





RightMate: Precision Breeding by Genomic Mating







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Breeding Decisions: Customer Needs and priorities

- Balanced Trait Mating?
- Terminal Mating?
- Heifer Mating?
- Multiply Priorities?
- Genetic Improvement, Genetic Diversity & Uniformity?







Complexity of Animal Breeding:

ORIGINAL ARTICLE



Complexity of animal breeding choice making

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Abstract

The last decades of developments in animal breeding, genetics, genomics and data recording technologies have allowed the evaluation of larger numbers of animal and animal traits than ever before. This should help make animal breeding choices more informed, but it also makes them far more complex. This complexity may be overwhelming farmers, thus compromising realization of potential genetic gain in livestock industries. However, the effect of complexity of animal breeding choices on farmers' selection of animals has received very little theoretical consideration to date. This paper reviews the theoretical principles of complex decisions, contextualizes the findings to the field of animal breeding, and analyses how farmers and the animal breeding industry are currently dealing with complexity. According to the findings of the analyses of complex decisions in other fields, the complexity of animal breeding choices is likely to lead to farmers using conscious or unconscious simplifying strategies (heuristics) to handle such complexity. When these heuristics are ineffective, poor selection decisions and a potential loss of genetic progress can be expected. Further, studies using survey experiments to understand farmer behaviour and selection preferences may be compromised by the complexity of the survey's

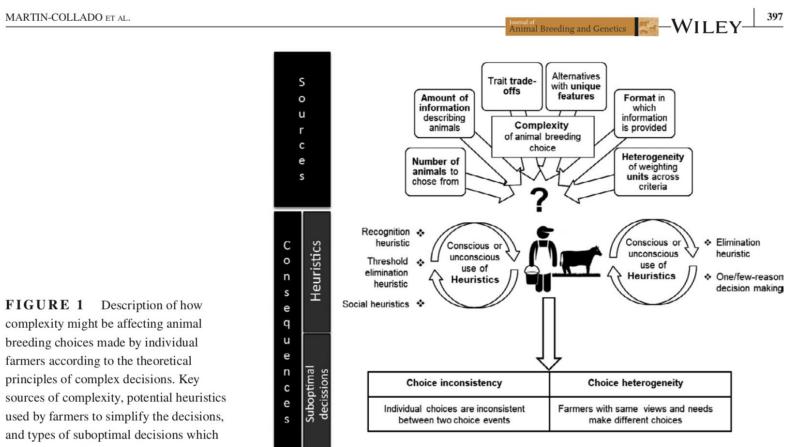


MARTIN-COLLADO ET AL.





Complexity of Animal Breeding:



principles of complex decisions. Key sources of complexity, potential heuristics used by farmers to simplify the decisions, and types of suboptimal decisions which may occur due to that complexity







So...What is RightMate?



- Genomic based Precision Breeding Tool.
- Primarily Designed for Seedstock Providers.
- Generations of Improvement per Mating Cycle.
- More Uniform Calf Crops.
- Optimized genetic diversity/Improvement with customized breeding objectives by integrating into MateSel software.
- Uses both EPDs and the Power of High Impact Genes.







Segelke et al. Genetics Selection Evolution 2014, **46**:42 http://www.gsejournal.org/content/46/1/42



RESEARCH Open Access

Prediction of expected genetic variation within groups of offspring for innovative mating schemes

Dierck Segelke^{1,2*}, Friedrich Reinhardt¹, Zengting Liu¹ and Georg Thaller²

Abstract

Background: Experience from progeny-testing indicates that the mating of popular bull sires that have high estimated breeding values with excellent dams does not guarantee the production of offspring with superior breeding values. This is explained partly by differences in the standard deviation of gamete breeding values (SDGBV) between animals at the haplotype level. The SDGBV depends on the variance of the true effects of single nucleotide polymorphisms (SNPs) and the degree of heterozygosity. Haplotypes of 58 035 Holstein animals were used to predict and investigate expected SDGBV for fat yield, protein yield, somatic cell score and the direct genetic effect for stillbirth.

Results: Differences in SDGBV between animals were detected, which means that the groups of offspring of parents with low SDGBV will be more homogeneous than those of parents with high SDGBV, although the expected mean breeding values of the progeny will be the same. SDGBV was negatively correlated with genomic and pedigree inbreeding coefficients and a small loss of SDGBV over time was observed. Sires that had relatively low mean gamete breeding values but high SDGBV had a higher probability of producing extremely positive offspring than sires that had a high mean gamete breeding value and low SDGBV.

Conclusions: An animal's SDGBV can be estimated based on genomic information and used to design specific genomic mating plans. Estimated SDGBV are an additional tool for mating programs, which allows breeders to identify and match mating partners using specific haplotype information.







Table 3 Correlations (r) between SDGBV with real progeny variations for different traits per minimum number of offspring per sire

Minimum number of offspring per sire	Number of sires	r _{FY}	r _{PY}	r _{scs}	r _{SBd}
10	409	0.65	0.56	0.60	0.50
50	146	0.90	0.78	0.80	0.72
100	84	0.93	0.83	0.88	0.69
150	48	0.93	0.90	0.90	0.78
200	32	0.93	0.91	0.87	0.85
300	20	0.96	0.93	0.94	0.82
500	7	0.98	0.88	0.90	0.90

PY = protein yield; FY = fat yield; SCS = somatic cell score; SBd = the direct genetic effect for stillbirth.







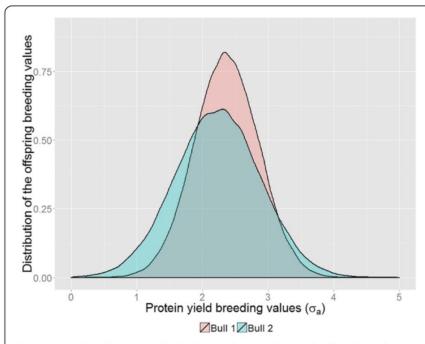


Figure 5 Distribution of the breeding values of offspring for protein yield. Two bulls (with MGBV equal to 1.81 σ_a and 1.68 σ_a and SDGBV equal to 0.29 σ_a and 0.52 σ_a , respectively) are mated with an average female of the population (MGBV equal to 0.55 σ_a , SDGBV equal to 0.39 σ_a).

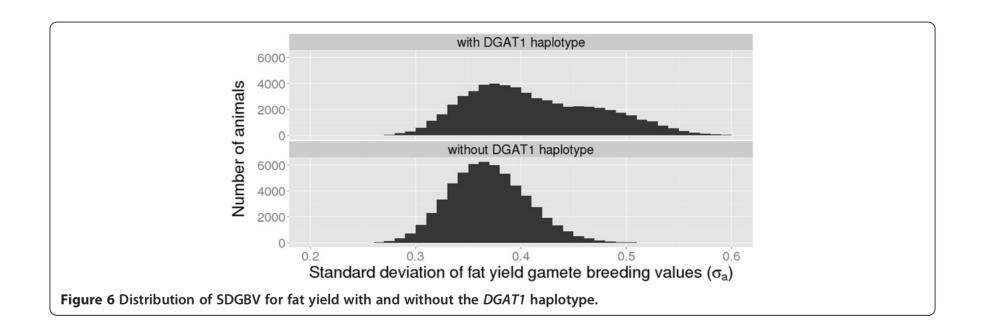






Segelke et al. Genetics Selection Evolution 2014, **46**:42 http://www.gsejournal.org/content/46/1/42

Page 8 of 10



GENETIC RESOURCES



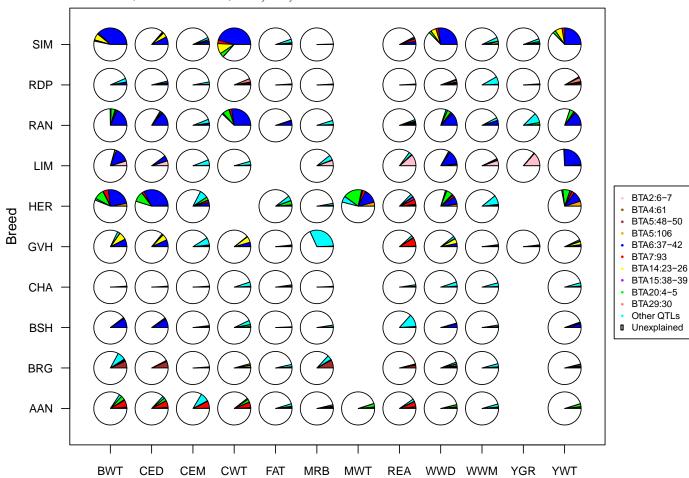


RESEARCH ARTICLE

Open Access

Large-effect pleiotropic or closely linked QTL segregate within and across ten US cattle breeds

Mahdi Saatchi¹, Robert D Schnabel², Jeremy F Taylor² and Dorian J Garrick^{1,3*}









High Impact Genes (markers): USAMRC

Could functional variants be more effective?

Genetic correlations between birth weight and GPE-trained birth weight MBV							
		GPE	<u>Eva</u>	Evaluated population			
Marker set	size	h ²	SFA	Red Angus	Simmental		
F250 shared with 50K	33,869	0.45	0.35	0.44	0.25		
Significant GPE effects	279	0.34	0.44	0.43	0.25		
LD reduced	12	0.30	0.49	0.47	0.28		
NCAPG	1	0.06	0.31	0.32	0.22		

Continued to the state of the s

Small sets of functional variants can explain meaningful phenotypic variation within and across populations

 depends on number and size of effects - difficult to identify variants causing sm effects, especially for traits influenced by many variants with small effects

BeefImprovement.org/Symposium • #BIF2020



Larry Kuehn, Warren Snelling, Mark Thallman, BIF 2020.







Which bull do you choose as a heifer bull?

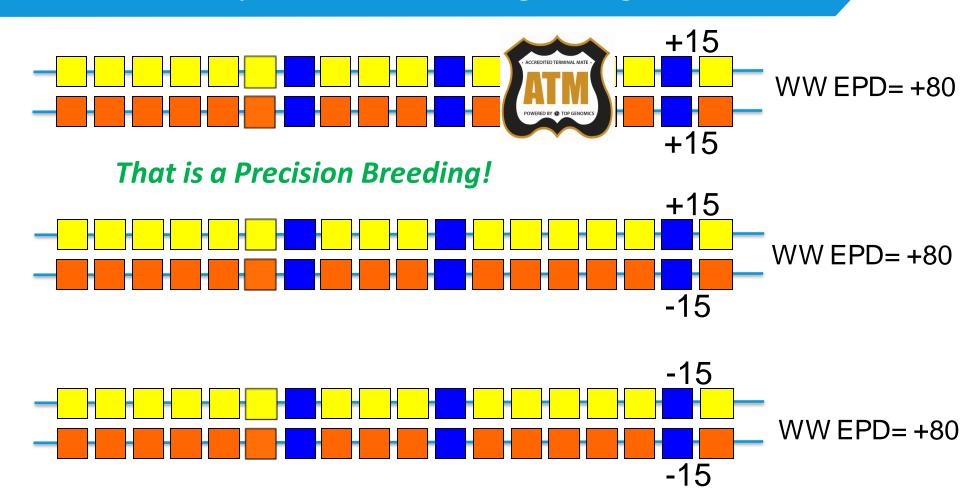








Which bull do you mate to a low growing cow?

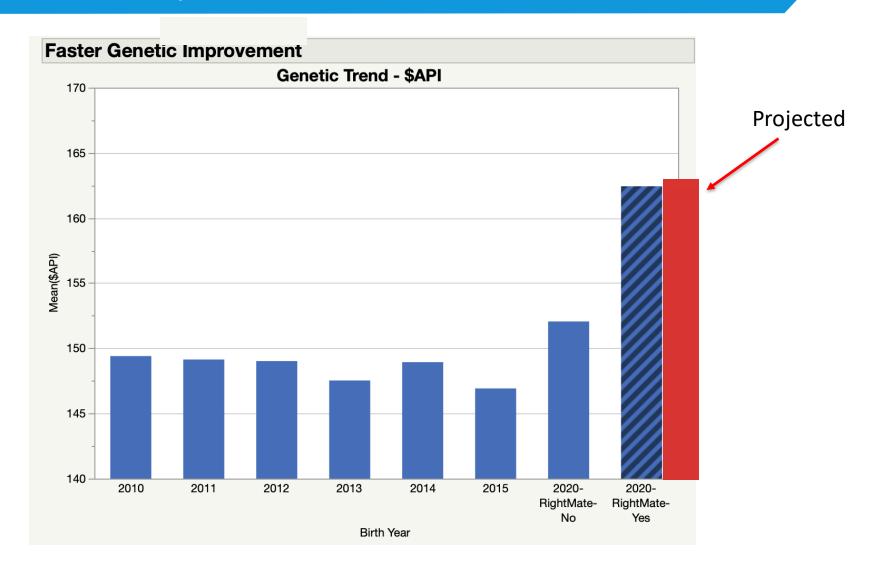








Faster Genetic Improvement:

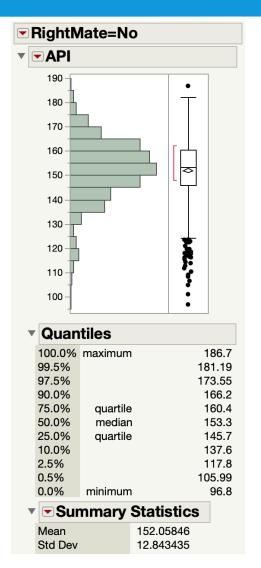


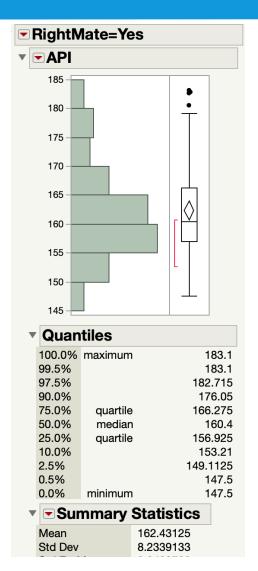






RightMate Validation: API



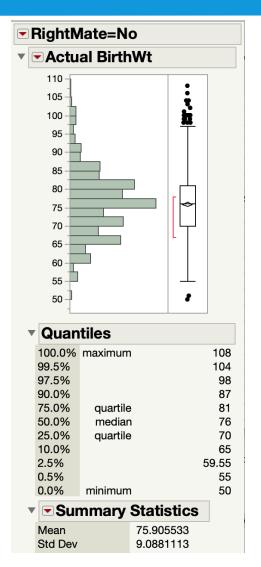


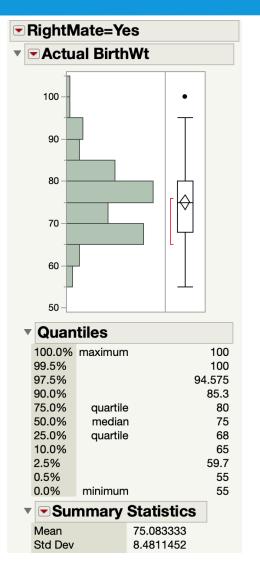






RightMate Validation: Actual Birth Weight



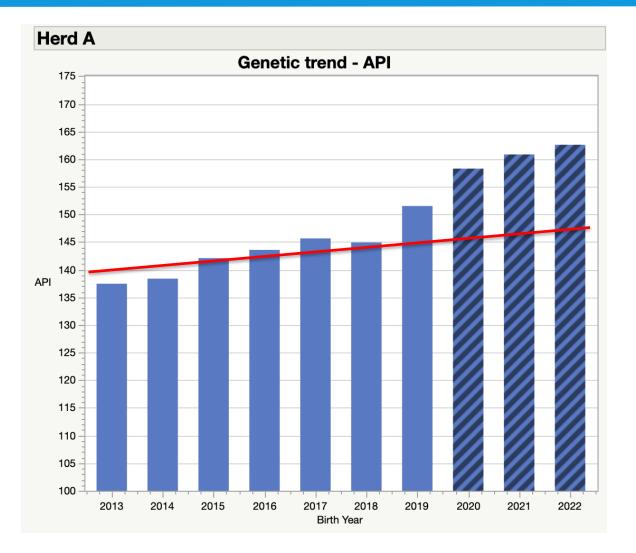








Herd A - Genetic Improvement:



Before RM: +2.09/year

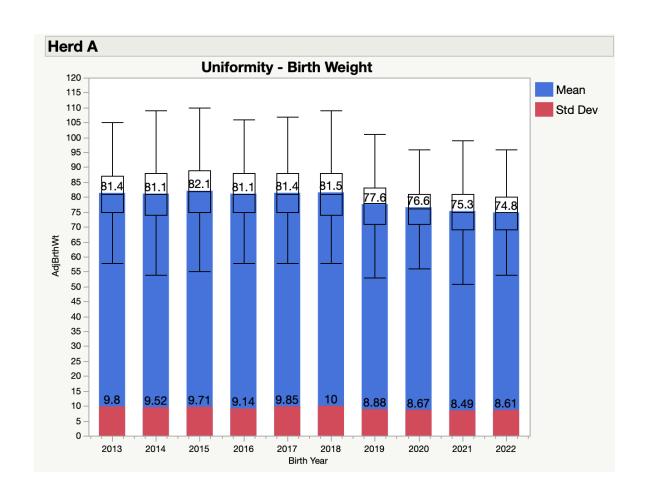
After RM: +3.60/year







Herd A - Uniformity:

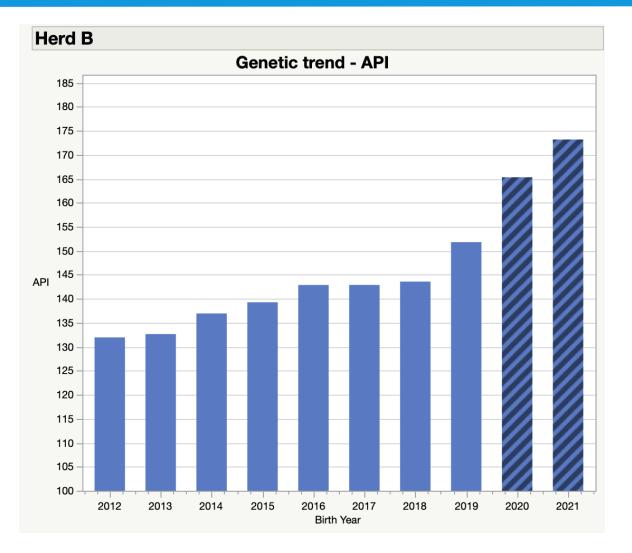








Herd B - Genetic Improvement:



Before RM: +2.55/year

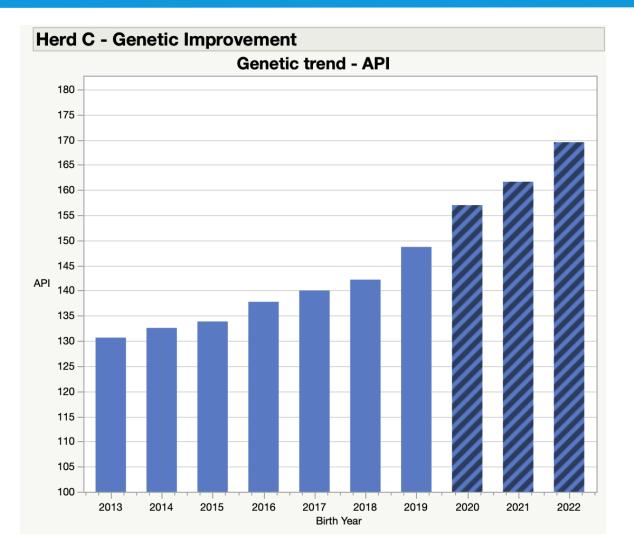
After RM: +11.2/year







Herd C - Genetic Improvement:



Before RM: +2.99/year

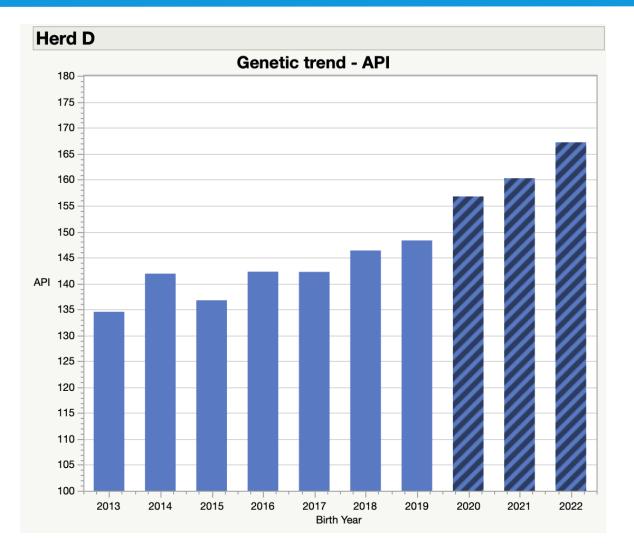
After RM: +6.67/year







Herd D - Genetic Improvement:



Before RM: +2.05/year

After RM: +5.91/year

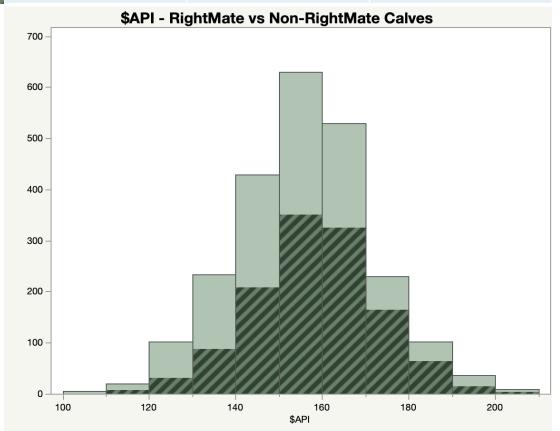






RightMate Results: 2021 Born Calves

Mate	N Calves	\$API		
Non-RightMate	1274	152.8		
RightMate	1242	158.2		



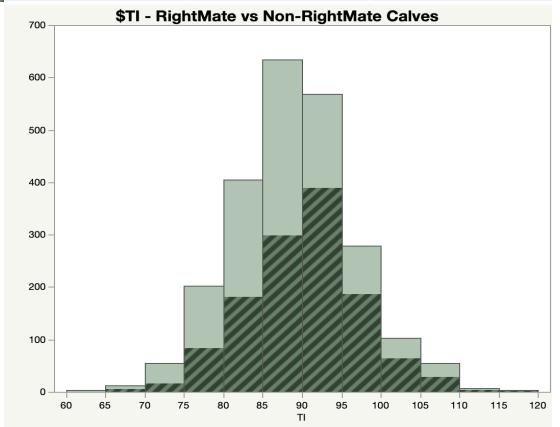






RightMate Results: 2021 Born Calves

Mate	N Calves	\$TI
Non-RightMate	1274	87.4
RightMate	1242	90.2









RightMate Results: 2021 Born Calves

Traits	Non- RightMate	RightMate	Diff.	Traits	Non- RightMate	RightMate	Diff.
CE	13.6	14.2	+0.8	STAY	17.7	18.1	+0.4
BW	-0.3	-0.6	-0.3	CW	36.4	37.3	+0.9
WW	80.2	81.7	+1.5	YG	-0.26	-0.26	0
YW	125.6	128.3	+2.7	MRB	0.46	0.51	+0.05
ADG	0.28	0.29	+0.01	BF	-0.05	-0.04	+0.01
MCE	7.8	8.1	+0.3	REA	0.80	0.83	+0.03
MILK	26.2	25.7	-0.5	SHR	-0.35	-0.37	-0.02
MWW	66.3	66.5	+0.2				







Commercial Application:



Simplify the decision making and reduce the risk

The 3 Accreditation Logos are designed to simplify bull selection and improve buyer confidence and reducing risk.





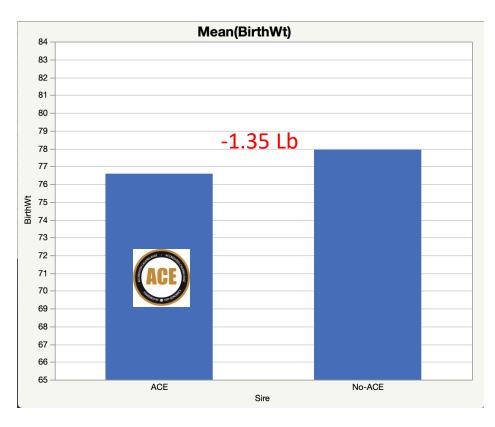


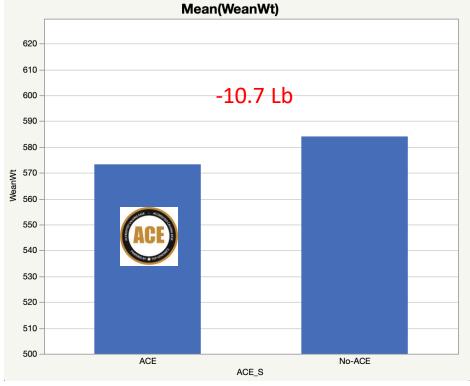






Right Choice powered by G-TOP GENOMICS



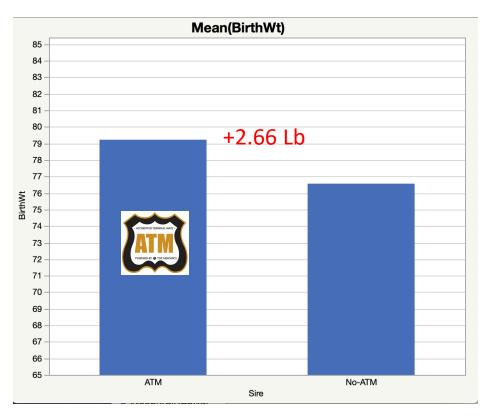


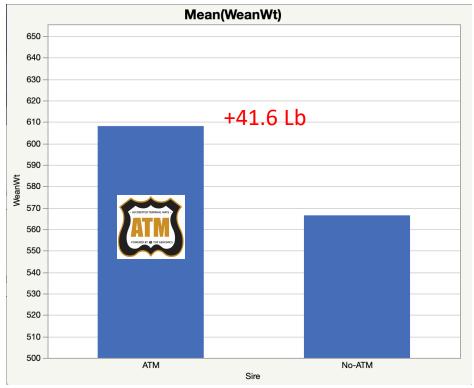






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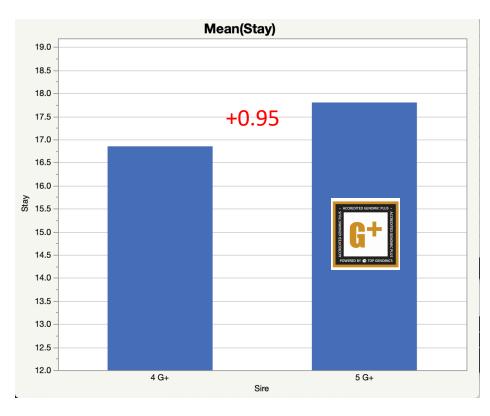


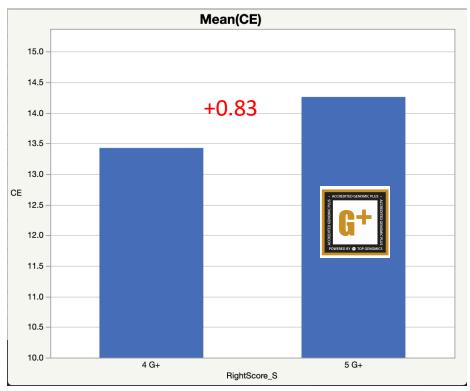










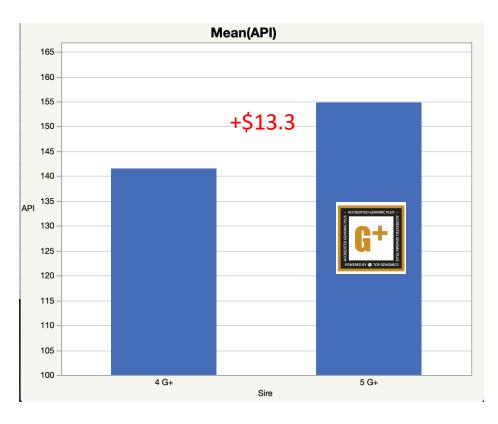








Right Choice powered by G-TOP GENOMICS











RightMate Benefits:

- Precision Breeding using Genomics
- Avoid Wasted Mating to offer more Marketable Bulls and Heifers
- Reduce Variation and More Uniformity
- Genetic Improvement at a Higher Speed
- Managing Inbreeding, Genetic Diversity and Customized Breeding using MateSel







2023 RightMate Statistics (So far):

✓ >11,681 Cows RightMated from >75 Herds

✓ > 4,400 Bulls/Cows RightChoice evaluated from >62 Herds







RightMate Added-Value for Beef Industry:

- √ >10,000 Cows RightMated
- ✓ >5,000 RightMate Calves
- ✓ >2,000 RightMate Bull Calves goes to Commercial Herds
- √ ~100 calves per commercial herd bull.
- √ >\$5 extra genetic improvement (\$API)
- ✓ > \$1,000,000/year RightMate added-value in commercial beef sector.
- ✓ Much more \$ added value for replacement heifers, Seed-stock and AI bulls!







Acknowledgments:

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Corey Wilkins, Rocky Forseth, Clint Berry, Jared and Jullie Murnin







Thank you!

Genomics wins!

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