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Beef Cattle Institute Research Summaries

August 8th, 2025

1

Evaluating the efficacy of Maternal
Bovine Appeasing Substance
(MBAS) (FerAppease®)
administration on pain outcomes
in calves after cautery dehorning
and surgical castration

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Background

- Maternal Bovine Appeasing Substance (MBAS) is a pheromone that is naturally secreted by the skin of the mammary gland and is thought to provide calming effect for nursing calves
- FerAppease® is a synthetic analog of MBAS that is administered topically



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Objective

- To determine if administering MBAS in addition to lidocaine, or in combination with lidocaine and meloxicam, would provide pain relief after surgical castration and cautery disbudding in calves

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Study Design

49 Calves Randomized Across 6 Treatments

| Lidocaine (Lid) | Lidocaine + MBAS (MBAS) | Lidocaine + Meloxicam (Mel) | Lidocaine + Meloxicam + MBAS (Combo) | Sham | No analgesia (Control) |
|-----------------|-------------------------|-----------------------------|--------------------------------------|------|------------------------|
| n=9 | n=10 | n=10 | n=9 | n=6 | n=5 |

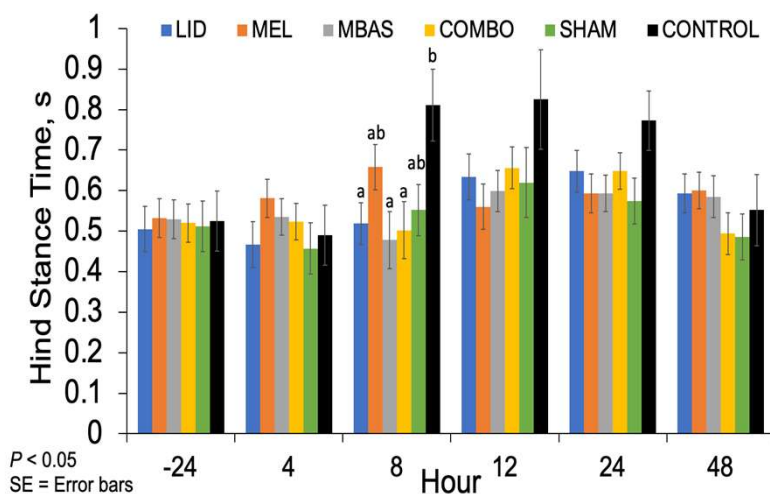
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Gait Analysis

| Hind Stance Time | |
|------------------|----------|
| Treatment | $P=0.99$ |
| Time | $P<0.01$ |
| Treatment x Time | $P=0.02$ |

- Calves that received no analgesia (CONTROL; $n = 5$) had a higher stance time than all other treatments at timepoint 8, 12, 24



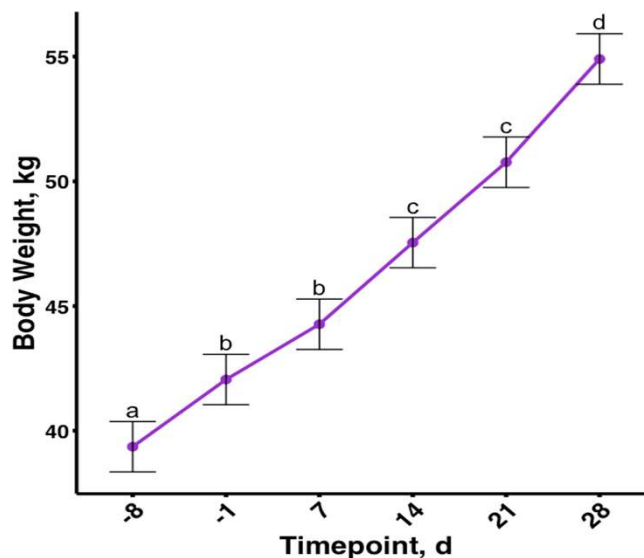
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Performance – Body Weight

| Body weight | |
|------------------|----------|
| Treatment | $P=0.58$ |
| Time | $P<0.01$ |
| Treatment x Time | $P=0.99$ |

- Steadily gained weight as study went on, regardless of treatment group



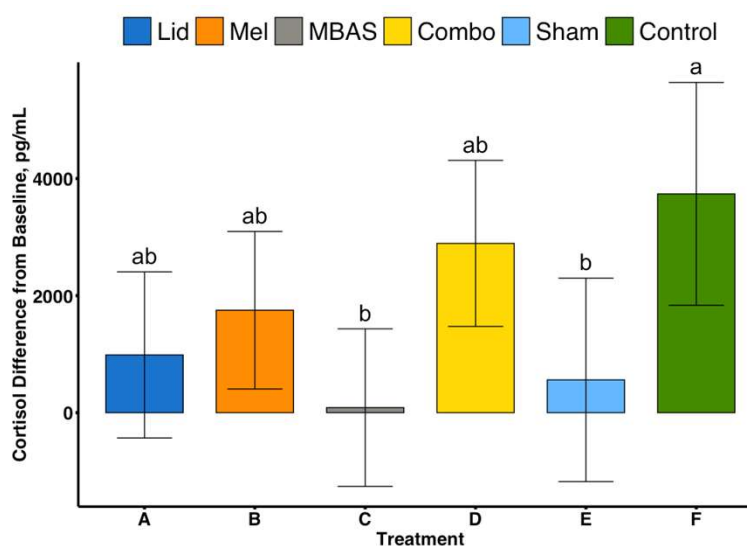
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Plasma Cortisol

| Cortisol | |
|------------------|----------|
| Treatment | $P<0.01$ |
| Time | $P<0.01$ |
| Treatment x Time | $P=0.12$ |

- CONTROL group significantly different than MBAS and SHAM groups



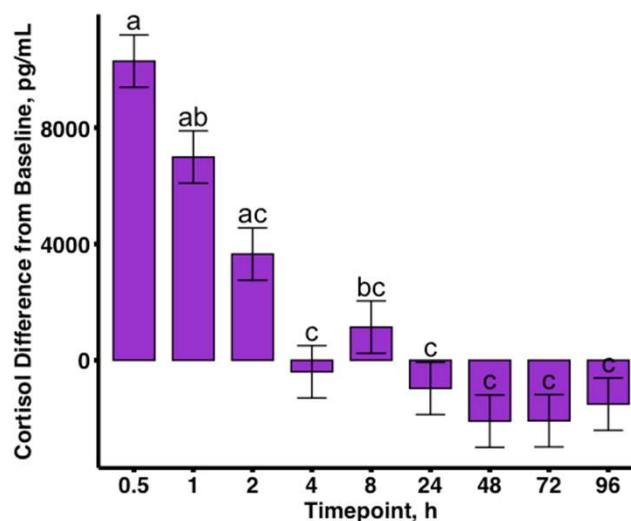
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Plasma Cortisol

| Cortisol | |
|------------------|------------|
| Treatment | $P < 0.01$ |
| Time | $P < 0.01$ |
| Treatment x Time | $P = 0.12$ |

- Cortisol peaked 0.5 h after castration and disbudding, regardless of treatment group
- Continued to decrease over time



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Conclusions

- Significant treatment and treatment x time differences in gait analysis and cortisol were due to CONTROL group, no differences between other treatment groups
- We do not have evidence that MBAS is more effective than only lidocaine when administered to dairy calves undergoing surgical castration and cautery disbudding

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Questions?
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The effect of shade on
steer performance after
terminal sort



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Study Objectives:

- Compare performance outcomes of steers allowed shade to steers allowed no shade
 - Panting behavior
 - Water consumption
 - Feed delivery
 - Health
 - Carcass traits (HCW, quality grade, yield grade)



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Study Design:

- Randomized controlled trial
- Pen: experimental and observational unit
- Study groups
 - Shade (S; n=12, ~350 hd/pen)
 - 30 ft²/head
 - 100% solar block
 - No shade (NS; n=12 , ~350 hd/pen)
- Blocked by week of allocation (n=4)
 - 6 pens enrolled each week
- T-sort pens (approx. 60 days from projected ship date; n=3)
 - T-1: small
 - T-2: medium
 - T-3: large
- Randomized individual animals to shade or no shade in respective T-sort group (n=6)

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The effect of shade on steer performance after terminal sort

MATERIALS AND METHODS



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| Beef Cattle Temperature Humidity Chart | | | | | | | | | | | | | |
|---|-----|-----------------------|----|----|----|----|----|----|----|----|----|----|----|
| | | Relative Humidity (%) | | | | | | | | | | | |
| | | 30 | 35 | 40 | 45 | 50 | 55 | 60 | 65 | 70 | 75 | 80 | 85 |
| Temperature (°F) | 100 | 84 | 85 | 86 | 87 | 88 | 90 | 91 | 92 | 93 | 94 | 95 | 97 |
| | 98 | 83 | 84 | 85 | 86 | 87 | 88 | 89 | 90 | 91 | 93 | 94 | 95 |
| | 96 | 81 | 82 | 83 | 85 | 86 | 87 | 88 | 89 | 90 | 91 | 92 | 93 |
| | 94 | 80 | 81 | 82 | 83 | 84 | 85 | 86 | 87 | 88 | 89 | 90 | 91 |
| | 92 | 79 | 80 | 81 | 82 | 83 | 84 | 85 | 86 | 87 | 88 | 89 | 90 |
| | 90 | 78 | 79 | 79 | 80 | 81 | 82 | 83 | 84 | 85 | 86 | 87 | 88 |
| | 88 | 76 | 77 | 78 | 79 | 80 | 81 | 81 | 82 | 83 | 84 | 85 | 86 |
| | 86 | 75 | 76 | 77 | 78 | 78 | 79 | 80 | 81 | 81 | 82 | 83 | 84 |
| | 84 | 74 | 75 | 75 | 76 | 77 | 78 | 78 | 79 | 80 | 80 | 81 | 82 |
| | 82 | 73 | 73 | 74 | 75 | 75 | 76 | 77 | 77 | 78 | 79 | 79 | 80 |
| | 80 | 72 | 72 | 73 | 73 | 74 | 75 | 75 | 76 | 76 | 77 | 78 | 78 |
| | 78 | 70 | 71 | 71 | 72 | 73 | 73 | 74 | 74 | 75 | 76 | 76 | 76 |
| | 76 | 69 | 70 | 70 | 71 | 71 | 72 | 72 | 73 | 73 | 74 | 74 | 75 |
| Temperature Humidity Index (THI) | | | | | | | | | | | | | |
| <div>Normal <75</div> <div>Alert 75-78</div> <div>Danger 79-83</div> <div>Emergency >84</div> | | | | | | | | | | | | | |

Figure 1: Cattle Temperature Humidity Index Chart

<https://beef.unl.edu/handling-cattle-through-high-heat-humidity-indexes/>

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Materials and Methods:

- Panting behavior: shaded pen



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Materials and Methods:

- Panting behavior: shaded pen



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Materials and Methods:

- Panting behavior: unshaded pen



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Materials and Methods:

- Panting behavior: unshaded pen

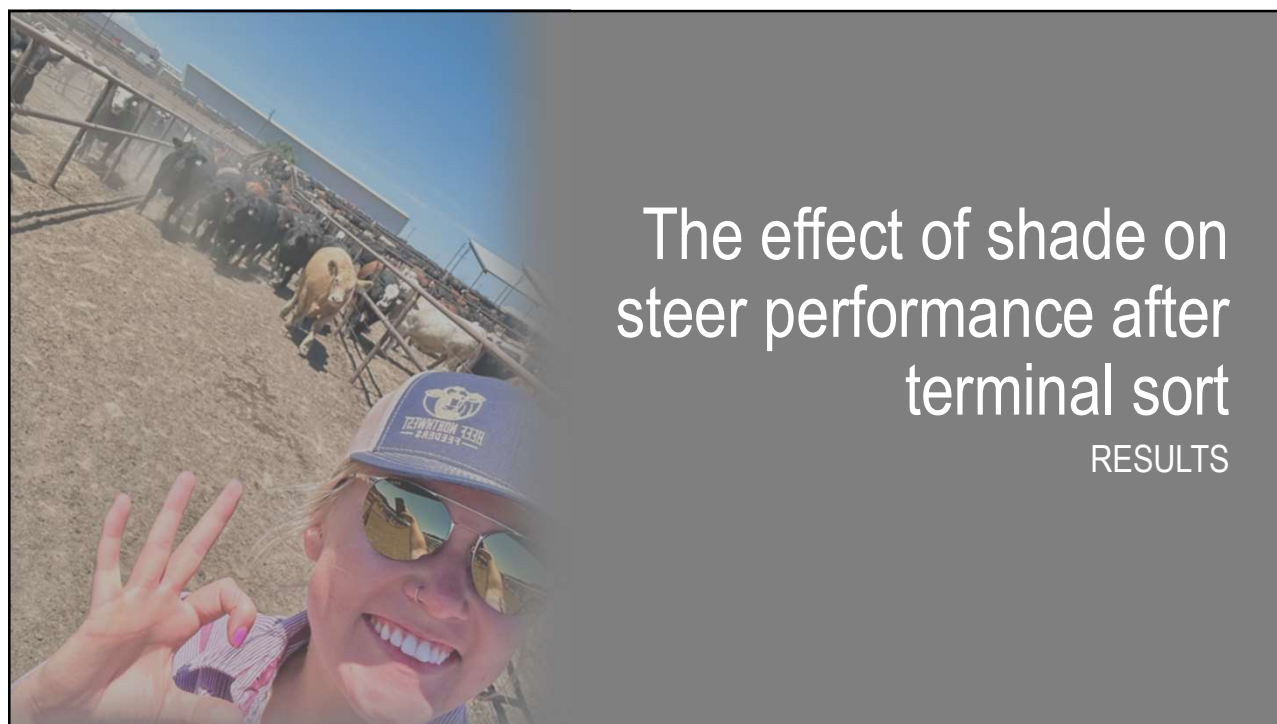


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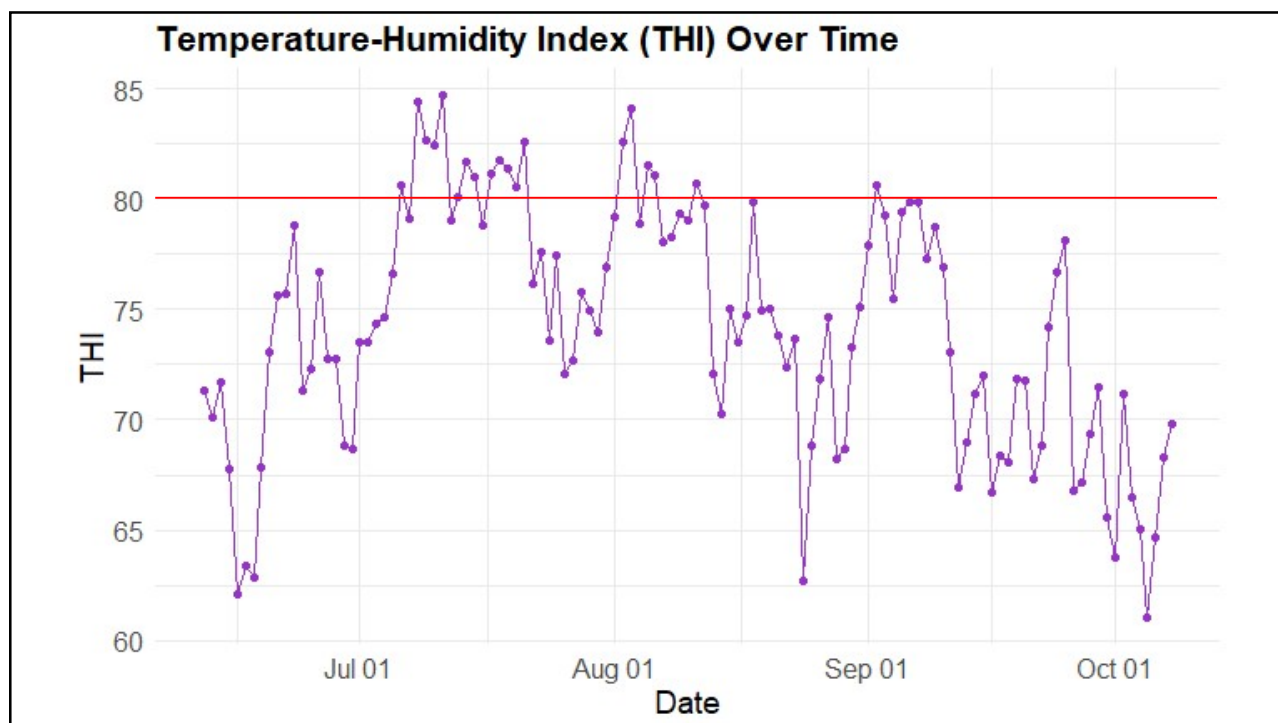
Materials and Methods:

- Pens shipped: August 4 – October 9, 2024
- Carcass collections:
 - Lot level data from commercial packing plant
 - Combined to pen level data
 - HCW/head
 - Quality grade
 - Yield grade
- Generalized linear mixed effects models
 - Outcome: count of quality grade/yield grade/dark cutters
 - Fixed effects: treatment
 - Random effects: sort group within block

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Results:

- Water data (gallons consumed per 1000 lbs BW)*
- treatment $p < 0.05$

| Treatment | Model estimated mean (back transformed) | SEM (transformed) |
|-----------|---|-------------------|
| No shade | 9.4 gal/1000 lbs | 0.008 |
| Shade | 8.0 gal/1000 lbs | 0.008 |

*Box-Cox transformation performed to obtain normality, SEM and P are reported as transformed, and means are reported as back-transformed

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Results:

- Water data (gallons consumed per 1000 lbs BW)*
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| No shade | 9.4 gal/1000 lbs | 0.008 |
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*Box-Cox transformation performed to obtain normality, SEM and P are reported as transformed, and means are reported as back-transformed

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Results:

- Water data (gallons consumed per 1000 lbs BW)*
- THI $p < 0.05$

| THI | Model estimated mean (back transformed) | SEM (transformed) |
|--------------------------|---|-------------------|
| Less than or equal to 80 | 7.8 gal/1000 lbs | 0.01 |
| Greater than 80 | 9.69 gal/1000 lbs | 0.01 |

*Box-Cox transformation performed to obtain normality, SEM and P are reported as transformed, and means are reported as back-transformed


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Results:

- Water data (gallons consumed per 1000 lbs BW)*
- THI $p < 0.05$

| THI | Model estimated mean (back transformed) | SEM (transformed) |
|--------------------------|---|-------------------|
| Less than or equal to 80 | 7.8 gal/1000 lbs | 0.01 |
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37% increase



*Box-Cox transformation performed to obtain normality, SEM and P are reported as transformed, and means are reported as back-transformed

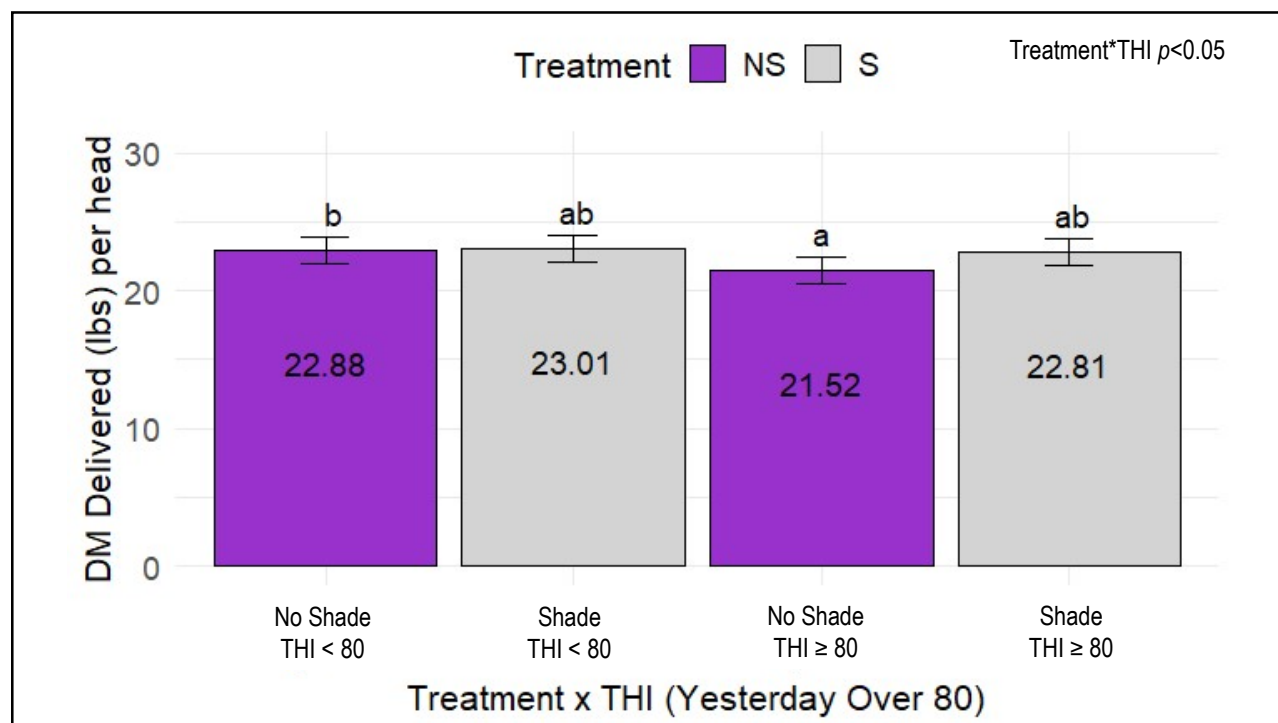
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Results:

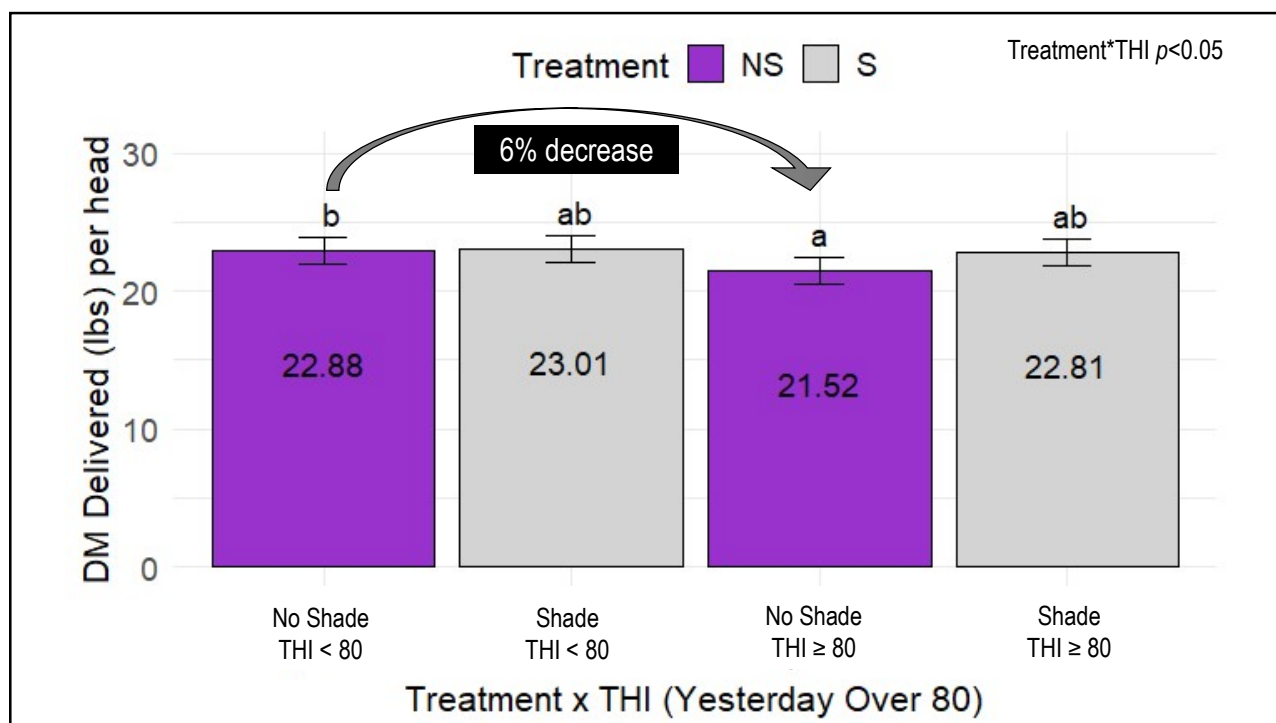
- Daily feed delivery data (DM delivered/head)
 - Previous days THI $p < 0.05$
 - Treatment * Yesterdays THI $p < 0.05$



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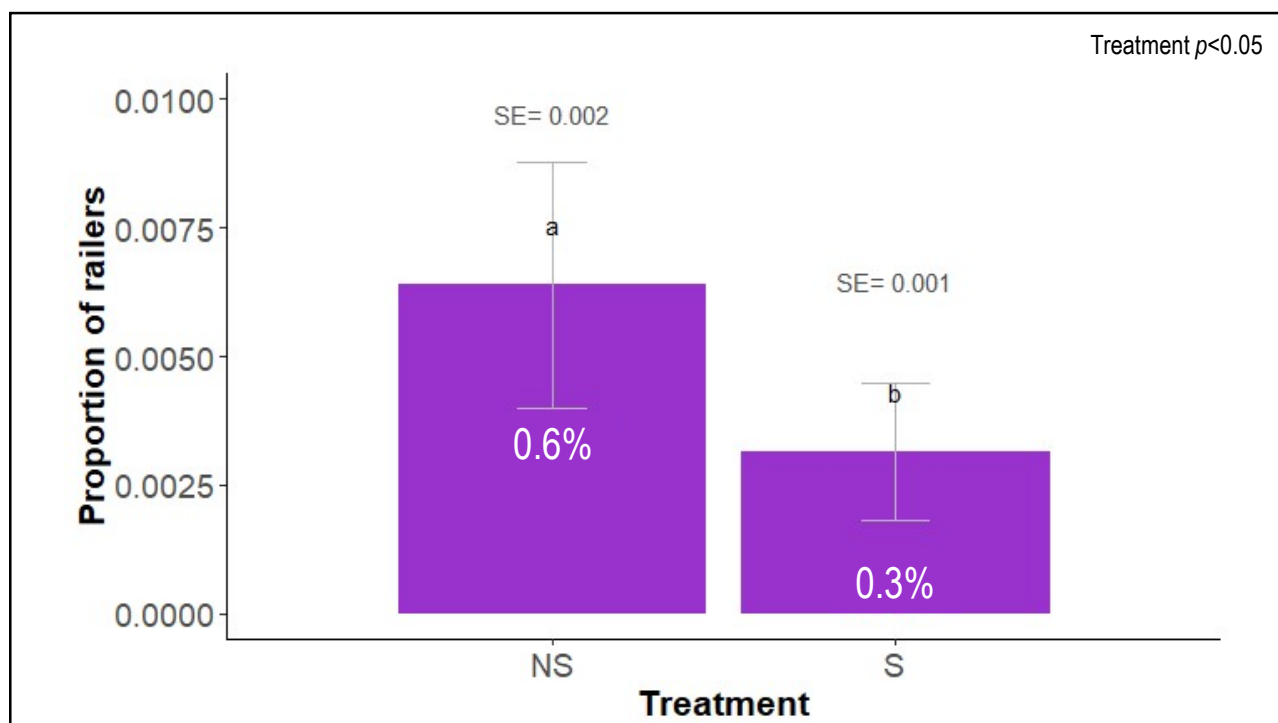
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Results:

- Health outcomes:
 - Morbidity $p > 0.05$
 - Mortality $p > 0.05$
 - Did not finish $p > 0.05$
 - Railers $p < 0.05$

| Treatment | morbidity (%) | mortality (%) | railer (%) |
|-----------|---------------|---------------|------------|
| shade | 9.2% | 0.9% | 0.4% |
| no shade | 9.8% | 0.9% | 0.7% |

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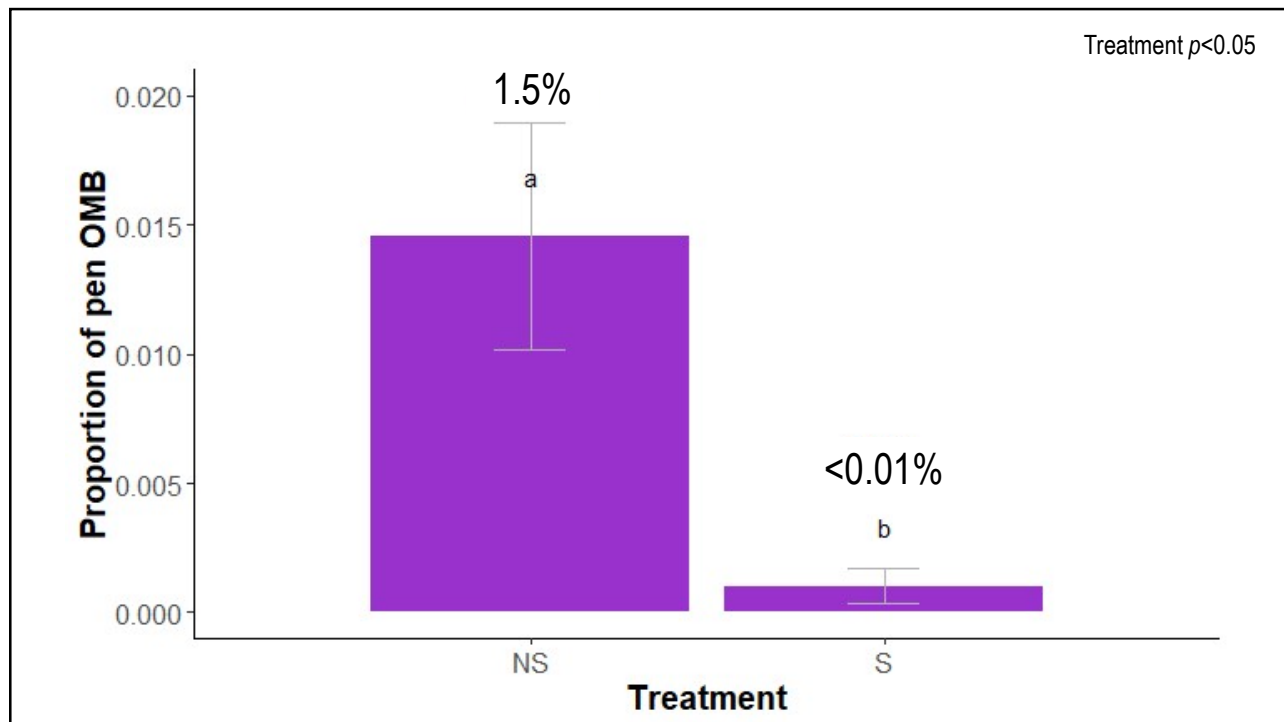


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Results:

- Panting behavior outcomes
 - Treatment $p < 0.05$
 - THI over 80 $p < 0.05$
 - Location $p < 0.05$
- Interactions tested ($p > 0.05$)

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Results:

- Pen level at enrollment

| Outcome | Model estimated means (SEM) | | <i>P-value</i> |
|---------------------------|-----------------------------|-----------------|----------------|
| | Shade | No shade | |
| Enrollment: | | | |
| Head count | 325 (7.1) | 324 (6.9) | 0.92 |
| Total pen weight (lbs) | 431,462 (21897) | 432,633 (20537) | 0.94 |
| Average weight/head (lbs) | 1327 (50) | 1336 (42) | 0.64 |

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Results:

- Pen level at enrollment

| Outcome | Model estimated means (SEM) | | <i>P-value</i> |
|---------------------------|-----------------------------|-----------------|----------------|
| | Shade | No shade | |
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| Total pen weight (lbs) | 431,462 (21897) | 432,633 (20537) | 0.94 |
| Average weight/head (lbs) | 1327 (50) | 1336 (42) | 0.64 |

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Results:

- Pen level at finish

*Box-Cox transformation performed to obtain normality, SEM and P are reported as transformed, and means are reported as back-transformed

| Outcome | Model estimated means (SEM) | | <i>P-value</i> |
|--------------------------------|-----------------------------|--------------|----------------|
| | Shade | No shade | |
| Finish: | | | |
| Average live weight/head (lbs) | 1593 (25.5) | 1589 (20.3) | 0.88 |
| Average HCW/head (lbs) | 1013 (15.8) | 1001 (12.6) | 0.53 |
| ADG/head (lbs)* | 3.5 (1.9) | 3.4 (1.8) | 0.75 |
| F:G/head (lbs)* | 6.12 (0.003) | 5.98 (0.003) | 0.54 |

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Results:

- Yield grade
 - Pen level
 - Categorized as yield grade 1, 2, and 3, versus 4, and 5
 - Treatment $P > 0.05$, no detectable difference

| | Average percent Yield Grade | | | | |
|-----------|-----------------------------|-----|-----|-----|----|
| Treatment | 1 | 2 | 3 | 4 | 5 |
| shade | 5% | 28% | 46% | 19% | 2% |
| no shade | 5% | 28% | 47% | 19% | 2% |

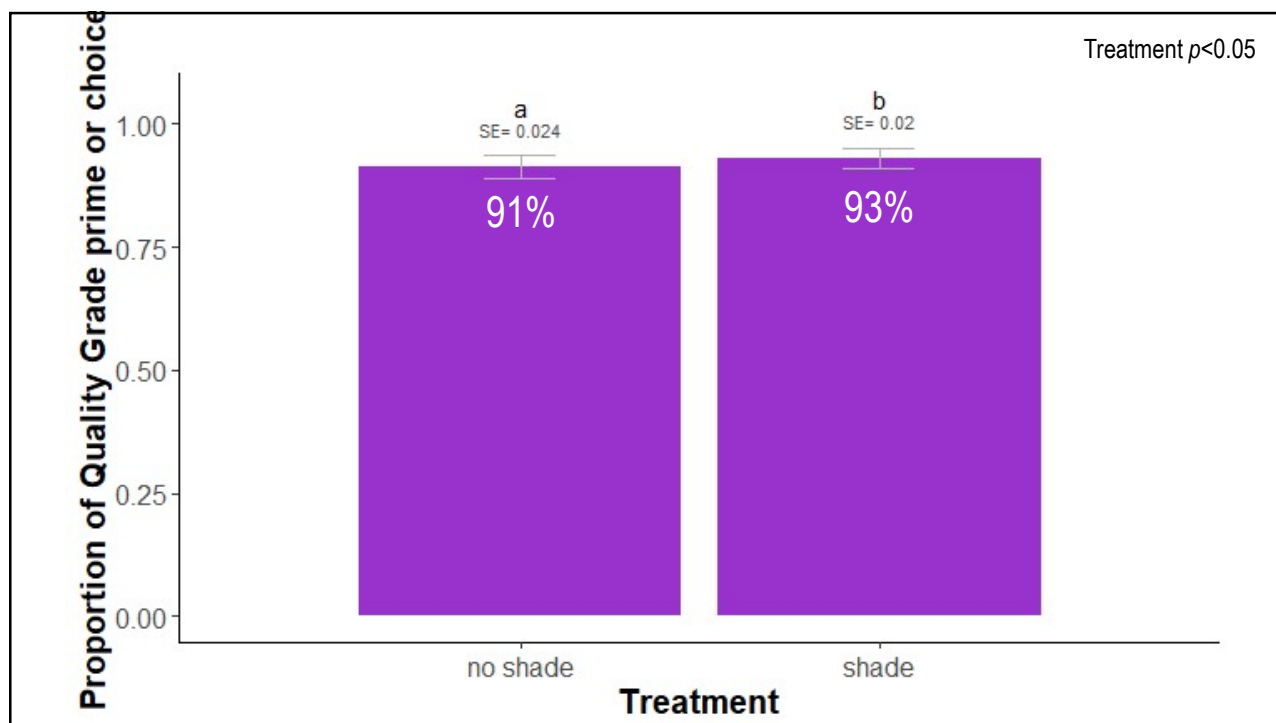
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Results:

- Quality grade
 - Pen level
 - Categorized as prime and choice versus select
 - Treatment $P < 0.01$, significantly different

| | Average percent Quality Grade | | | |
|-----------|-------------------------------|--------|--------|-------|
| Treatment | prime | choice | select | other |
| shade | 3% | 88% | 9% | 1% |
| no shade | 3% | 85% | 10% | 2% |

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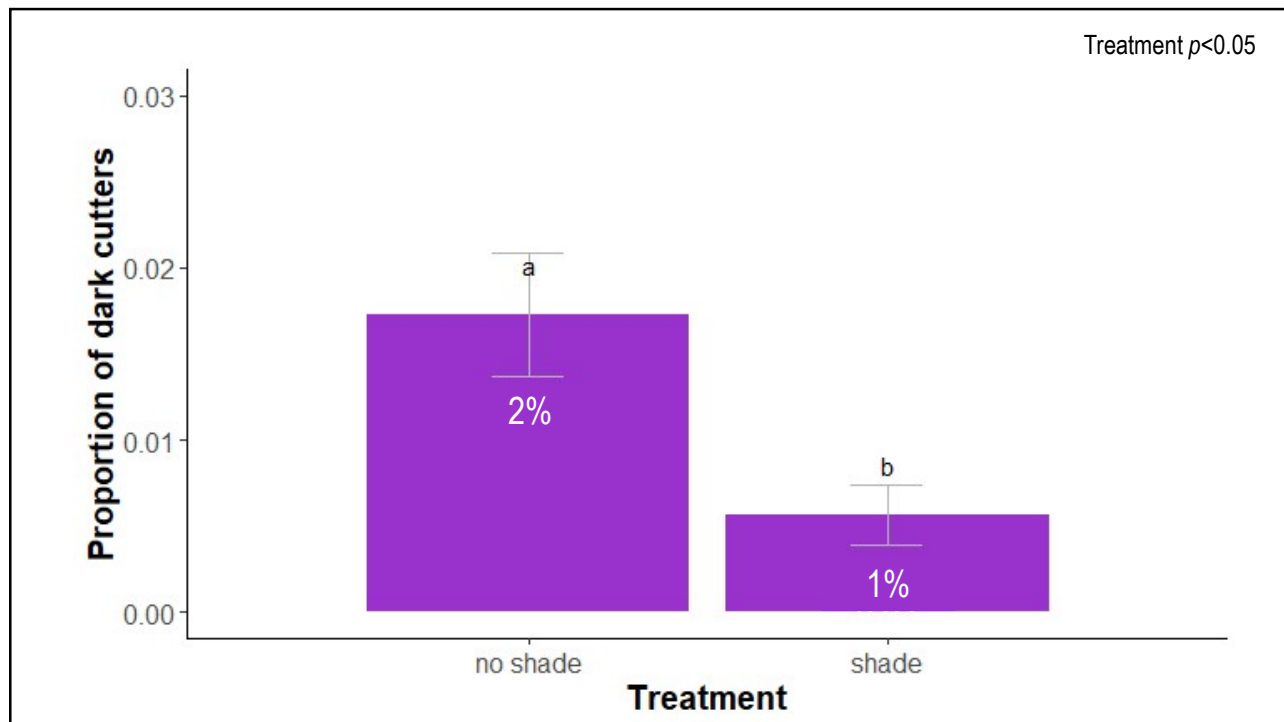
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Results:

- Dark cutters
 - Pen level
 - Treatment $P < 0.001$, significantly different

| Treatment | Avg percent dark cutter |
|-----------|-------------------------|
| shade | 1% |
| no shade | 2% |

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Conclusions:



- **Shade** was shown to:

- 
 - Increase
 - Quality grade prime and choice vs. select

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Conclusions:

- **Shade** was shown to:

- 
 - Increase
 - Quality grade prime and choice vs. select
- 
 - Decrease
 - Water consumption
 - Panting behavior
 - Railer count
 - Dark cutter count

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Conclusions:

- **Shade** was shown to:

- Increase
 - Quality grade prime and choice vs. select
- Decrease
 - Water consumption
 - Panting behavior
 - Railer count
 - Dark cutter count
- No difference in feed delivery between THI categories

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Thank you!

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Management practices of pre-weaned beef-on-dairy calves on commercial calf ranches

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Background

- Increasing number of beef-on-dairy calves
- A calf ranch raises calves from a young age to a targeted weight or age



<https://www.dairyforward.com/news/424224/Halls-Calf-Ranch-Innovative-caring.htm>



<https://www.northernag.net/wp-content/uploads/2024/09/Beef-on-Dairy.jpg>

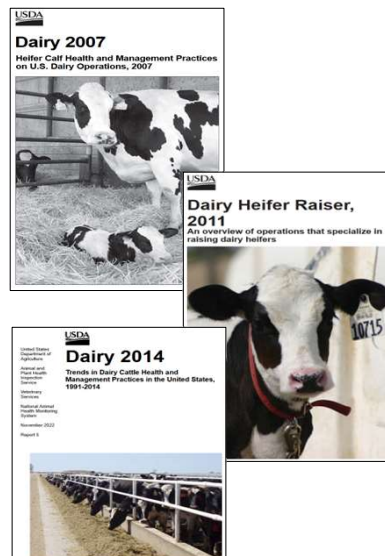
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Background

- Multiple national surveys have been conducted on **management of replacement dairy heifers**

- Dairy 2007
- Dairy Heifer Raiser 2011
- Dairy 2014



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Objective

- To describe management practices of beef-on-dairy calves implemented on commercial calf ranches

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Materials and Methods

- A comprehensive survey was developed
 - Designed in discussion with commercial calf ranch owners, managers and consulting veterinarians
- A total of 15 calf ranches were surveyed
- Surveys conducted in-person or via video call between November 2023 and June 2024

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Materials and Methods

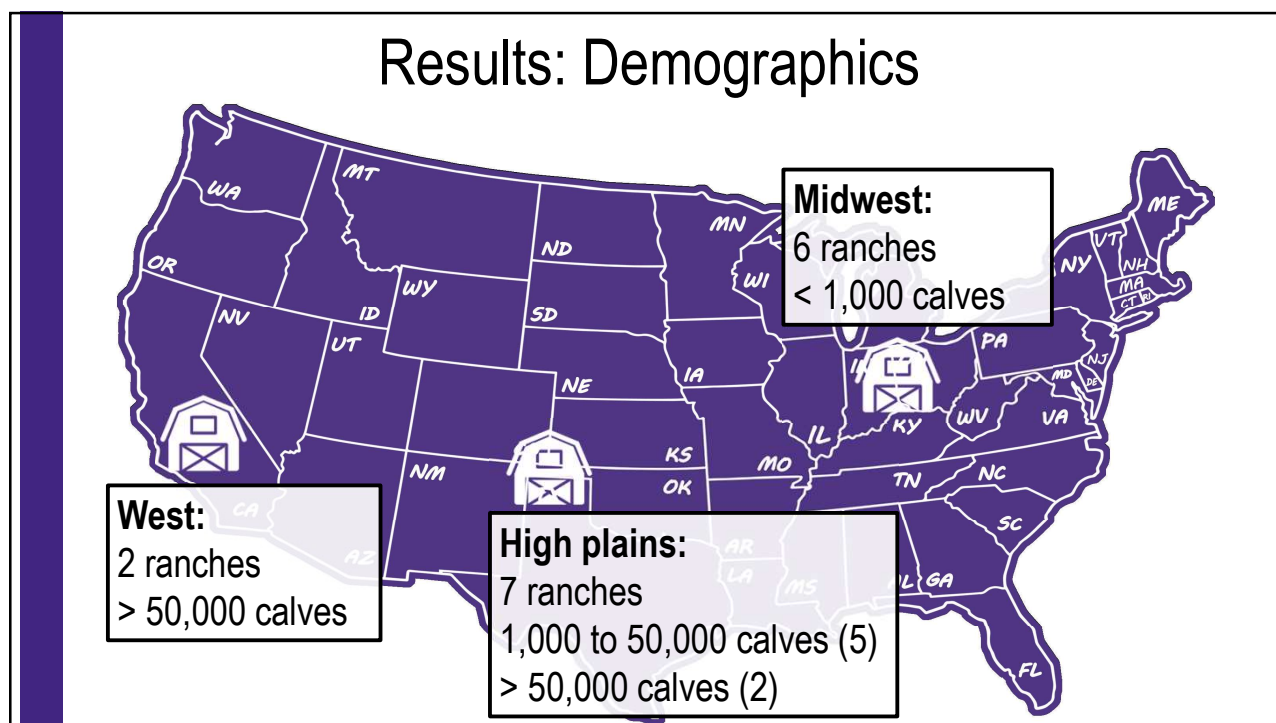
- Survey consisted of 10 sections:
 - General information/ranch demographics
 - Calf arrival processing procedures
 - Pre-weaning housing
 - Milk feeding protocol
 - Starter feed formulation and feeding protocol
 - Weaning protocol
 - Health challenges and vaccine/treatment protocols
 - Water offerings
 - Movement/management of group pens
 - Transition/grower diet formulation and feeding protocol

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Results: Demographics



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Results: Calf Arrival & Pre-weaning Housing

- All ranches received calves that were 4 days old or younger
- 4 types of pre-weaning housing:
 - Group housing (7%; 1 ranch)
 - Plastic hutches with runs (27%, 4 ranches)
 - Wooden hutches (33%, 5 ranches)
 - Individual pens (33%, 5 ranches)



Washington State University Extension, EM045E



<https://calfhutch.com/products/rancher-calf-hutches/>

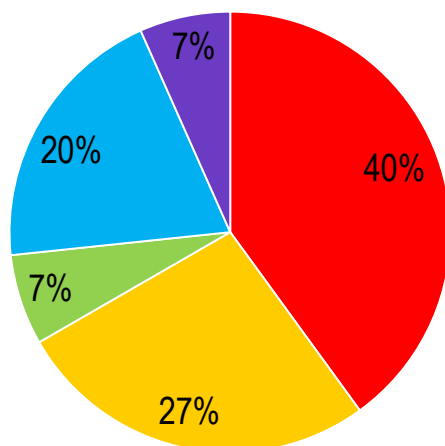


<https://www.agriland.ie/farming-news>

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Results: Milk Feeding Protocol

Types of milk fed:



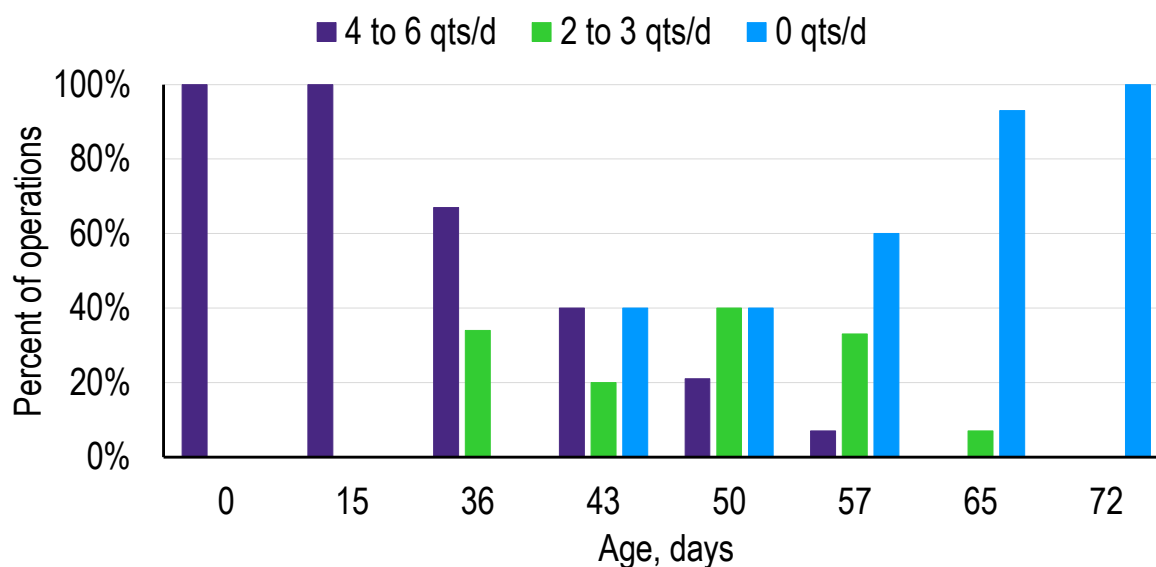
- Milk replacer
- Milk replacer; Nonsaleable milk
- Milk replacer; Saleable milk
- Milk replacer; Nonsaleable milk; Saleable milk
- Saleable milk

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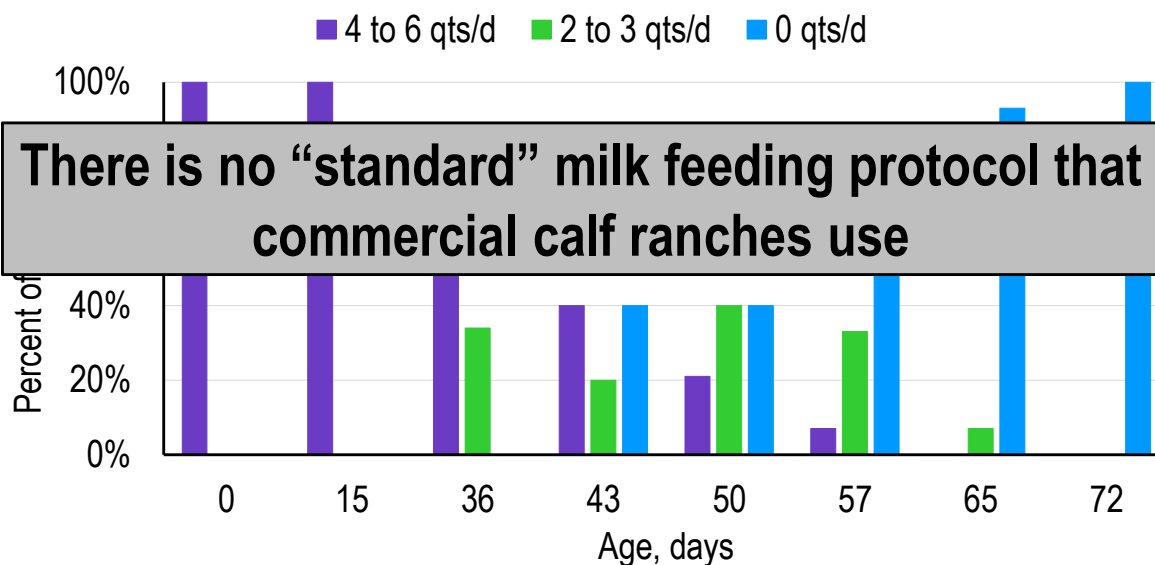
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Results: Milk Feeding Protocol



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Results: Milk Feeding Protocol



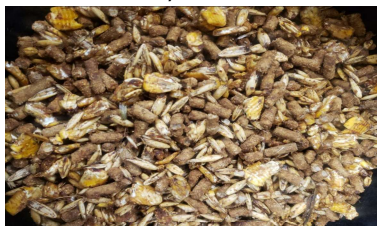
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Results: Calf Starter Diets

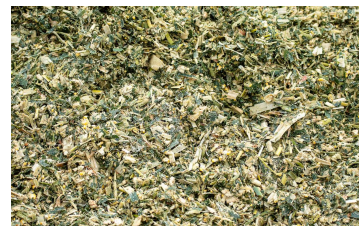
- Calf starter was offered upon arrival at all ranches
- Three types of feed:
 - Pelleted feed (40%; 6 ranches)
 - Texturized feed (47%; 7 ranches)
 - Total mixed ration (13%; 2 ranches)



https://cms-static.wehaacdn.com/hoards-com/images/201025_648-Calf-Starter.19570.jpg



<https://www.farmerscoop.com/wp-content/uploads/2022/05/18-Calf-Grower-2021-768x1024.jpg>

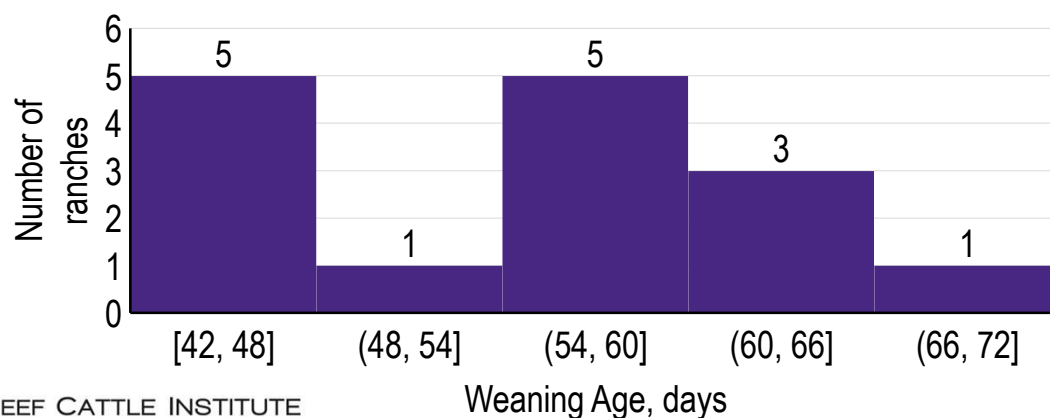


https://cdn11.bigcommerce.com/s-h1vr2w36j/images/stencil/1280x1280/products/19419/22305/KJC50406__54049.1696449085.jpg?c=1

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Results: Weaning

- Weaning age ranged between 42 and 72 days
- 87% of operations gradually weaned calves



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Results: Calf Health

- Treatment of respiratory disease:
 - Antimicrobial (73%; 11 ranches)
 - Antimicrobial & anti-inflammatory drug (27%; 4 ranches)
- Treatment of digestive disease:
 - IV fluids (7%; 1 ranch)
 - Electrolytes (7%; 1 ranch)
 - Electrolytes and additional therapies (86%; 13 ranches)

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Conclusions

- This research highlights the knowledge gap in the industry
- Management of beef-on-dairy calves within commercial calf ranches has not been previously described
 - Speculation about differences between management of dairy and beef-on-dairy calves (Machado & Ballou, 2022)
- There is no “standard” way to manage these calves in this setting

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Questions?

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Automated Machine Learning and Facial Imaging for Feedyard Cattle Outcome Prediction

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Objective

This study evaluated the potential of facial images taken at time of BRD treatment to predict feedyard cattle outcomes (Finish or Did not Finish) following 60 days post-treatment

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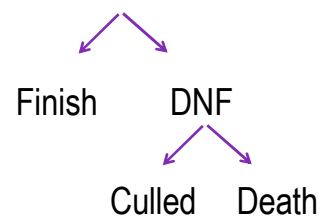
Methods

Cross-sectional
observational study

2 hospitals in a
feedyard in the High
Plains region



Outcomes
determined 60 days
post-enrollment



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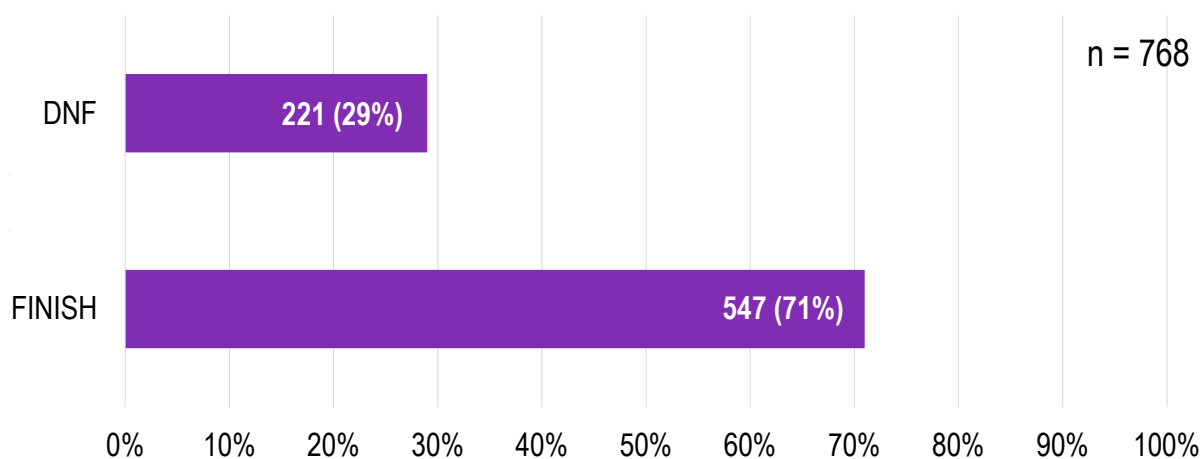
Microsoft Azure



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Label Classification Distribution



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Guess what Finish or Did not Finish?

- Hands up for Finish
- Hands down for DNF

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DNF

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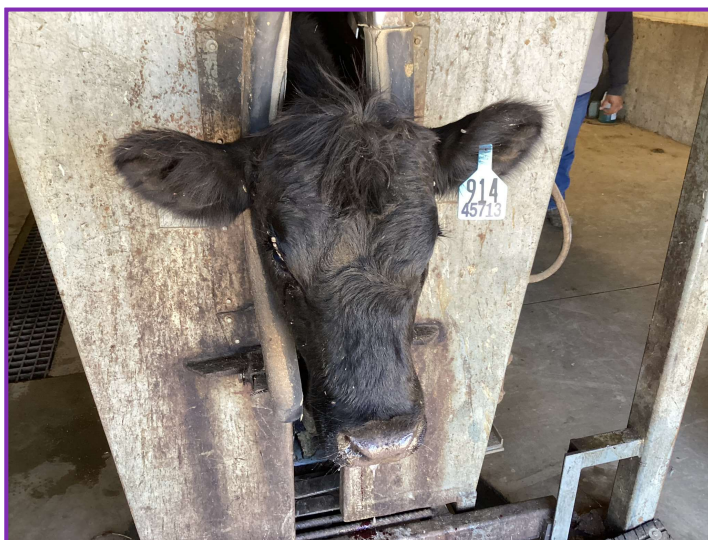


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Finish

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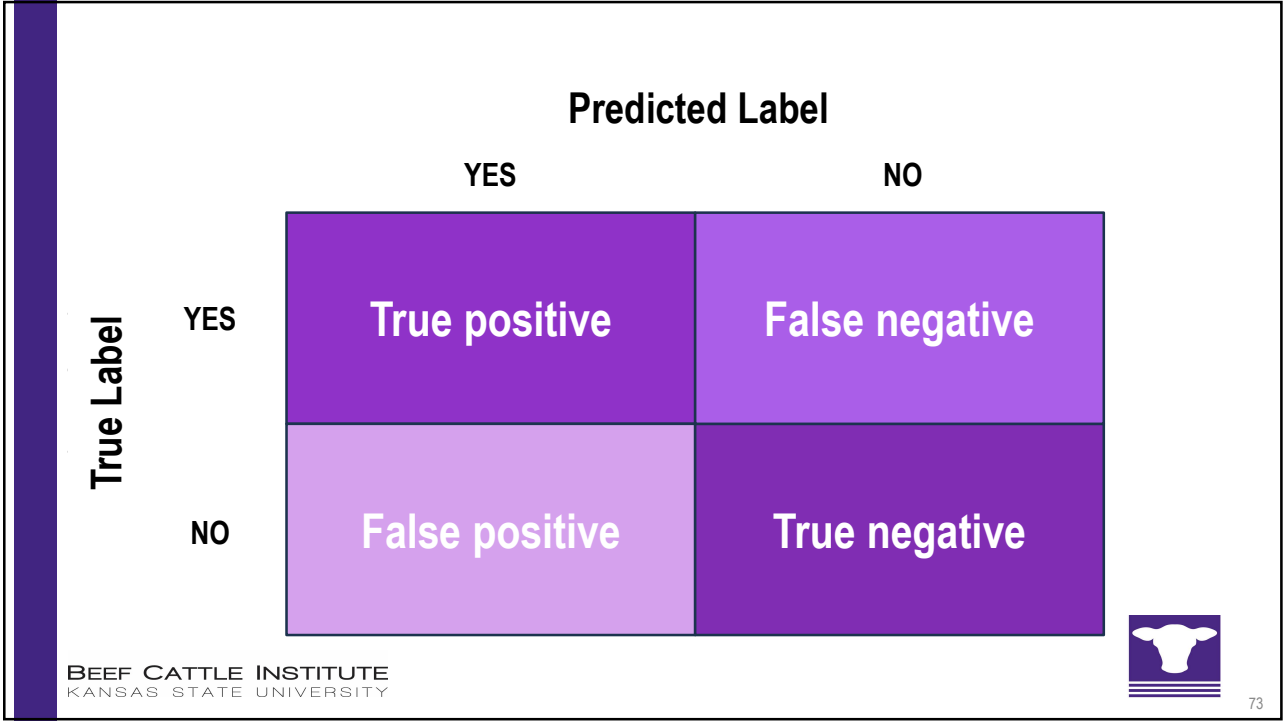


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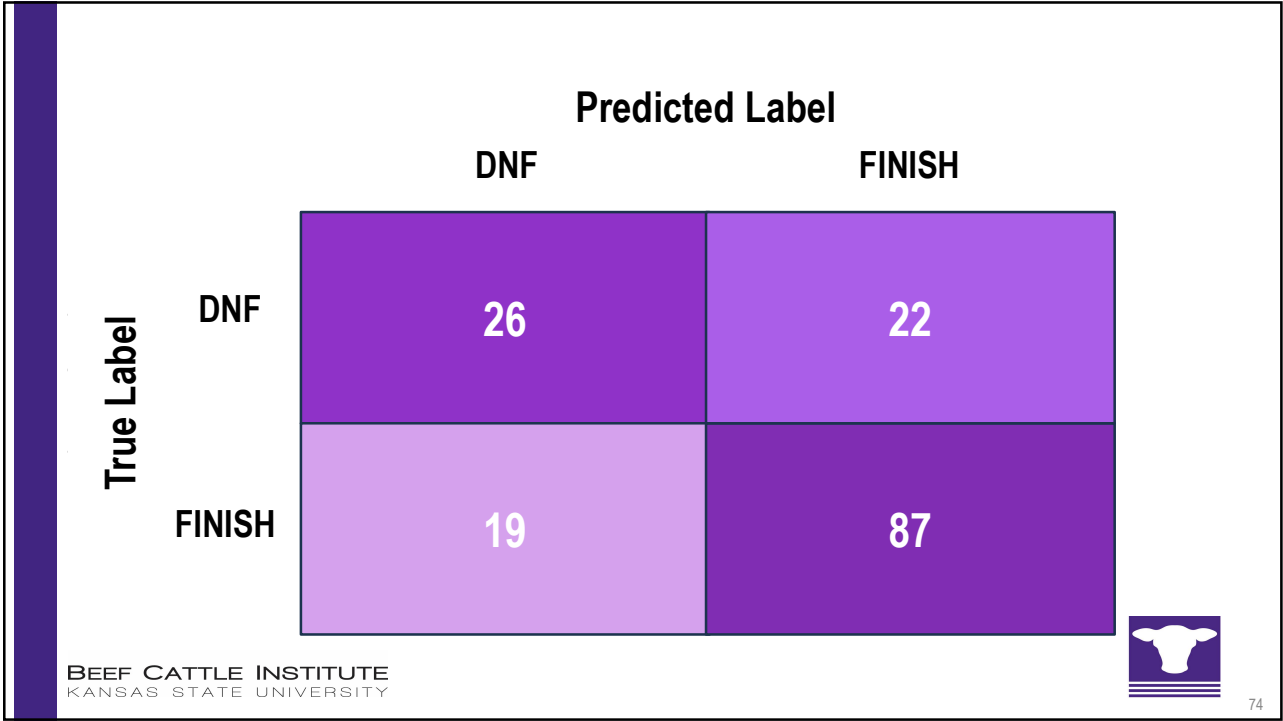
DNF

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72




73



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| | | Predicted Label | |
|------------|--------|-----------------|--------|
| | | DNF | FINISH |
| True Label | DNF | 26 | 22 |
| | FINISH | 19 | 87 |

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Model performance

| Class | Se% ² | Sp% ³ | PPV% ⁴ | NPV% ⁵ | AUC ⁶ | Acc ⁷ |
|------------------|------------------|------------------|-------------------|-------------------|------------------|------------------|
| DNF ¹ | 54 | 82 | 57 | 79 | 0.69 | 0.73 |

¹DNF – Did Not Finish; ²SE – Sensitivity; ³SP – Specificity; PPV - ⁴Positive Predictive Value; ⁵NPV – Negative Predictive Value; ⁶AUC – Area Under the Curve; ⁷Acc – Max Accuracy.

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Conclusion

Facial imaging-based models showed potential for predicting cattle outcomes in this dataset.

Combining these models with other diagnostic tools could improve management strategies in feedyard operations.

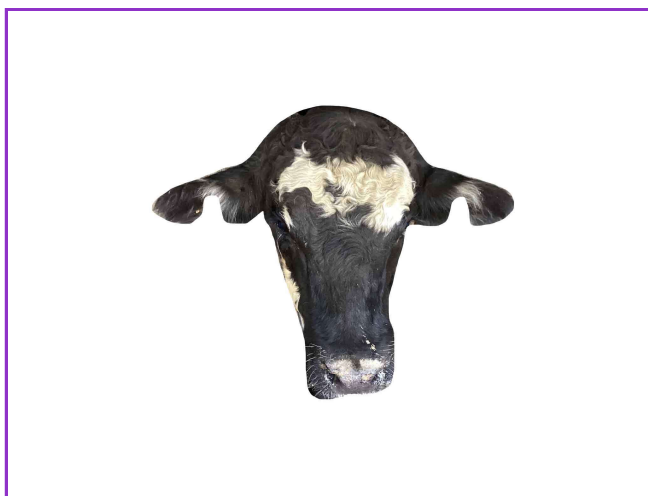
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Limitations and next steps



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Limitations and next steps

- Background and ear tag removal
- Light exposure and image shadow
- Model refinement

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Questions?
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What can predictive models do for the industry?

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Issues we face...

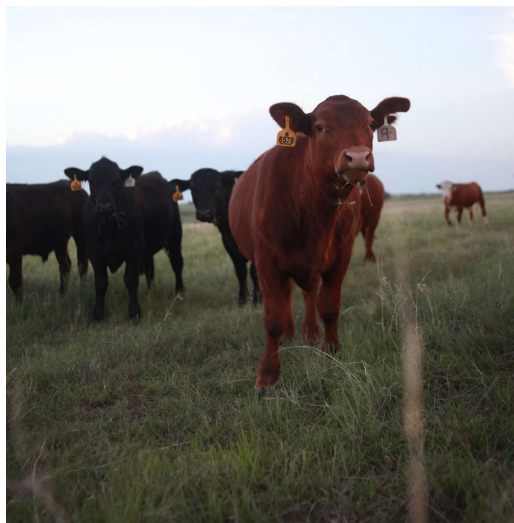
- Demand for a high-quality product at an affordable price
 - How do I still make money?
 - How do I account for the ever-changing market?
 - Efficiency!
- Huge amounts of data sitting unused
- Balance the consequences of risk-taking

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Where do predictive models fit in?



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BCI Predictive Models

- Utilize vast amounts of feedlot data to...
 - Predict individual animal outcome at the time of first and second BRD treatment
 - Predict which lots of cattle will experience high morbidity due to BRD ($\geq 15\%$)
 - Determine which lots of cattle should receive metaphylaxis based on economic outcome

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BCI Predictive Models

- Types of data
 - Feed delivery data
 - Cattle demographic characteristics → lot arrival weight, sex, etc.
 - Data collected at treatment → rectal temperature, weight
 - Weather data → precipitation, ambient temperature, wind speed, etc.
- Types of models/algorithms
 - Linear techniques → logistic regression
 - Non-linear techniques → decision tree, neural network, random forest

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Evaluation of predictive models to determine metaphylaxis application to cattle arriving at the feedlot

Outcome of interest: best metaphylaxis application strategy (no metaphylaxis or metaphylaxis)

Data

- Demographic data at arrival
- Origin data
- External economic data

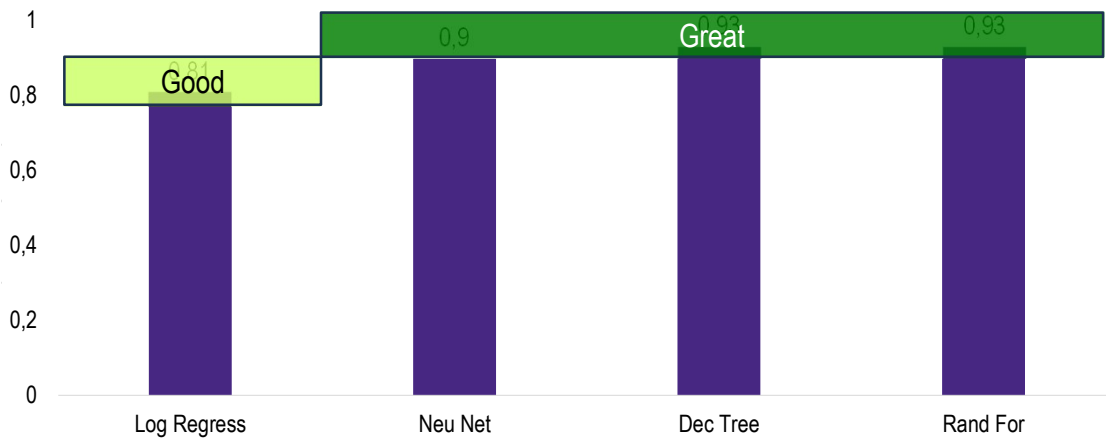


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Model Results – Demographic Data Alone



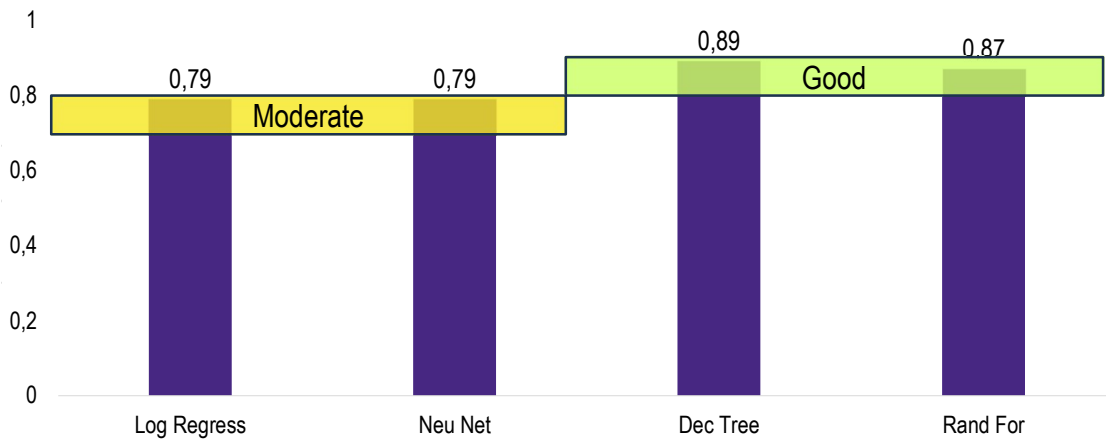
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Model Results – Incl. Origin Data



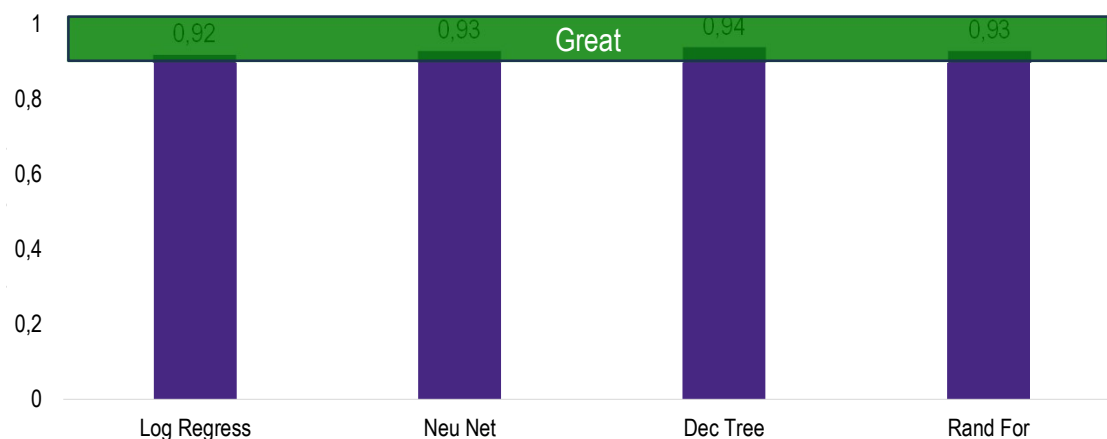
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Model Results – Incl. External Econ Data



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Key Takeaways

- Predictive model application to feedlot challenges is feasible
- Process and good data matters
- Will not and CANNOT replace the labor on the ground



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Questions?
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