



Sequence-based GWAS meta-analyses for beef production traits

M.-P. Sanchez*, T. Tribout, N. Kadri, P.K. Chitneedi, S. Maak, C. Hozé, M. Boussaha, P. Croiseau, R. Philippe, M. Spengeler, C. Kühn, Y. Wang, H. Pausch, D. Boichard

* marie-pierre.sanchez@inrae.fr

INRAE

ETH zürich













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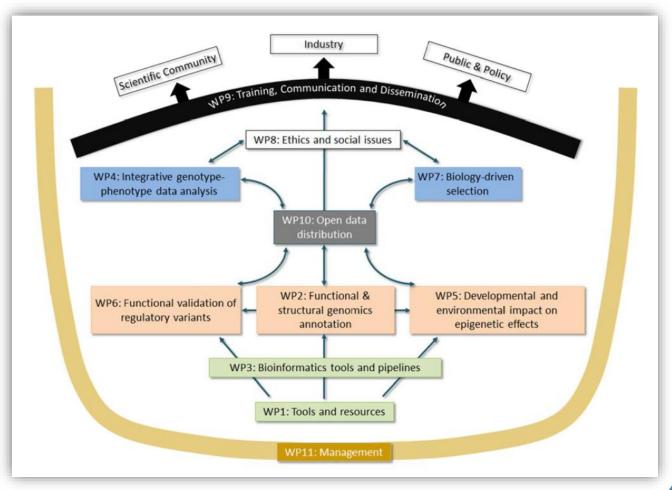
H2020 BovReg project

Identification of functionally active genomic features relevant to phenotypic diversity

and plasticity in cattle

20 partners from 14 countries





11 WP incl. WP4
Integrative analysis of genotype-phenotype

WP4 – Integrative genotype-phenotype data analysis

WP leader: Hubert PAUSCH (ETH, Switzerland) - 11 partners / 20 involved in BovReg

T4.1 – Hubert PAUSCH (ETH, Switzerland)

GWAS and meta-analyses from whole-genome sequences (WGS) for biological efficiency, disease resistance and fertility traits

T4.2 – Carole CHARLIER (GIGA, Belgium)
Phenotypic impact of mobile element integration

T4.3 – Christa KUHN (FBN, Germany) eQTLs and mQTLs analyses

T4.4 – Emily CLARK (UEDIN, UK)
Tools to prioritize candidate causative variants





T4.1 – GWAS & meta-analyses

Within-breed GWAS







4 groups of traits

Mastitis resistance

Milk yield & fertility

Feed efficiency

Beef traits

Meta-analyses









8-13 populations > 120 000 anim.

7-12 populations > 125 000 anim.

3-9 populations> 13 000 anim.

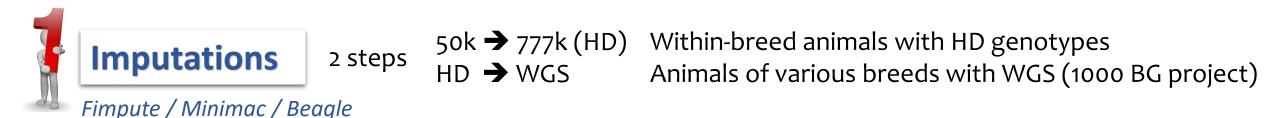
3-10 populations > 25 000 anim.

Bulls, cows, steers from purebred or crossbred populations Phenotypes (with weights): YD, DYD, DRP, AP



T4.1 – Within-breed GWAS & meta-analyses

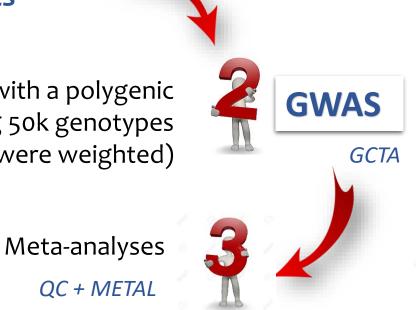
All partners applied similar imputation and GWAS workflows before meta-analyses





tens of millions of variants incl. causal variants

Linear mixed model to test individual variant effect together with a polygenic effect estimated from a GRM built using 50k genotypes (when required, phenotypes were weighted)





T4.1 – « Beef » MA – populations

A large number and a large diversity of populations

図 8 purebred populations from 5 French breeds (NOR, MON, CHA, LIM, BLA)













■ 4 populations from Swiss breeds (BSW, OBR)





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☑ 2 crossbred populations from Germany (HOL x CHA)









□ 1 composite line from Canada (ANG, CHA, beef)











T4.1 – « Beef » MA – traits analyzed

A large number and a large diversity of traits

☑ Growth (6)

☑ Morphology (6)

☑ Carcass (21)

1 Growth	Birth Weight	BW
2 Growth	weight at month 15	W15
3 Growth	weight at 18 months	W18
4 Growth	weight at 24 months	W24
5 Growth	average daily gain	ADG
6 Growth	average daily gain during fattening	ADG
7 Morphology	muscularity score	MS30
8 Morphology	skeletal score	SS30
9 Morphology	thickness of bones	TB30
10 Morphology	Thighs	THIGHS
11 Morphology	Wither	WITHER
12 Morphology	Fat score	FS
13 Carcass	carcass weight	CW
14 Carcass	fat coverage	CF
15 Carcass	meatiness	MT
16 Carcass	Area of longissimus thoracis	ALT
17 Carcass	Carcass conformation	CC
18 Carcass	carcass fat score	FS
19 Carcass	carcass yield	CY
20 Carcass	Internal fat weight	IFW
21 Carcass	length of the leg	LL
22 Carcass	Rib Eye Area	REA
23 Carcass	Weight at slaughter	WS
24 Carcass	Maximum width of the thigh	WT
25 Carcass	age at slaughter	AS
26 Carcass	carcass grade	CG
27 Carcass	average backfat thickness	ABT
28 Carcass	hot carcass weight	CW
29 Carcass	lean meat yield	LMY
30 Carcass	fat content of 6th rib	FC6
31 Carcass	fat content measured by ultrasound	FCU
32 Carcass	muscular development	MD
33 Carcass	skeletal development	SD

Grouping of traits in 16 MA

MA Trait type	Traits	# traits	# pop.	# partners	# anim.
1 Growth	W15/W18/ADG	3	7	4	18774
2 Growth	BW	1	5	2	2720
3 Morphology	MS30/THIGHS/CC	3	6	2	17418
4 Morphology	MS30/WITHER/CC	3	6	2	17418
5 Morphology	LL	1	5	2	3695
6 Morphology	WT	1	5	2	3695
7 Morphology	SS30/SD	2	4	2	12140
8 Carcass	CW	1	7	4	19989
9 Carcass	AS	1	6	2	12208
10 Carcass	CY	1	5	2	3694
11 Carcass	CG/LMY/MT/CC	4	10	5	25367
12 Carcass	FS/ABT/FC6/FCU/CF	5	8	5	14622
13 Carcass	WS	1	5	2	2636
14 Carcass	ALT	1	5	2	3692
15 Carcass	IFW	1	5	2	3686
16 Carcass	REA	1	3	2	4453

☑ 1 to 5 traits / MA

☑ 3 to 10 populations / MA

☑ 2 to 5 partners / MA

☑ 2600 to 20,000 animals / MA





T4.1 – « Beef » MA – methods

2 methods used

METAL software (Willer et al., 2010)

z-score

For each variant, Z is calculated by combining the p-value (p_i) associated to its effects in the different GWAS, weighted by the sample size (w_i)

$$Z = \frac{\sum_{i} Z_{i} w_{i}}{\sqrt{\sum_{i} w_{i}^{2}}}$$

$$Z_{i} = \Phi^{-1} \left(1 - \frac{P_{i}}{2}\right)^{*} \text{(effect direction for study } i\text{)}$$

Fixed effects

Normalized <u>effect</u> of each variant estimated in the GWAS $i(\theta_i)$ combined and weighted by the inverse of the error variance (w_i)

$$\hat{\theta}_{F} = \frac{\sum_{i} w_{i} \hat{\theta}_{i}}{\sum_{i} w_{i}}$$

→ The fixed effects method is generally more powerful but as variant effects are considered as identical between GWAS, traits analyzed in GWAS need to be identical and measured in the same unit => standardization of the effects by the genetic SD of the trait specific to each population

Effect of a variant was considered significant if $-\log_{10}(p\text{-value}) \ge 8.7$

=> 5% threshold after Bonferroni correction (~25M variants)

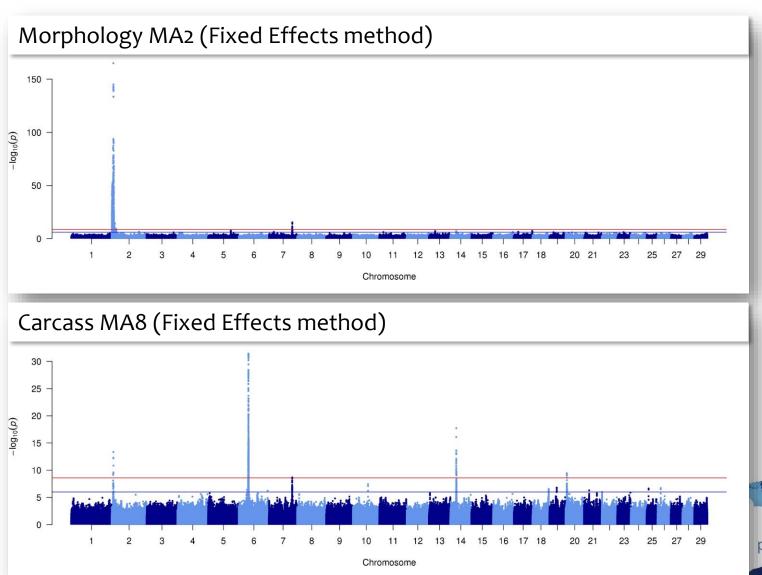
Significant results (QTL) for 15 of the 16 MA on 11 bovine autosomes

The most significant QTL located

On chromosome 2

On chromosome 6

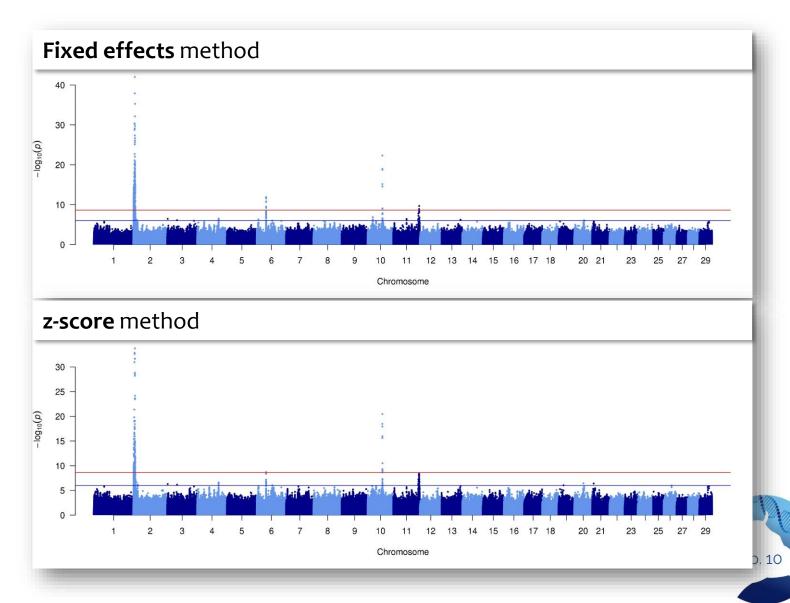




Ex: carcass MA11

Fixed effects vs **z-score** method:

- → QTL generally found with more significant effects
- → More QTL detected





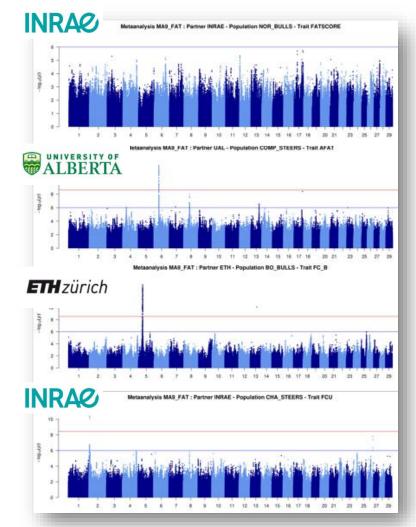
MA vs within-breed GWAS

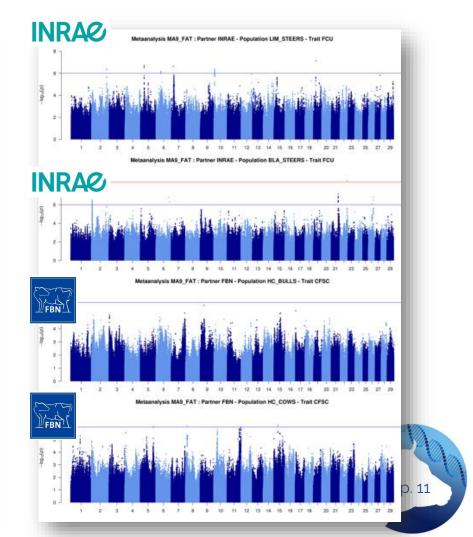
- Confirm QTL detected in within-breed GWAS with generally more significant effects
- Detect QTL not found in within-breed GWAS

Ex: carcass MA9

Within-breed GWAS

=> 3 QTL on chrom. 2, 5, and 6







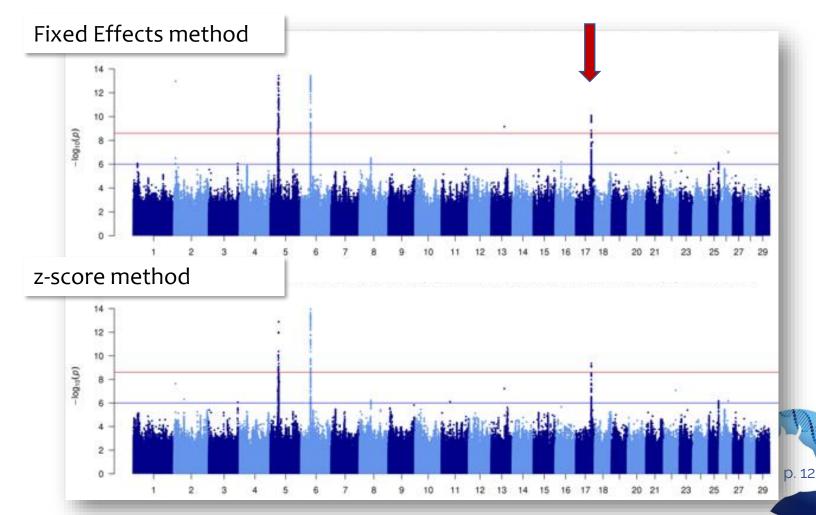
MA vs within-breed GWAS

- Confirm QTL detected in within-breed GWAS with generally more significant effects
- Detect QTL not found in within-breed GWAS

Ex: Carcass MA9

Meta-analyses

=> 3 QTL on chrom. 2, 5, and 6 + 1 new QTL on chrom. 17





MA vs within-breed GWAS

Functional annotation of TOP1 variants (variants with the most significant effects)

Functional annotation	Within-breed GWAS (%)	Fixed effects MA (%)
downstream_gene_variant	1.9	13.9
frameshift_variant	1.9	2.8
intergenic_region	37.7	27.8
intron_variant	39.6	36.1
missense_variant	1.9	2.8
stop_gained	11.3	11.1
synonymous_variant	1.9	0.0
upstream_gene_variant	3.8	5.6
% TOP1 variants in genes	62.3	72.2

Do the MA help to target causal variants?

=> Various situations depending on the QTL





MA vs within-breed GWAS

• In some cases, MA appear to better target causal variants

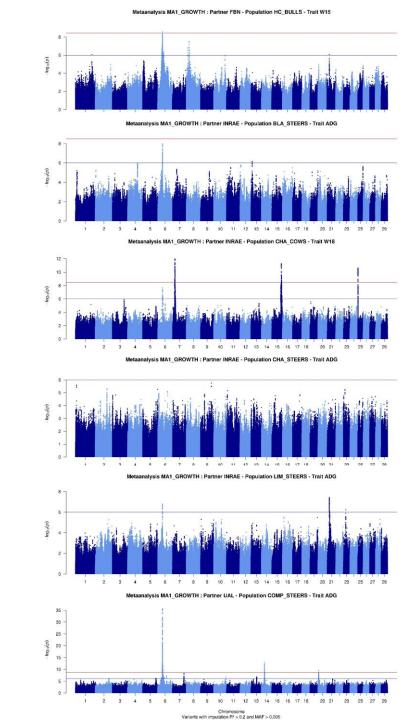
Ex: Growth MA1

MA (Fixed Effects)

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 20 21 23 25 27 29

QTL found in 10 / 16 MA

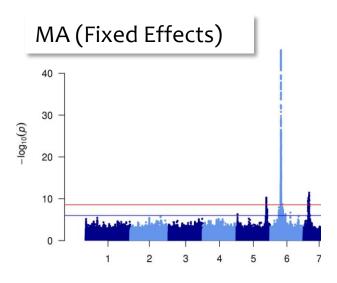




MA vs within-breed GWAS

• In some cases, MA appear to better target causal variants

Ex: Growth MA1



QTL found in 10 / 16 MA

In all cases, the TOP1 variant is in (or near) LCORL (≠ within-breed GWAS)

MA	Trait type	-Log10(p) max	Annotation	GENE
MA1	Growth	45.5	intergenic_region	LCORL-SLIT2
MA ₂	Growth	22.0	intron_variant	LCORL
MA5	Morphology	37.1	intron_variant	LCORL
MA6	Morphology	21.1	intron_variant	LCORL
MA7	Morphology	18.0	intron_variant	LCORL
MA8	Carcass	31.4	intergenic_region	LCORL-SLIT2
MA11	Carcass	11.9	frameshift_variant	LCORL
MA12	Carcass	14.7	intron_variant	LCORL
MA13	Carcass	29.2	intron_variant	LCORL
MA16	Carcass	18.1	missense variant	LCORL

LCORL => transcription factor

Associated to different traits (growth, carcass, stature, ingestion...) in various populations (e.g. Doyle et al., 2020)





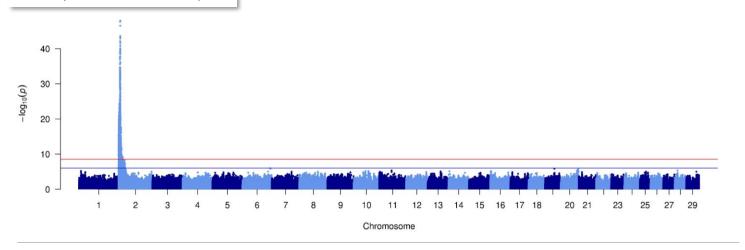
If 1 causal mutation shared between breeds => less extent LD between breeds (MA) can help to better target the causal mutation

MA vs within-breed GWAS

In other cases, MA appear to « dilute » the signal

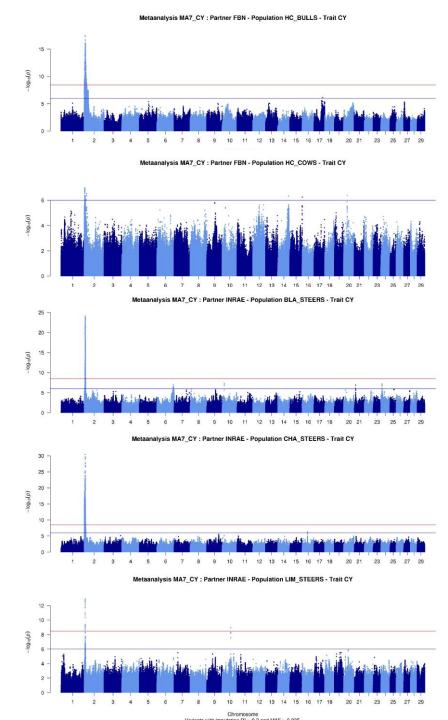
Ex: Carcass MA10 (carcass yield)

MA (Fixed Effects)



Larger confidence interval in MA than in some within-breed GWAS





MA vs within-breed GWAS

In other cases, MA appear to « dilute » the signal

Chromosome 2

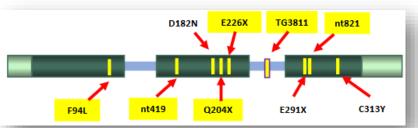
MSTN gene (myostatin)

Muscular hypertrophia



McPherron and Lee, 1997

9 known mutations in MSTN



The most frequent mutation differs depending on the breed (Renand et al., 3R 2020)

Allelic frequencies	n	+	F94L	Q204X	nt419	E226X	TG3811	nt821
Salers	1 099	99,6	0,1					0,2
Charolais	42 780	84,2	6,5	9,2				
Aubrac	775	8,1	87,0	0,2			2,0	2,6
Parthenais	1 350	0,4	0,2	0,3	4,3	7,6	0,3	86,9
Blonde d'Aquitaine	16 343	0,2	0,5	0,2	0,6		98,5	
Limousine	11 677	0,1	99,1	0,2				0,5







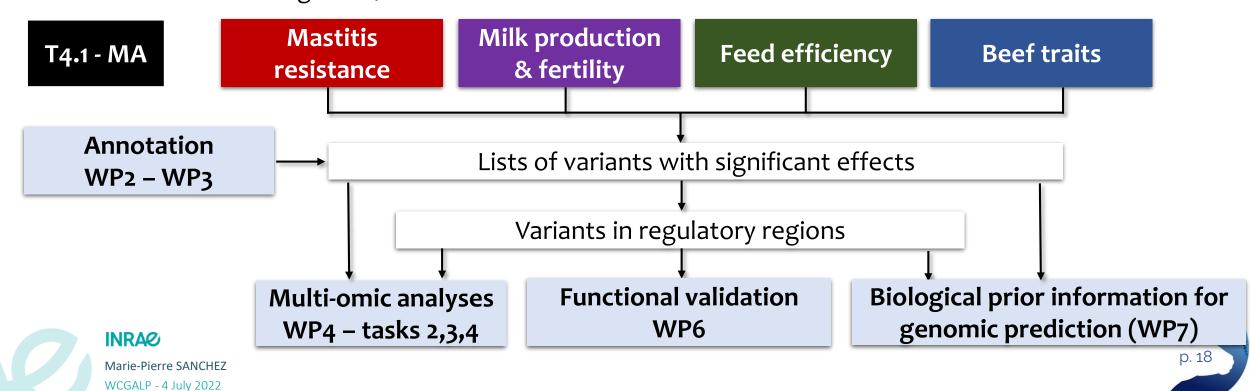
If **different causal mutations** in the different breeds => The GWAS signal can be diluted

T4.1 – « Beef » MA – discussion / conclusions

Meta-analyses conducted from within-breed GWAS results (without elementary data)

- => QTL with more significant effects (+ new QTL)
- => Appear to better target causal variants in some cases (shared mutation)

New investigations have to be conducted to identify genes and causal variants Links with other BovReg tasks/WP



KO BovReg meeting, Sept. 2019









