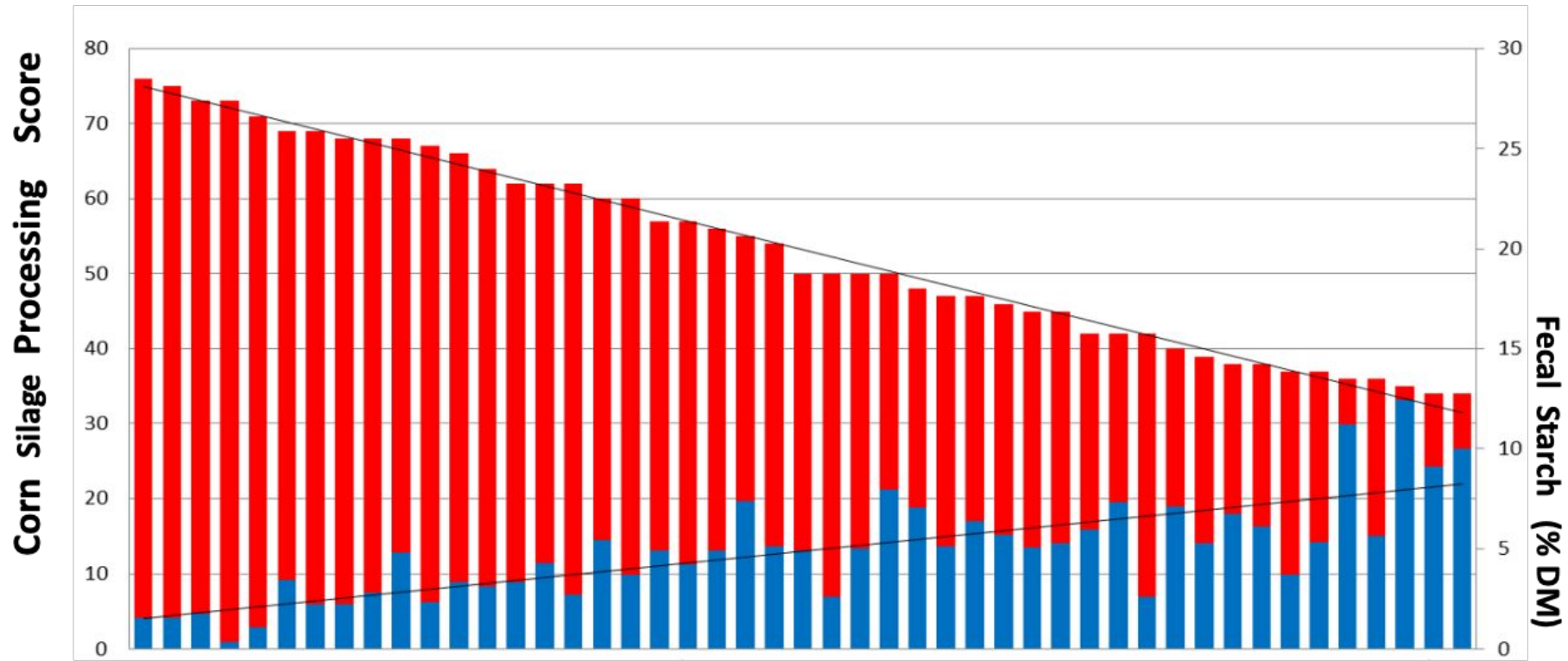
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²Facultad de Ciencias Agrarias, Pontificia Universidad Católica Argentina, Buenos Aires, Argentina

³USDA-ARS, U.S. Dairy Forage Research Center, Madison, WI 53706

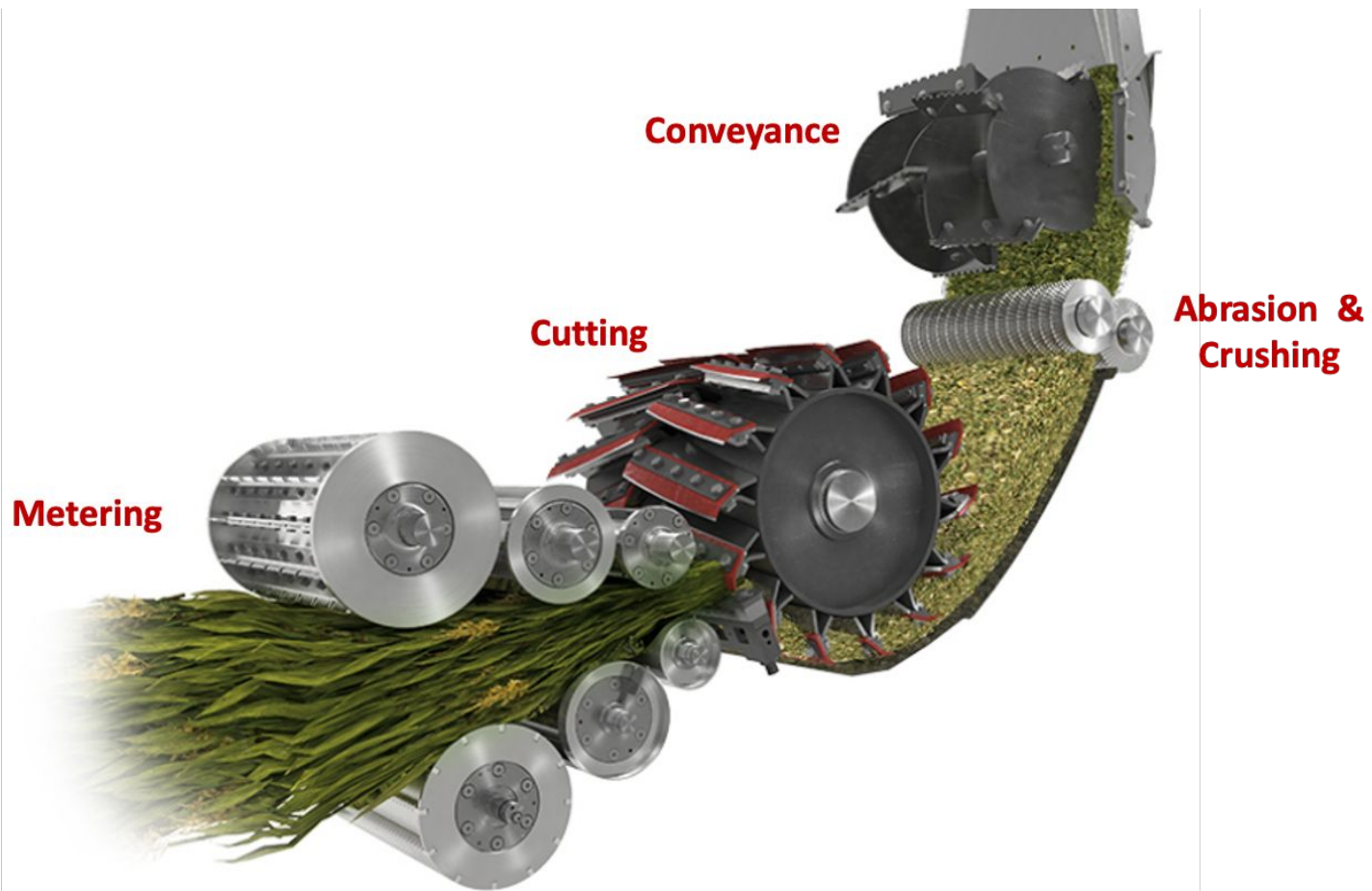


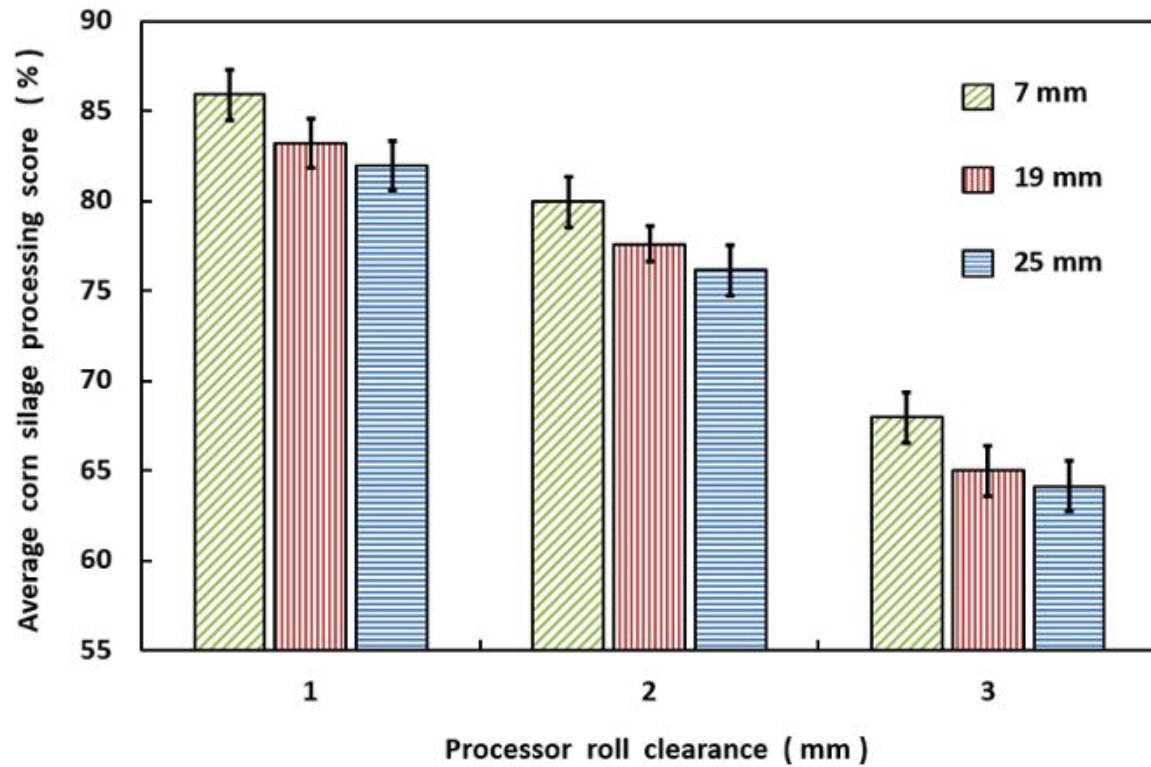
Excellent	$>70\%$
Average	50-70%
Bad	$<50\%$



"One-unit change in fecal starch equals the energy equivalent for 0.72 pounds of milk production."

How do we manipulate CSPS?





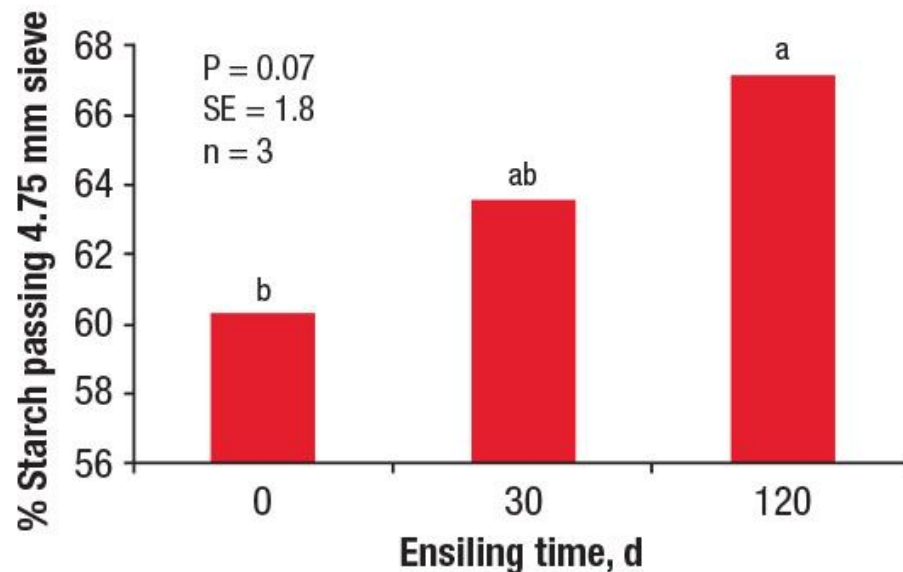


Table 1. Least squares means and SE for the effect of fermentation time on IVSD, NDFd30, and CSPS.

Forage Metric	Fermentation Timepoint, d				SE	P-value
	0	45	90	135		
IVSD, % starch	58.2 ^d	63.2 ^c	67.2 ^b	69.4 ^a	1.7	< 0.001
NDFd30, % NDFom	57.4	57.1	57.0	56.8	0.8	0.23
CSPS, % starch	60.3 ^A	60.0 ^{BA}	59.3 ^{BA}	58.4 ^{BC}	5.9	0.07

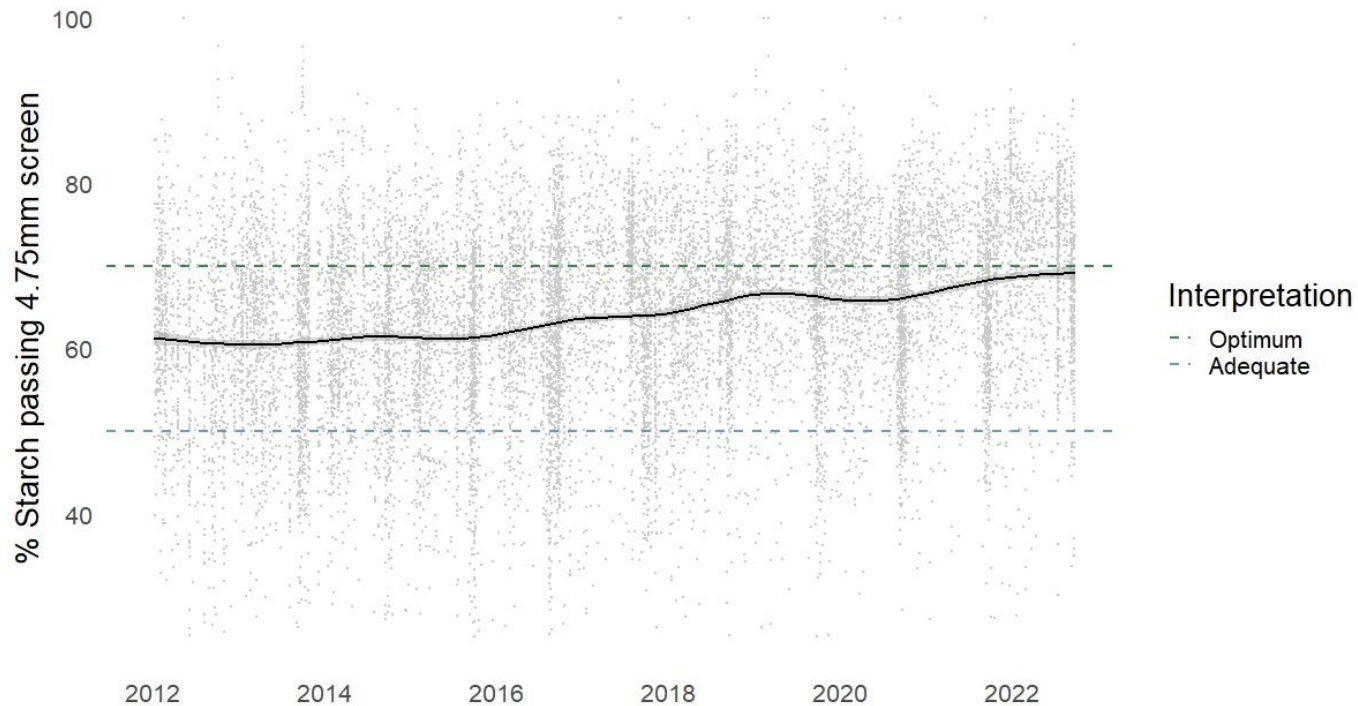
^{a,b,c,d}Means with different superscripts differ ($P < 0.05$).

^{A,B,C}Means with different superscripts differ ($P < 0.10$).

Are we doing a good job?

CSPS has improved over 10 years

(n = 16,879)



Data source: Dairyland Laboratories Inc.

Can we optically measure CSPS?



(a) Sample from the plot with lowest CSPA (59.9%)



(b) Sample from the plot with highest CSPA (89.3%)



(a) Original image.



(b) 2015 model.



(c) 2016 model.

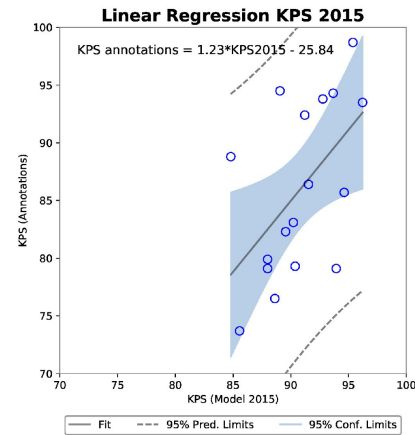


(d) 2017 model.

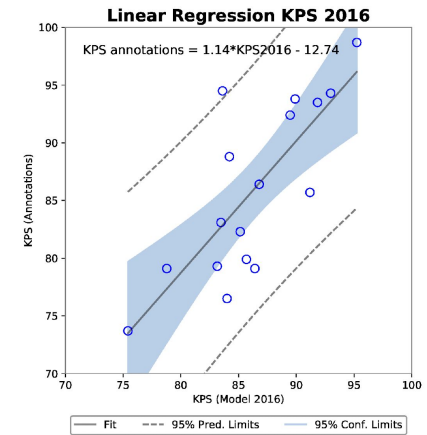


(e) 151617 model.

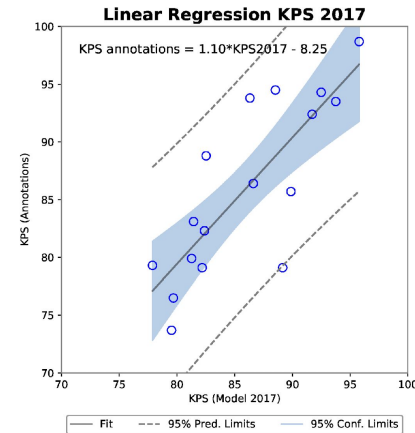
- Unsorted kernels, RGB images
- Strong correlation ($r^2 = 0.78$) between R-FCN predictions and CSPS



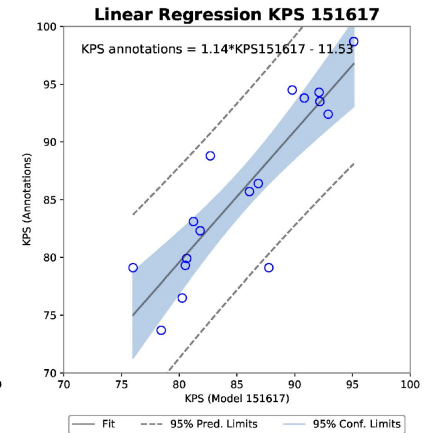
(a) 2015 model.



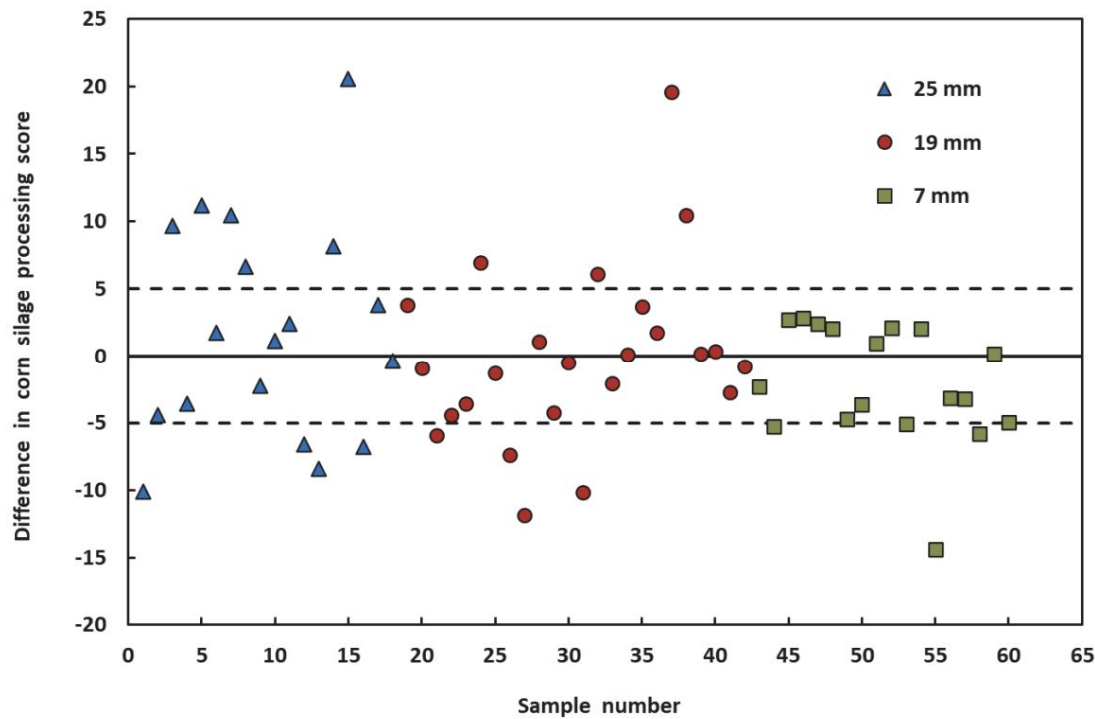
(b) 2016 model.



(c) 2017 model.

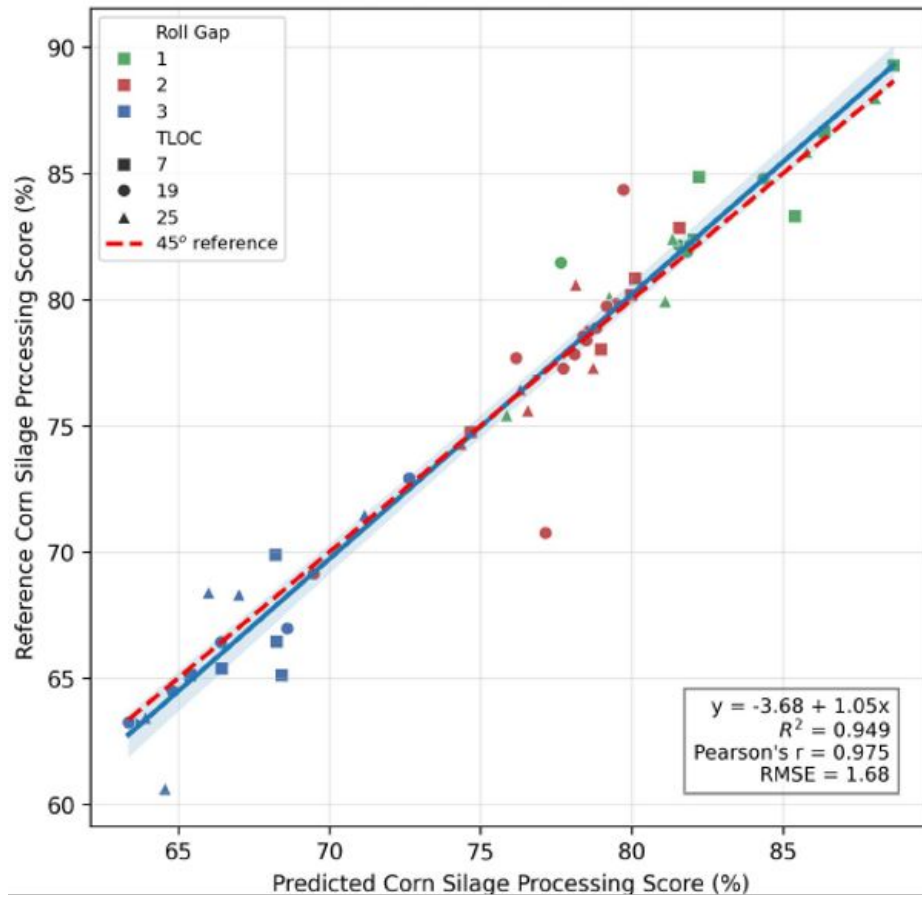


(d) 151617 model.



SEL was 6.7 and 6.9
for Labs A and B.

Random Forest Fusion



This is a big step forward!

So we're done? Not quite:

- **In my opinion** the lab assay is too noisy for precise control of CSPS
- We don't have enough information to understand the cost of attaining a one point change in CSPS
- Many of these systems use **CVML** and can be easily overfit to the data - it will take time to understand the limitations and improve the models

Questions?

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