

A New Era: Technologies for Beef Yield Grade

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YG-1



YG-3



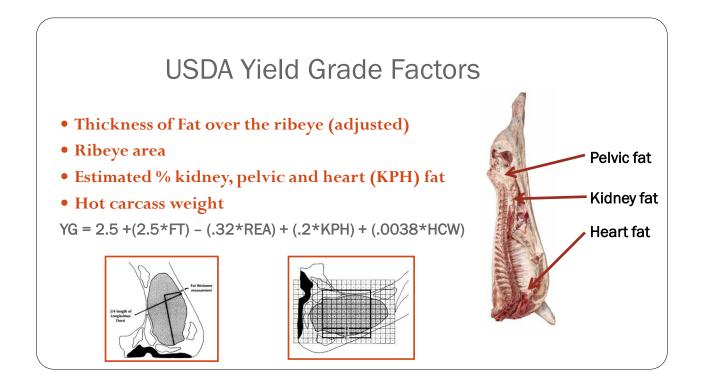
YG-5

USDA Yield Grades

<u>Yield Grades</u>: Reflect differences in yield of closely trimmed, boneless retail cuts from the round, loin, rib, and chuck.

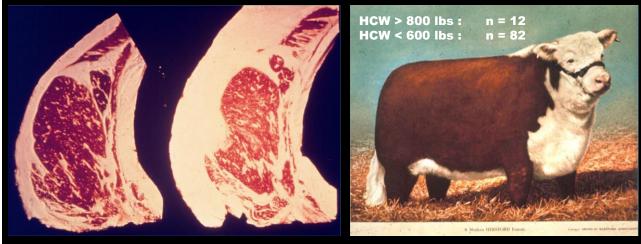
YG-1 more than 52.3%
YG-2 50.1 to 52.3%
YG-3 47.8 to 50.0%
YG-4 45.5 to 47.7%
YG-5 45.4% or less

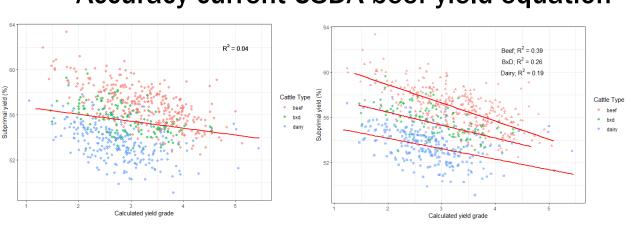




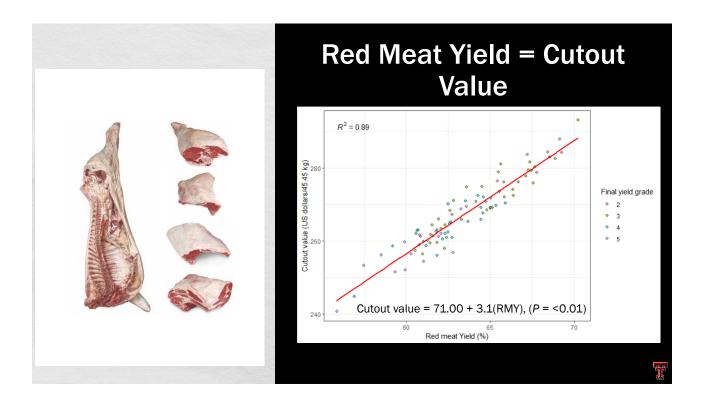
Murphey, **1**960 (N = **1**62)

BCTRC (R-L-R-C) = 51.34 - (5.78*FT) - (.462*%KPH) - (.0093*HCW) + (.74*REA)One unit YG (e.g., 2.0 to 3.0) = 2.3% BCTRC





Accuracy current USDA beef yield equation



UPDATING THE YIELD GRADE EQUATION

Updating Coefficients (Traditional Statistics)

Limited to linear relationships

Current Predictors (HCW, REA, PYG) Accuracy = 17%

Yield Grade = $X_1 + (X_2 \times HCW) + (X_3 \times PYG) + (X_4 \times KPH\%) - (X_5 \times REA)$

Current Predictors + Cattle Type Accuracy = 61%

Yield Grade = $X_1 + (X_2 \times HCW) + (X_3 \times PYG) + (X_4 \times KPH\%) - (X_5 \times REA) + (X_6 \times Cattle Type)$

Artificial Intelligence

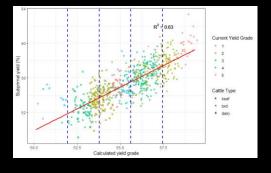
Current Predictors (HCW, REA, PYG) Accuracy = 19%

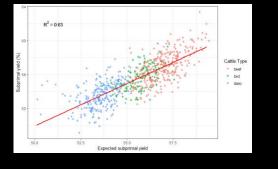
61% improvement *Temporary solution **Current Predictors** + Cattle Type Accuracy = 66%

Accuracy modified subprimal yield equation ~ Adjusted for cattle type

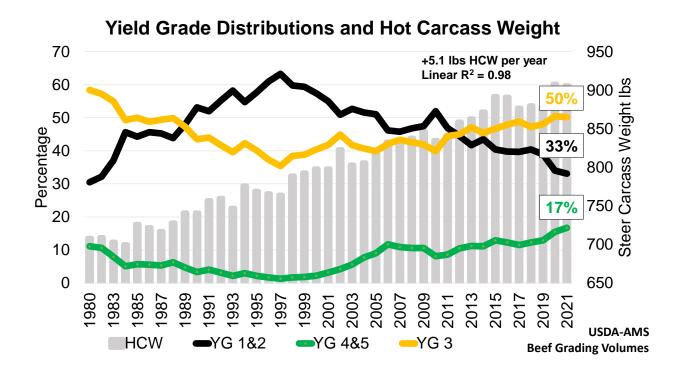
Subprimal yield = 56.94+(0.40*REA)-(0.0042*HCW)-(3.57*FT)

- Beef Adjustment = 0 (baseline) BeefxDairy Adjustment = -1.76 Dairy Adjustment = -4.02



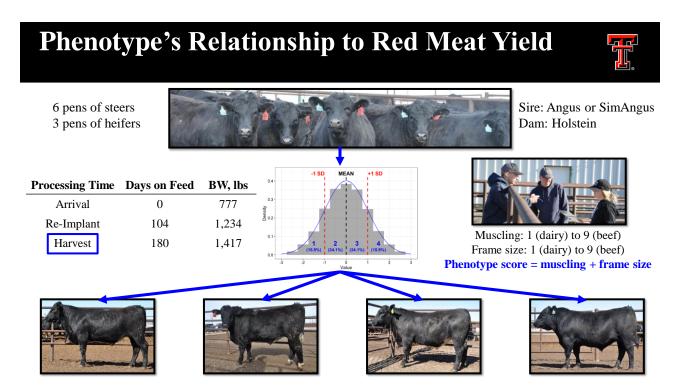


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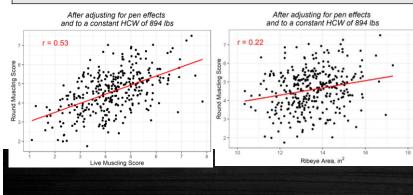
Expression of Phenotype & Red Meat Yield





Muscling Considerations

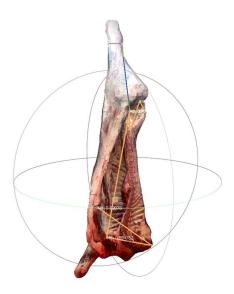
Trait	Fully Dairy-type	Partially Dairy-type	-	Fully Beef-type	P-value
Live muscling score	2.8 ^d	4.0 ^c	4.5 ^b	5.6ª	<0.01
Ribeye area, in ²	13.2	13.5	13.6	13.5	0.30
Round muscling score	3.8°	4.5 ^{bc}	4.8 ^{ab}	5.3ª	<0.01











3-DIMENSIONAL IMAGES TO PREDICT RED MEAT YIELD

• Volume

- Volume
 Linear Measurements
- Cross Sections
- Shape Indicators
- Surface Area
- X, Y, Z Coordinates

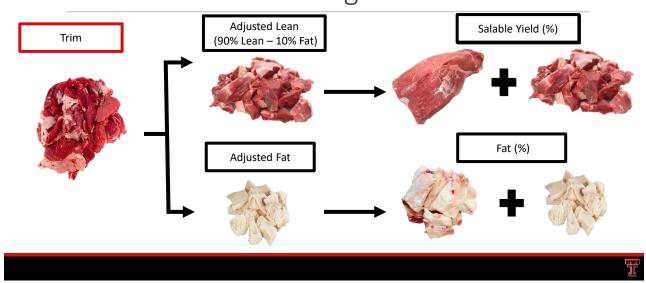


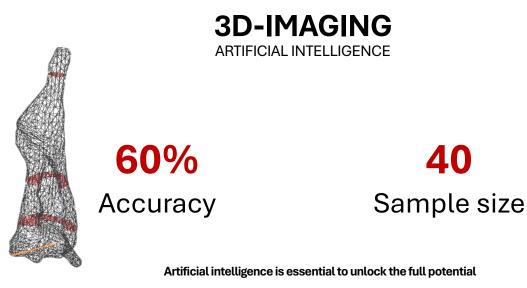
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Materials and Methods Yield Data Collection

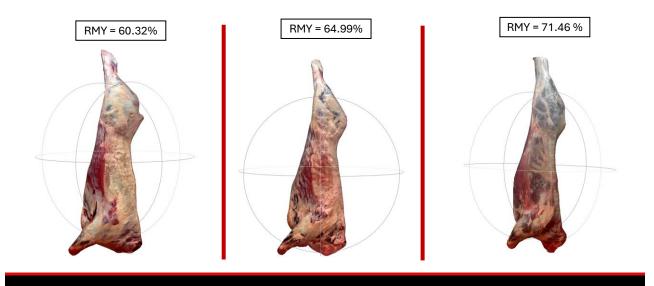


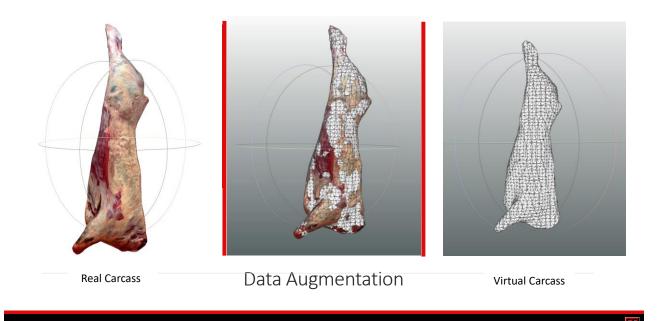
Materials and Methods Data Management





DATA AUGMENTATION

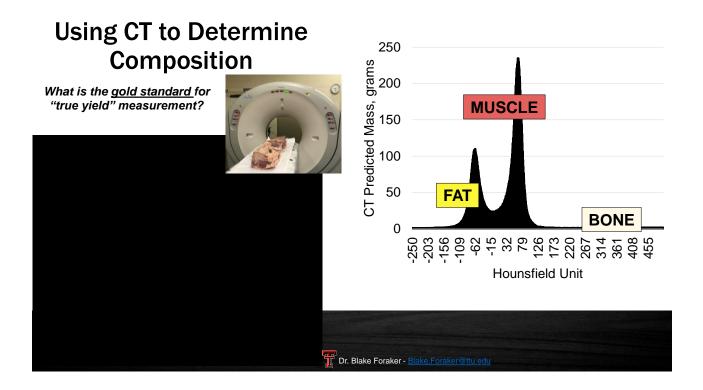




Conclusion

- 1.3D imaging combined with deep learning outperforms current USDA Yield Grade predictors.
- 2. Future research should focus on increasing sample sizes and exploring alternative modeling techniques.







CT data for 3D rendering

Texas Tech and Australian Wagyu Collaboration

Objective I: Development of a Conformation-Based Carcass Scoring System

Objective II: Establishing Relationship Between Ribeye Image Data and Carcass Conformation

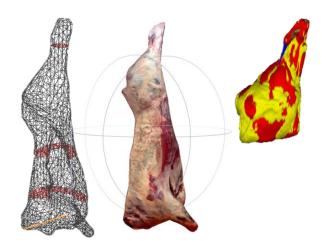
Objective III: Utilizing Advanced 3-Dimensional imaging Technologies for Predicting Red Meat Yield

Your Role...

We are currently looking for **purebred and crossbred Wagyu carcasses in Texas** to support a research project using **3dimensional imaging technologies**.

If you have carcasses available or expect to in the near future, we would be grateful for the opportunity to connect.

Please contact: Dr. Dale Woerner — (970) 980-4386 ; dale.woerner@ttu.edu Cooper Carter — (806) 471-1493 ; coopcart@ttu.edu



Thank You!

Questions ?

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