

Genomic insights for feeding behavior traits in beef cattle

Tiago S. Valente, Mohammed K. Abo-Ismail, John Crowley, John A. Basarab <u>tsvalente@ualberta.ca</u> & Graham Plastow



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Introduction

Animal behavior studies have been viewed as an attempt to look at the entire production system, assessing individual activities regarding the social and physical environment that animals are exposed to.

Feeding behavior related traits:

- Respiratory diseases (Quimby et al., 2001; González et al., 2008)
- Metritis (Weary et al., 2009)
- Evaluation of acidosis (Schwartzkopf-Genswein et al., 2003; Robles et al., 2007)
- Strategies for bunk management (Schwartzkopf-Genswein et al., 2002; 2004)





Introduction

Important aspects of feeding behavior:

> Individual variability in feed efficiency and performance traits;

- > Individual capacity of adaptability to the environment provided by man;
- > Animal health status and welfare

All these aspects are of considerable importance for the development of sustainable breeding programs





Introduction

There is a need to understand the genetic and biological mechanisms underlying feeding behavior traits and their association with feed efficiency, performance and carcass traits.

Identify genomic regions and candidate genes associated with feeding behavior traits including feeding event frequency and daily feeding duration in crossbred beef cattle





Automated GrowSafe feeding system (GrowSafe Systems Ltd., Airdrie, Alberta, Canada)

Performance tests carried out between 2003 and 2013

Average test duration was 80 ± 6 days

Crossbred beef cattle

• Angus, Hereford, BeefBooster, Kinsella Composite and Charolais

Illumina BovineSNP50 BeadChip array

- SNP markers with unknown genomic position
- located on sex chromosomes
- monomorphic and markers with MAF < 0.01
- HWE (p-value < 1⁻⁶)
- SNPs and Individuals with call rate < 0.90







Feeding Event Frequency

 \sum feed events/performance duration (d)

event = bunk visit with feed consume > 0Kg and intervals between the last 2 readings of the same transponder > 300s. Feeding Event Frequency

 \sum feed duration time/performance duration (d)

duration time (min) = difference between the feeding event end and start times.

Basarab et al. (2003)





Weighted single-step GBLUP (WssGBLUP) methodology using BLUPF90 programs (Wang et al., 2012; Misztal et al., 2016)

Single-trait animal model

$$y = X\beta + Za + e$$

- direct additive genetic effects = $a \sim N(0, H\sigma_a^2)$
- residual effect = $e \sim N(0, I\sigma_e^2)$

Fixed effect - Contemporary groups (animal sex, herd of origin, year of birth, and management group)

Covariates - Animal weight at the beginning of the test Breed composition (ADMIXTURE software, Alexander et al. 2009)





Summary of data set structure and descriptive statistics

	FREQ (events/d)	DUR (min/d)
No. animals	3,509	3,511
Mean ± SD	55.94 ± 29.92	100.89 ± 37.53
Live weight (kg)	373	± 74
No. genotypes (after QC)	3,5	29
No. SNPs (after QC)	37,2	298

Post GWAS analysis

andForestry





CattleQTLdb

(Durinck et al., 2009; Huang et al., 2009; Hu et al., 2013)





Results

Manhattan plot of additive genetic variance explained by windows of 20 adjacent SNPs for:



Chromosome



Results

Summary of top 10 SNP-windows with major effects on feeding behavior traits.

Feeding event frequency						
BTA ¹	Location (bp)	Var (%) ²				
1	25,685,179 - 27,465,349	1.20				
2	63,516,470 - 65,069,037	1.82				
6	44,622,597 - 45,412,181	1.03				
7	29,579,813 - 31,136,178	0.94				
9	20,247,290 - 21,490,009	0.85				
9	49,865,815 - 50,804,011	1.38				
12	45,002,070 - 45,952,853	1.11				
13	8,347,568 - 9,556,357	1.97				
14	11,154,590 - 11,983,913	1.08				
16	35,723,500 - 36,565,106	1.58				
	17 genes	12.96%				
) J	ivestock ¹ B ⁻	TA = Bos tauru				

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	Daily feeding duration					
	BTA¹	Location (bp)	Va	Var (%) ²		
Aoficultu	1	52,748,123 - 53,656,	,600	0.80		
	3	18,520,392 - 19,860,	,064	0.88		
	4	68,528,799 - 69,831,	,612	0.80		
	5	119,261,609 - 120,378	8,417	0.78		
	8	33,747,904 - 34,714,	,805	0.97		
	13	63,257,337 - 65,006,	,713	2.32		
	20	22,823,334 - 24,228,836		0.91		
	23	5,896,623 - 7,276,902		1.15		
	23	9,453,816 - 10,688,252		0.93		
	28	22,760,601 - 23,74 <mark>2,9</mark> 25		0.73		
		163 genes		10.27%		

¹BTA = *Bos taurus* chromosomes; ²Var = Proportion of additive genetic variance



Results

SNP windows overlapped with previously reported QTLs for feed efficiency (QTLdb database)

• DMI; ADG; RFI; FCR; YG

Feeding Event Frequency

• *PPARGC1A* gene on BTA6 (44Mb) (peroxysome proliferator-activated receptor-gamma coactivator-1alpha)

Modulate feed intake (GO: 0002021)

Associated with intramuscular fat deposition, productive and reproductive traits in beef cattle (Ramayo-Caldas et al., 2014)

Feeding Event Frequency

RORC, TNFAIP8L2 on BTA3

- *BPIFB1* on BTA 13
- DBS, PSMB8, PSMB9, BOLA-DMA, BOLA-DMB, BOLA-DOA, BOLA-DOB, BOLA-DYB on BTA23

Inflammatory bowel disease (KEGG_PATHWAY)

Associated with intramuscular fat, marbling and carcass weight in different cattle breeds (Barendse *et al.,* 2010)





Conclusion

Identified major SNP windows and potential candidate genes, providing new insights for the biological understanding of feeding behavior in beef cattle.

Corroborate the possibility to use animal behavior as tool for early detection of health status and performance.

Further functional genomic studies are important to better understand the relevant pleiotropic pathways underlying feeding behavior and feed efficiency traits.





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