



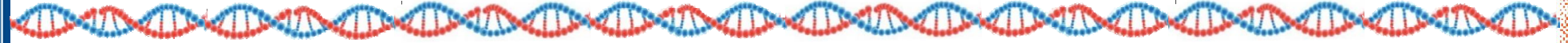
GENOMIC ANALYSES FOR PALATABILITY OF BEEF

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UF UNIVERSITY of
FLORIDA

Outline



1 Research program overview

2 Genomic analyses for palatability

2.1 Genome-wide association (GWAS)

2.2 Genomic Prediction

2.3 GWAS signals → underlying biology

Research program focus

- Genetic control of complex traits
 - Major advances in animal genomics
 - Improve animal production efficiency & enhance animal products for human health

Healthfulness Project: development of genetic tools to improve **nutritional** and **health** value of beef.



Sustainable small ruminant production through selection for **resistance to internal parasites**



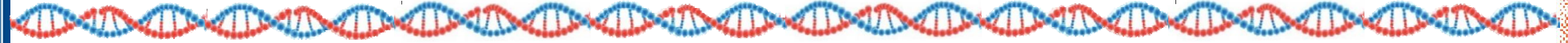
Genomic tools to improve **meat quality** traits in Angus-Brahman cattle.

Genomics of **resilience** in sheep to climatic stressors



Mapping QTL for **breeding out of season** and **milk production** in sheep

“Healthfulness Project”



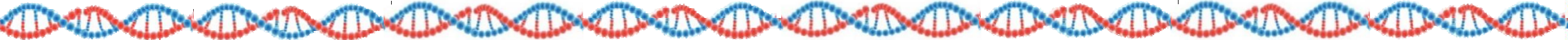
- **Project Goal:**

- Assess the **natural variation** in healthfulness traits
- Identify **genetic mechanism** controlling these traits
- Develop tools to **select for**:
 - More nutritious beef
 - Tasty beef
 - Improved production

Genetic evaluation for nutrient composition of beef could result in development of **gEBV** to aid selection for:

- **favorable fatty-acid** profile
- **lower** levels of **cholesterol** and **saturated** fat
- **higher** concentrations of **minerals** and **vitamins**

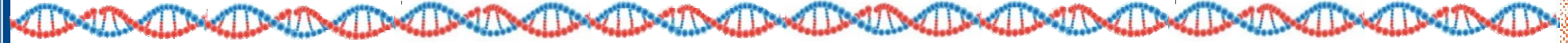
Beef Healthfulness Project

- 
- **3 Angus herds (n = 2,285): Iowa, Oklahoma, California**
 - **Harvest: October 2007 – May 2008**
 - **Growth**
 - Birth, weaning, yearling, slaughter weights
 - **Carcass**
 - Hot carcass weight, dressing %, ribeye area, back fat thickness, yield grade, quality grade, KPH
 - **Meat Quality**
 - WBSF, Sensory panels (Juiciness, Tenderness, Connective tissue & Flavor: beef, painty/fishy, livery/metallic), TBARS
 - **Nutritional value and Healthfulness**
 - Fatty acids, main FA groups (SFA, MUFA, PUFA, n3, n6), AI – Triacylglycerol, Phospholipid, Composite
 - Cholesterol, sphingolipids, creatine, creatinine, carnitine, carnosine, anserine
 - Minerals: iron, sodium, magnesium, manganese, zinc, phosphorus, potassium, calcium, copper
 - Vitamins: E, B6

All animals genotyped with the Illumina Bovine **50K SNP** Bead Chip

5 generation pedigree (n = 5,907)

What do consumers want?



- **The National Beef Quality Audit**

- Seedstock Producers, Cow-Calf Producers, Stockers/Backgrounders, Feedlot Operators, Packers, Purveyors, Retailers, & Restaurateurs
- “**Eating satisfaction**” - only quality category for which the packers, food service buyers, and retailers are willing to pay a premium

- **NBQA Strategy Workshop**

- Discuss the implications of research for the U.S. beef industry, provides the beef industry with a blueprint for the next five years
- **Top 10 Quality Challenges:**
 1. Low overall uniformity & consistency of cattle, carcasses, cuts
 2. Inappropriate carcass size & weight
 3. Inadequate **tenderness** of beef
 4. Insufficient marbling

National Beef Quality Audit

Quality Challenges

Ranked according to priority, 1991 to 2011



1991	1995	2000	2005	2011
External Fat	Overall Uniformity	Overall Uniformity	Traceability	Food Safety
Seam Fat	Overall Palatability	Carcass Weights	Overall Uniformity	Eating Satisfaction
Overall Palatability	Marbling	Tenderness	Instrument grading	How and where cattle were raised
Tenderness	Tenderness	Marbling	Market Signals	Weight and Size
Overall Cutability	External and Seam Fat	Reduced quality due to implants	Segmentation	Lean, Fat and Bone
Marbling	Cut weights	External Fat	Carcass Weights	Cattle Genetics

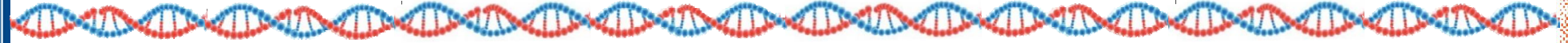
National Beef Quality Audit – 2011



Specific recommendations to reduce barriers and improve beef profitability

- Increase funding to improve **eating satisfaction**
- Use genetics to optimize cutability and **palatability**
- **Fit Cattle with Customer Expectations**
 - Assist producers with use of selection and management techniques to produce cattle that fit customer expectations and other value-determining attributes

Selection/ Prediction challenges



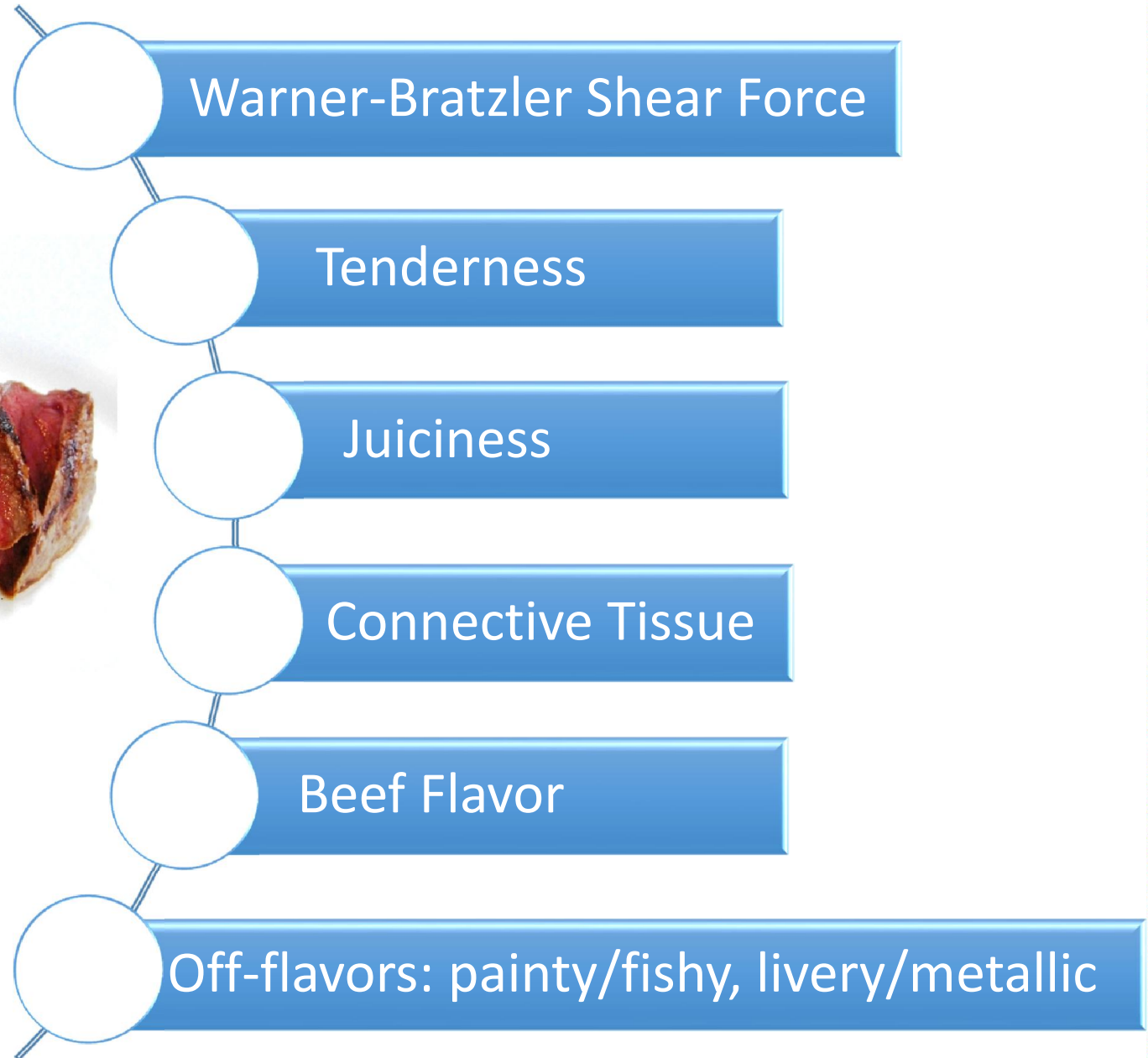
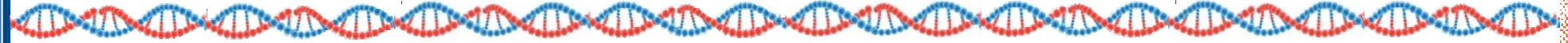
- Trait measured after slaughter, expensive to measure routinely, low consistency
- Currently USDA grading system (marbling and maturity) is used to predict palatability of beef
 - Changes in consumer preference
 - Limitation in the ability to predict eating quality
 - Limited consumer understanding of the system

Higher quality grade = more tender and palatable meat

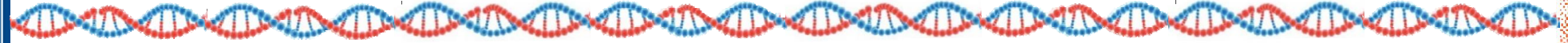
- Programs to **improve** eating quality
- Ability to better **predict** the eating quality level for market purposes

Increase consumer's confidence that quality expectations are met

Palatability – eating satisfaction



Sample Collection and Preparation

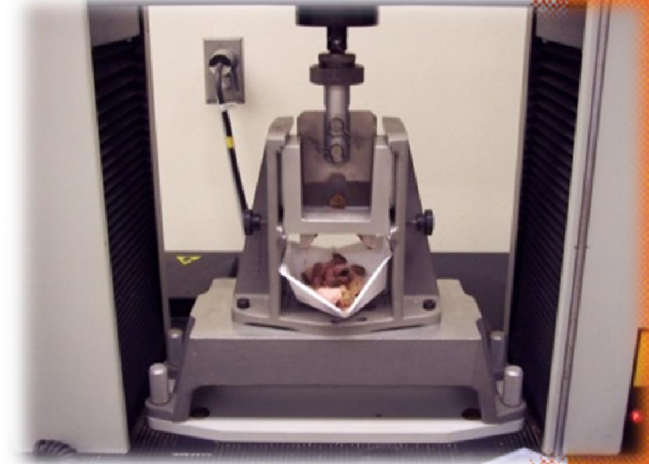


- Rib sections:
 - Transported to ISU Meat Laboratory, or shipped from California to OSU
 - Fabricated into steaks
 - Frozen at 14d postmortem



Warner-Bratzler Shear Force (WBSF)

- Broiled on impingement oven at 200°C to internal temperature of 68°C
- Cooled at 4°C for 18-24 h
- **Six** 1.27 cm cores removed and sheared
- Average peak load (**kg**) was analyzed



Sensory Panels

- Cooked similarly to steaks for WBS
- Sessions conducted once or twice per day
- 12 samples served randomly to panelists
- Served to eight member trained panel



Each steak

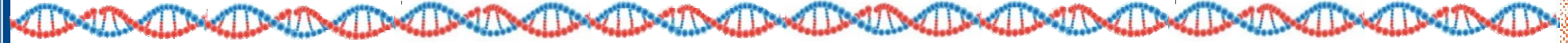


8 trained panelists



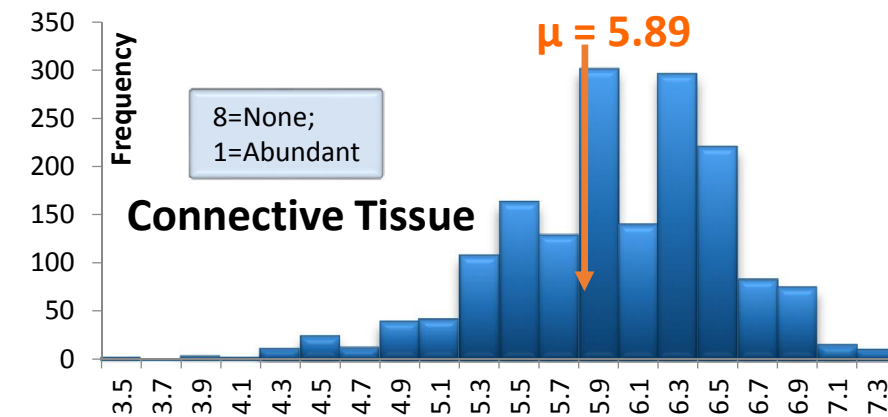
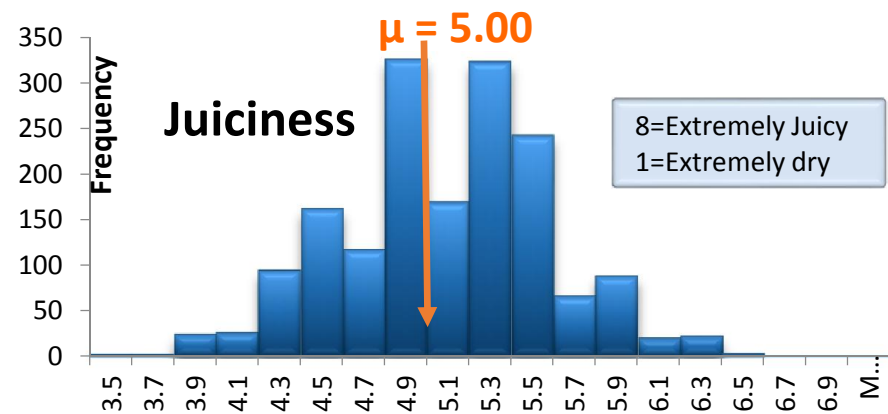
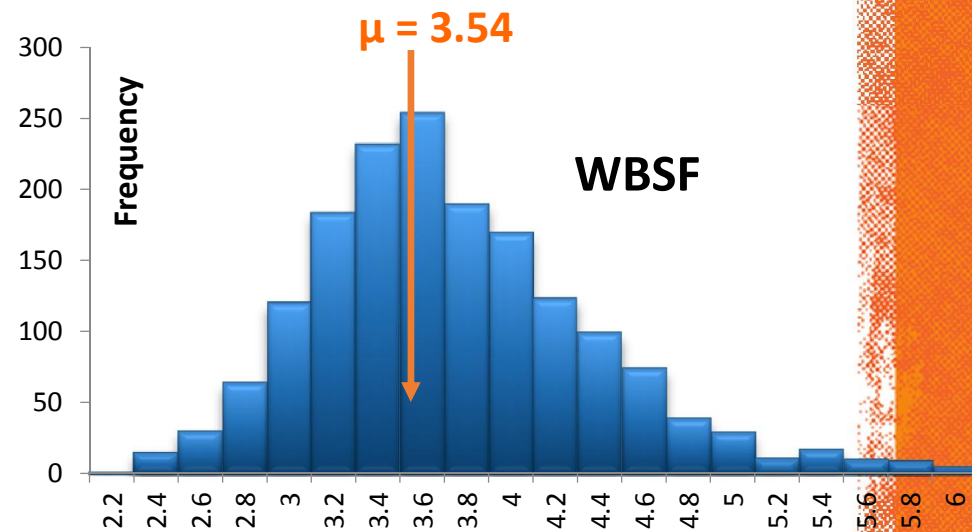
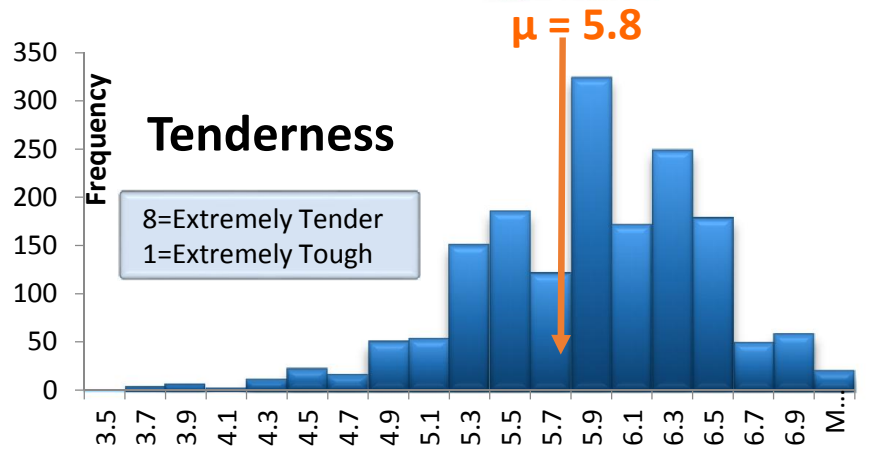
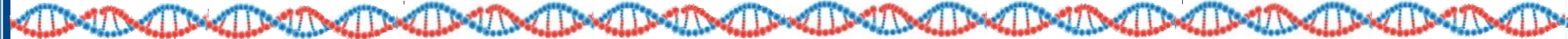
Average
score of
all
panelists

Sensory Panel Evaluation



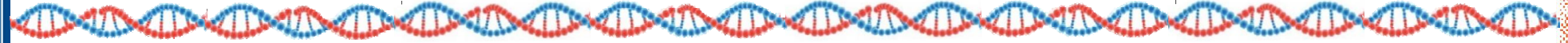
- Evaluated juiciness, tenderness, flavor
 - **Juiciness** – 8 point scale (1 = extremely dry and 8 = extremely juicy)
 - **Tenderness** – 8 point scale (1 = extremely tough and 8 = extremely tender)
 - **Connective Tissue** – 8 point scale (1 = abundant and 8 = none)
 - **Beef**, Painty/Fishy, and Livery/Metallic Flavors – 3 point scale (1 = not detectable, 2 = slightly detectable, and 3 = strong)

Palatability statistics



Trait	N	Mean ± SD
WBSF	2,076	3.54 ± 0.77
Tenderness	1,591	5.80 ± 0.59
Juiciness	1,591	5.00 ± 0.49
Connective Tissue	1,591	5.89 ± 0.59
Beef Flavor	1,591	2.50 ± 0.23
Painty	1,591	1.13 ± 0.17
Livery	1,591	1.10 ± 0.12

Palatability of Beef - goals

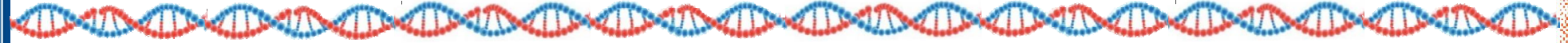


1 Genome-wide association (GWAS)

2 Genomic Prediction

3 GWAS signals → underlying biology

Genome-wide association



- Identify **SNPs** and chromosomal regions associated with palatability traits
 - WBSF
 - Tenderness (sensory panel)
 - Juiciness (sensory panel)
 - Connective tissue (sensory panel)
 - Flavor (sensory panel)
- **SVS (SNP & Variation Suite) v8.3.4 (Golden Helix)**
 - Mixed Model GWAS using single locus models (**EMMAX**) and multi-locus models (**MLMM**)

GWAS in SVS - initial steps



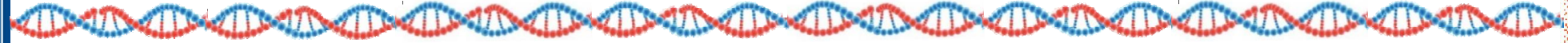
QualityPhenotypes + Genotypes 50K - Mapped WBSF [1519]

File Edit Select DNA-Seq Genotype Numeric RNA-Seq GenomeBrowse Plot Scripts Help

WBSAVG (Quantitative), 2,110 x 53,465

Unsort		G 2752	G 2753	G 2754	G 2755	G 2756	G 2757	G 2758
Map	HEALTHID	Hapmap42876-BTA-23541	ARS-BFGL-NGS-115015	BTB-00011225	BTB-00011494	ARS-BFGL-NGS-46004	BTA-59258-no-rs	Hapmap50048
Chromosome		1	1	1	1	1	1	1
Position		27145992	27181523	27249385	27324745	27357510	27434665	27465
dbSNP Strand		A/G	T/C	G/C	A/G	G/C	T/C	T/
Strand Versus dbSNP		same	reverse	same	same	same	reverse	reve
Observed		A/G	A/G	G/C	A/G	G/C	A/G	A/
GenTrain Score		0.9364	0.8837	0.7879	0.8868	0.9532	0.8531	0.88
Strand		TOP	TOP	BOT	TOP	BOT	TOP	TO
2	20081015111597		B_B	B_B	A_A	B_B	B_B	A_A
3	20081015111589		A_B	B_B	A_A	B_B	B_B	B_B
4	20081015111547		B_B	B_B	A_A	B_B	B_B	A_B
5	20081015111553		B_B	B_B	A_A	B_B	B_B	A_A
6	20081015111577		A_A	B_B	A_A	B_B	B_B	B_B
7	20081015111605		B_B	B_B	A_A	B_B	B_B	B_B
8	20081015111593		A_B	B_B	A_A	B_B	B_B	A_B
9	20081015111581		B_B	A_A	B_B	A_B	B_B	A_B
10	20081015111601		A_B	B_B	A_A	B_B	B_B	A_B
11	20081015111539		A_B	A_B	A_B	A_B	B_B	B_B
12	20080502C40313		B_B	B_B	A_A	B_B	B_B	B_B

GWAS in SVS - genotype filtering



QualityPhenotypes + Genotypes 50K - Mapped - Sheet 5 [3057]

File Edit Select DNA-Seq Genotype Numeric RNA-Seq GenomeBrowse Plot Scripts Help

All: 2,110 x 53,465
Active: 2,076 x 53,465

Unsort	G	4333	G	4334	G	4335	G	4336	G	4337	G	4338	G	4339	G	4340	G	4341
Map	HEALTHID	BTB-00045786	BTB-00045751	Hapmap38948-BTA-28757	BTB-00046247	BTB-02007023	Hapmap42893-BTA-27908	BTB-01831301	BTB-01579733	BTB-01579794								
Chromosome		1	1						1	1								
Position		103926075	103950225	103950225					104280525	104317463								
dbSNP Strand		T/C	T/C						A/C	T/C								
Strand Versus dbSNP		same	same						same	reverse								
Observed		T/C	T/C						A/C	A/G								
GenTrain Score		0.8871	0.9151						0.9366	0.8853								
Strand		BOT	BOT						TOP	TOP								

Genotype Filtering by Marker

(No variable is set as dependent.)

Classify alleles by allele frequency Classify alleles by reference/alternate
(Marker map "Reference" field required)

Filter Genotype Columns

General Statistics Filtering

- Drop if call rate
- Drop if number of alleles
- Drop if Minor Allele Frequency (MAF)
- Drop if carrier count

Hardy Weinberg Equilibrium (HWE) Filtering

Perform HWE filtering based on:

- Drop if Hardy Weinberg Equilibrium (HWE) P-Value
- Drop if Fisher's exact test for HWE P-Value
- Drop if signed HWE R (positive if more homozygous)

Actions

- Inactivate genotype columns that meet above criteria for filtering
- Output spreadsheet with marker statistics and 'Drop?' columns

Additional Output

- Output -log10(Value)

Help Restore Options Save Options Run Cancel

GWAS for WBSF – SVS

- Mixed Model GWAS using a single locus (**EMMAX**) and multi-locus models (**MLMM**)
 - Genomic relationship matrix
 - Contemporary groups

The screenshot shows the 'Mixed Linear Model Analysis' dialog box. The 'MLM Parameters' tab is active. Under 'Regression Model(s) To Use', the following options are checked: 'Mixed Model GWAS', 'Single-locus mixed model GWAS (EMMAX)', and 'Use Pre-Computed Kinship Matrix (Cov. Matrix of Random Effects)'. The 'Number of steps to use' is set to 10. A 'GBLUP Genomic Relationship Matrix' is selected, with a 'Select Sheet' button. The 'Correct for Additional Covariates' section is checked, and 'NUMCONTG' is listed in the box. The 'Genetic Model and Imputation' section shows 'Additive' selected for the genetic model and 'Numerically as average value' for imputation. The 'Correct For Hemizygous Males' section is unchecked, with 'X' entered for the chromosome. The 'OK', 'Cancel', and 'Help' buttons are at the bottom.

Mixed Linear Model Analysis

MLM Parameters Additional Outputs

Regression Model(s) To Use

Linear regression (fixed effects only)

Mixed Model GWAS

Single-locus mixed model GWAS (EMMAX)

Multi-locus mixed model GWAS (MLMM)

Number of steps to use: 10

Use Pre-Computed Kinship Matrix (Cov. Matrix of Random Effects)

GBLUP Genomic Relationship Matrix Select Sheet

NOTE: If no pre-computed kinship matrix spreadsheet is selected, an IBS spreadsheet will be computed from the genotype data and used for this analysis.

Correct for Additional Covariates

NUMCONTG

Add Columns

Remove Selected

Clear List

Genetic Model and Imputation

Genetic model used for recoding the original spreadsheet:

Additive Dominant Recessive

Impute missing data as:

Homozygous major allele Numerically as average value

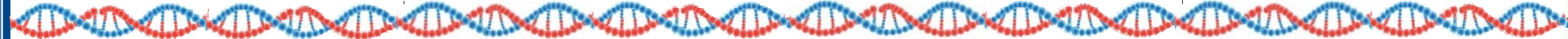
Correct For Hemizygous Males

Choose Sex Column: Select Column

Chromosome that is hemizygous for males: X

OK Cancel Help

GWAS for WBSF - SVS - output



P-Values from Single-Locus Mixed Model WBSF [1484]

Unsort		R	1	R	2	R	3	R	4	R	5	R	6	R	7
Map	Marker		P-Value		$-\log_{10}(\text{P-Value})$		Regression Beta		Beta Standard Error		Expected P		$-\log_{10}(\text{Expected P})$		FDR
1	Hapmap43437-BTA-101873		0.383821648310084		0.415870533264946		-0.0285368845077287		0.032761059113204		0.382716411489803		0.417122914136511		1
2	ARS-BFGL-NGS-16466		0.515419782177384		0.287838916934501		0.018682979636141		0.0287194749870717		0.51214289904382		0.290608844478641		1
3	ARS-BFGL-NGS-19289		0.0636798254424759		1.19599813574579		0.384513604774792		0.207239366901906		0.0696212432295027		1.15725822576744		0.915909379250977
4	Hapmap34944-BES1_Contig627_1906		0.785545607914954		0.104828595231531		0.0114773283735648		0.042175185831537		0.781554916798655		0.107040500402728		1
5	BTA-07251-no-rs		0.917230288158053		0.037521612726741		-0.00381395499449193		0.0366952686512595		0.915009483584599		0.0385744046798728		1
6	ARS-BFGL-NGS-98142		0.823512154836004		0.0843299863676352		0.00729278064823647		0.0326945134104316		0.820408282982343		0.0859699635701682		1
7	Hapmap53946-rs29015852		0.149964121774113		0.824012631472996		-0.0395628967966938		0.027470573428266		0.155677440800923		0.807774316433205		0.964690076491648
8	ARS-BFGL-NGS-114208		0.389176081721017		0.409853858783378		-0.0234085431879875		0.0271783613648939		0.38754619581158		0.41167652182794		1
9	ARS-BFGL-NGS-66449		0.575477852905042		0.23997138540727		-0.0162121902459932		0.0289456335028307		0.572270780782542		0.242398427897064		1
10	ARS-BFGL-BAC-32770		0.791571825152898		0.101509672270577		0.0222013840837164		0.0839993730972714		0.787108191079564		0.103965568073668		1
11	ARS-BFGL-NGS-65067		0.97693805630211		0.0101329722323896		0.00078799290569142		0.0272554627035422		0.977464265462154		0.00989911075693163		1
12	ARS-BFGL-BAC-31497		0.937666701496802		0.0279515064223586		0.010124352585938		0.129447012170112		0.935228094874954		0.0290824551547167		1
13	ARS-BFGL-BAC-32722		0.65532008905181		0.18354651840031		0.0415682827653159		0.0931088410705422		0.652754150290374		0.185250358199039		1
14	ARS-BFGL-BAC-34682		0.95907832417845		0.0181459205170035		0.00498446835454126		0.0971320462825188		0.960550243444594		0.0174799135924503		0.999960793701189
15	ARS-BFGL-NGS-3964		0.90695956655662		0.0424120738763232		0.00404998772846728		0.0346482366110295		0.903531413151874		0.0440567436881523		1

P-Values from Single-Locus Mixed Model WBSF [1484]

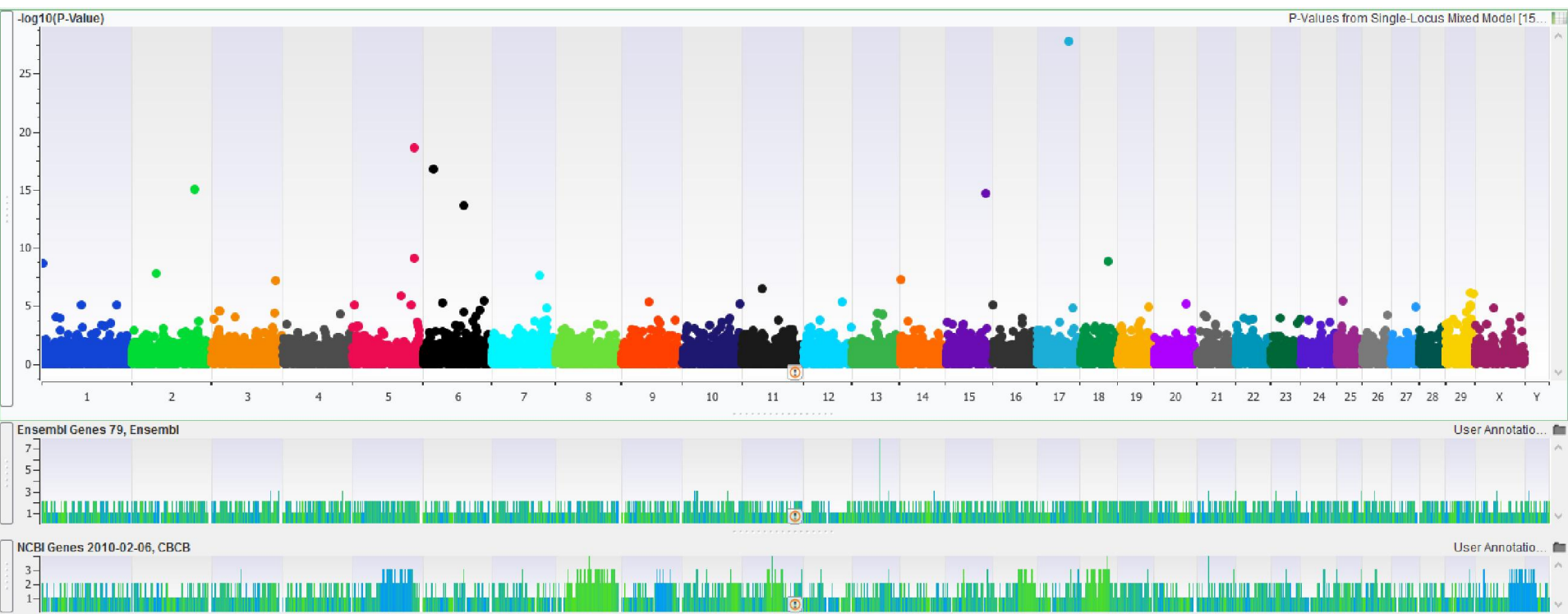
Unsort		R	8	R	9	R	10	R	11	R	12
Map	Marker		Proportion of Variance Explained		Mahal. RSS		Call Rate		Minor Allele D Frequency		Major Allele d Frequency
1	Hapmap43437-BTA-101873		0.000365880292415577		438.1050549380863		0.988439306358382		0.262426900584795		0.737573099415205
2	ARS-BFGL-NGS-16466		0.000204103915236731		438.175955928754		1		0.474470134874759		0.525529865125241
3	ARS-BFGL-NGS-19289		0.00165790251263243		437.538806293878		0.979287090558767		0.183472700442696		0.816527299557305
4	Hapmap34944-BES1_Contig627_1906		3.57234001253648e-005		438.249751283844		0.986994219653179		0.197657393850659		0.802342606149341
5	BTA-07251-no-rs		5.21110014883863e-006		438.26312376943		1		0.495905587668593		0.504094412331407
6	ARS-BFGL-NGS-98142		2.40008609111442e-005		438.254888867271		0.953275529865125		0.326932794340576		0.673067205659424
7	Hapmap53946-rs29015852		0.00099955591911649		437.827336832036		0.988439306358382		0.139863547758285		0.860136452241715
8	ARS-BFGL-NGS-114208		0.000357723563513046		438.108629750985		0.991811175337187		0.416221466731423		0.58377853268577
9	ARS-BFGL-NGS-66449		0.000151304640196304		438.199096024552		0.977842003853565		0.365270935960591		0.634729064039409
10	ARS-BFGL-BAC-32770		3.3697220037987e-005		438.250639288486		0.95616570327553		0.449622166246851		0.550377833753149
11	ARS-BFGL-NGS-65067		4.03216242395033e-007		438.265230898631		0.96242774566474		0.470720720720721		0.529279279279279
12	ARS-BFGL-BAC-31497		2.95086864521288e-006		438.264114350712		0.981695568400771		0.122914622178606		0.877085377821394
13	ARS-BFGL-BAC-32722		9.61393692351376e-005		438.223273054516		0.971579961464355		0.116261774913237		0.883738225086762
14	ARS-BFGL-BAC-34682		1.27031889107787e-006		438.264850877535		0.959537572254335		0.151104417670683		0.848895582329317
15	ARS-BFGL-NGS-3964		6.59087351229104e-006		438.262519062495		1		0.467003853564547		0.532996146435453



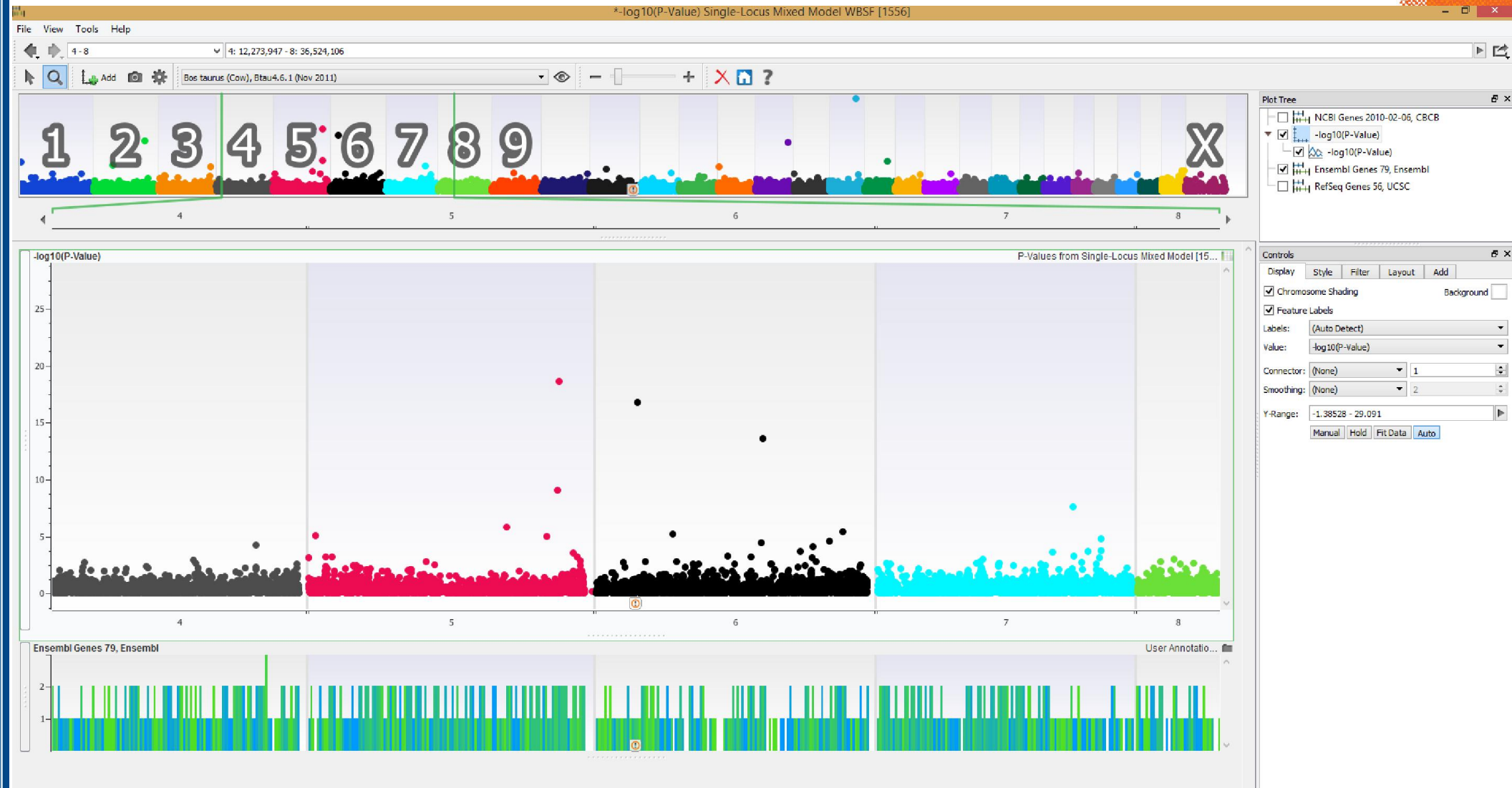
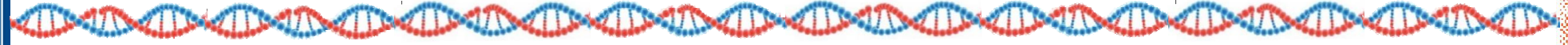
GWAS for WBSF – SVS

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 - Genomic relationship matrix
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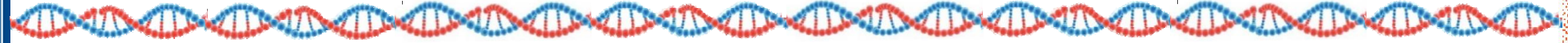
Samples scanned: 2,076
Markers scanned: 51,218
Markers analyzed: 51,141
Pseudo-heritability: 0.37



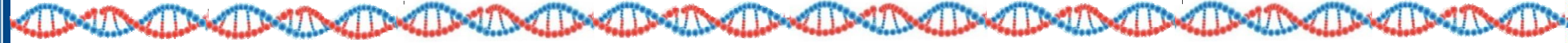
Exploring data - GenomeBrowser



Exploring data - GenomeBrowser



Exploring data - GenomeBrowser

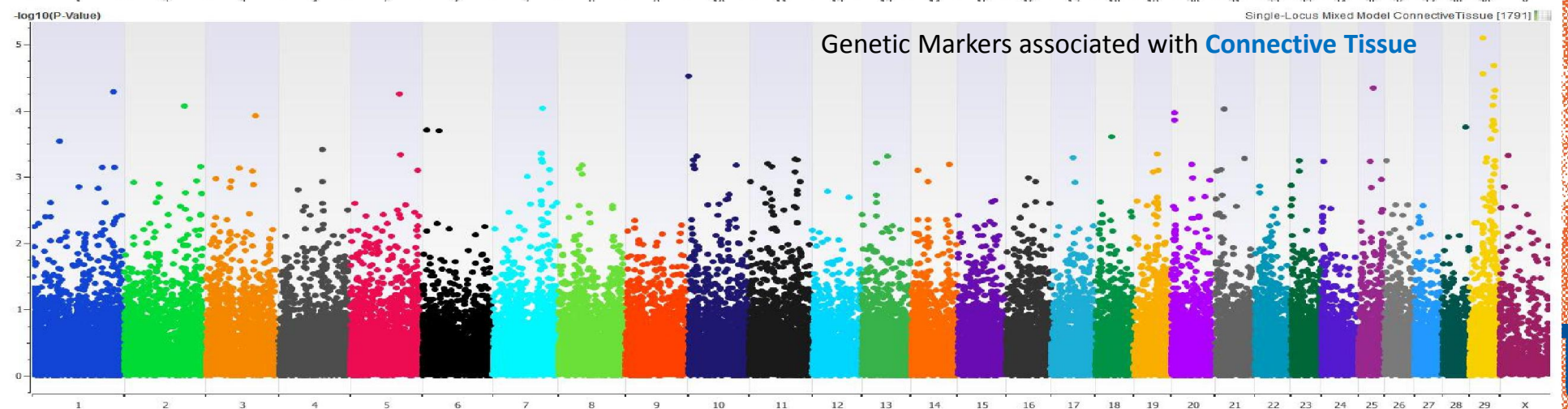
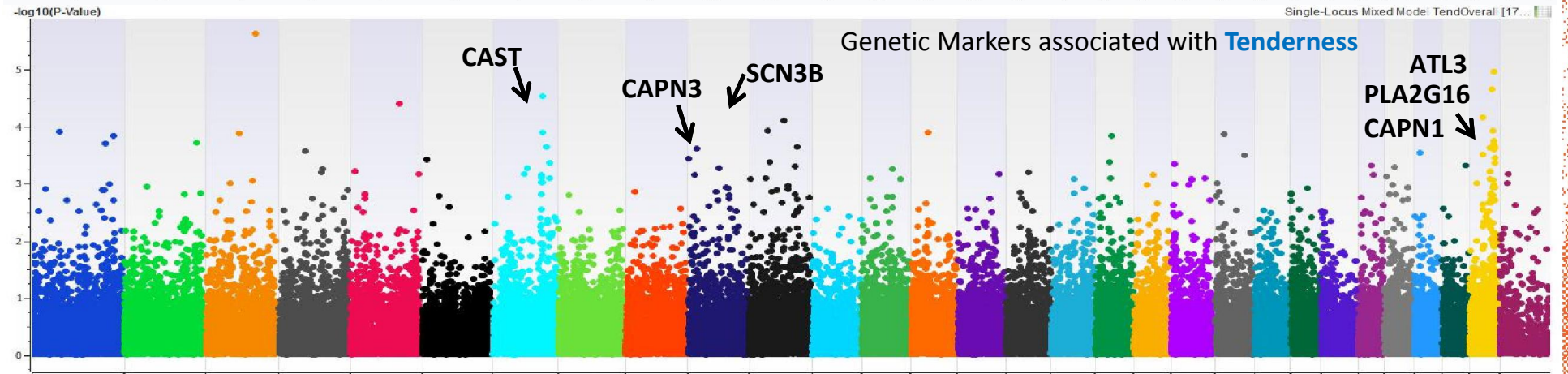
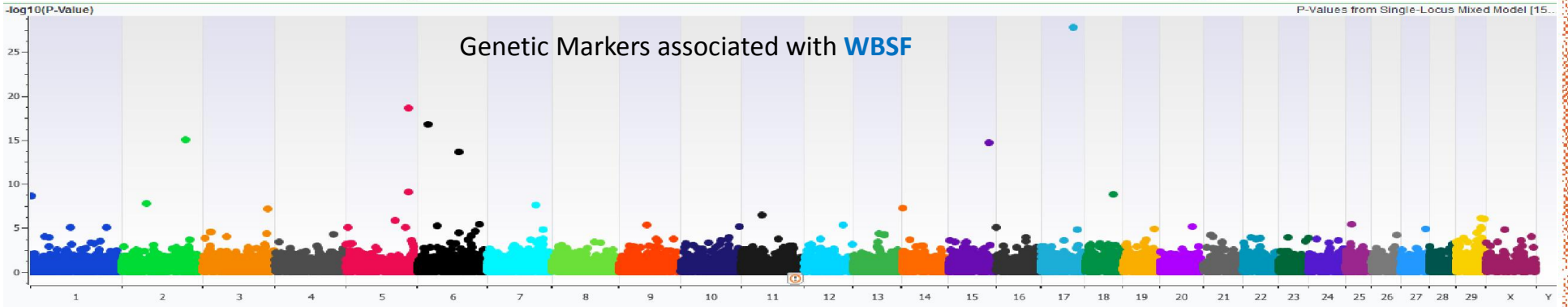
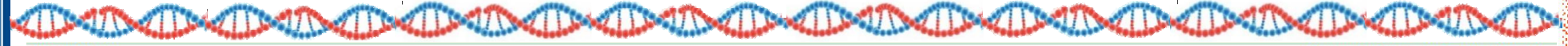


Feature List																					
Ensembl Genes 79, Ensembl																					
Chr	Start	Stop	Gene Name	Transcript Name	CDS Start	CDS Stop	Exon Starts	Exon Stops	Strand	Exon ID	n num	n Vers	Gene Biotype	Gene ID	ene Versic	Protein ID	stein Versi	ernate transcript	cript Ver		
1	5	102715224	102729724	ENSBTAG000000...	ENSBTAT000000...	102715223	102729582	102715223,1027...	102715467,1027...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	2	?	2	
2	5	102722829	102726245	ENSBTAG000000...	ENSBTAT000000...	102722828	102726245	102722828,1027...	102723177,1027...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	1	?	1	
3	5	102726402	102729724	ENSBTAG000000...	ENSBTAT000000...	102726401	102729582	102726401,1027...	102726486,1027...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	4	?	4	
4	5	102894599	102923181	ENSBTAG000000...	ENSBTAT000000...	102894598	102923181	102894598,1029...	102894917,1029...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	1	ENSBTAP000000...	2	?	2	
5	5	102940224	102966384	ENSBTAG000000...	ENSBTAT000000...	102940223	102966384	102940223,1029...	102940296,1029...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	2	?	2	
6	5	102940224	102966384	ENSBTAG000000...	ENSBTAT000000...	102940223	102966384	102940223,1029...	102940296,1029...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	2	?	2	
7	5	102968842	102991326	ENSBTAG000000...	ENSBTAT000000...	102968841	102991326	102968841,1029...	102968957,1029...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	3	ENSBTAP000000...	2	?	2	
8	5	103032491	103060434	ENSBTAG000000...	ENSBTAT000000...	103032490	103060434	103032490,1030...	103032618,1030...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	1	ENSBTAP000000...	2	?	2	
9	5	103081967	103082596	ENSBTAG000000...	ENSBTAT000000...	?	?	103081966	103082596	+	ENSBTAE000000...	1	1	processed_pseu...	ENSBTAG000000...	1	?	?	?	1	
10	5	103110444	103113609	ENSBTAG000000...	ENSBTAT000000...	103110443	103113609	103110443,1031...	103110607,1031...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	3	ENSBTAP000000...	1	?	1	
11	5	103132709	103149919	ENSBTAG000000...	ENSBTAT000000...	103132708	103149919	103132708,1031...	103132854,1031...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	1	?	1	
12	5	103132709	103261137	ENSBTAG000000...	ENSBTAT000000...	103132708	103261137	103132708,1031...	103132854,1031...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	4	?	4	
13	5	103170324	103207769	ENSBTAG000000...	ENSBTAT000000...	103170323	103207769	103170323,1031...	103170619,1031...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	2	?	2	
14	5	103276708	103332323	ENSBTAG000000...	ENSBTAT000000...	103276707	103332323	103276707,1032...	103276850,1032...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	2	?	2	
15	5	103334470	103335099	ENSBTAG000000...	ENSBTAT000000...	?	?	103334469	103335099	+	ENSBTAE000000...	1	1	processed_pseu...	ENSBTAG000000...	1	?	?	?	1	
16	5	103375408	103417370	CD163L1	ENSBTAT000000...	103375451	103417370	103375407,1033...	103375597,1033...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	1	?	CD163L1-201	1
17	5	103375408	103426531	CD163L1	ENSBTAT000000...	103375451	103426531	103375407,1033...	103375597,1033...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	2	?	CD163L1-202	2
18	5	103389271	103439148	CD163L1	ENSBTAT000000...	103389270	103439095	103389270,1033...	103389613,1033...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	2	?	CD163L1-203	2
19	5	103458792	103460314	ENSBTAG000000...	ENSBTAT000000...	103458791	103460314	103458791,1034...	103458943,1034...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	?	4
20	5	103530546	103546956	PEX5	ENSBTAT000000...	103531697	103546254	103530545,1035...	103531899,1035...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	5	ENSBTAP000000...	2	?	PEX5-201	2
21	5	103530565	103547919	PEX5	ENSBTAT000000...	103531697	103546254	103530564,1035...	103531899,1035...	-	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	5	ENSBTAP000000...	4	?	PEX5-202	4
22	5	103581388	103611327	CLSTN3	ENSBTAT000000...	103582113	103611068	103581387,1035...	103582254,1035...	-	ENSBTAE000000...	1	2	protein_coding	ENSBTAG000000...	5	ENSBTAP000000...	5	?	CLSTN3-201	5
23	5	103581388	103611327	CLSTN3	ENSBTAT000000...	103582358	103611068	103581387,1035...	103582254,1035...	-	ENSBTAE000000...	1	2	protein_coding	ENSBTAG000000...	5	ENSBTAP000000...	4	?	CLSTN3-202	4
24	5	103612392	103617657	RBP5	ENSBTAT000000...	103613069	103617361	103612391,1036...	103612559,1036...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	RBP5-201	4
25	5	103633960	103644323	C1RL	ENSBTAT000000...	103633959	103644323	103633959,1036...	103633974,1036...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	C1RL-201	4
26	5	103634595	103644323	C1RL	ENSBTAT000000...	103634594	103644323	103634594,1036...	103634677,1036...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	1	?	C1RL-202	1
27	5	103664843	103687826	C1R	ENSBTAT000000...	103681239	103687826	103664842,1036...	103664937,1036...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	2	ENSBTAP000000...	2	?	C1R-201	2
28	5	103737161	103747960	C1R	ENSBTAT000000...	103737216	103745018	103737160,1037...	103737218,1037...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	C1R-201	4
29	5	103737177	103747996	C1R	ENSBTAT000000...	103737216	103747851	103737176,1037...	103737218,1037...	+	ENSBTAE000000...	1	1	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	C1R-202	4
30	5	103768066	103779250	C15	ENSBTAT000000...	103768508	103778083	103768065,1037...	103769308,1037...	-	ENSBTAE000000...	1	2	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	C15-201	4
31	5	103841289	103847254	LPCAT3	ENSBTAT000000...	103841288	103846610	103841288,1038...	103841397,1038...	+	ENSBTAE000000...	1	2	protein_coding	ENSBTAG000000...	5	ENSBTAP000000...	2	?	LPCAT3-201	2
32	5	103847447	103853139	ENSBTAG000000...	ENSBTAT000000...	103853139	103853006	103847446,1038...	103847709,1038...	-	ENSBTAE000000...	1	3	protein_coding	ENSBTAG000000...	4	ENSBTAP000000...	4	?	?	4
33	5	103853322	103857548	PHB2	ENSBTAT000000...	103853381	103857548	103853321,1038...	103853508,1038...	+	ENSBTAE000000...	1	2	protein_coding	ENSBTAG000000...	3	ENSBTAP000000...	3	?	PHB2-201	3
34	5	103856022	103856266	snoU89	ENSBTAT000000...	?	?	103856021	103856266	+	ENSBTAE000000...	1	1	snoRNA	ENSBTAG000000...	1	?	?	?	snoU89-201	1

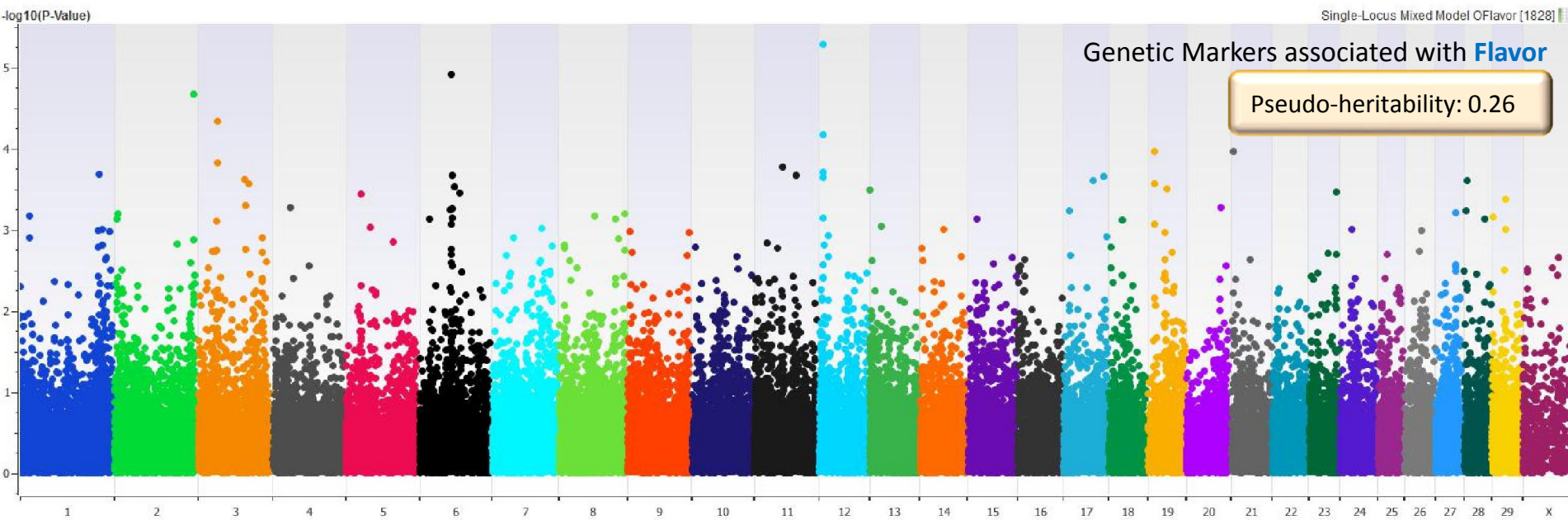
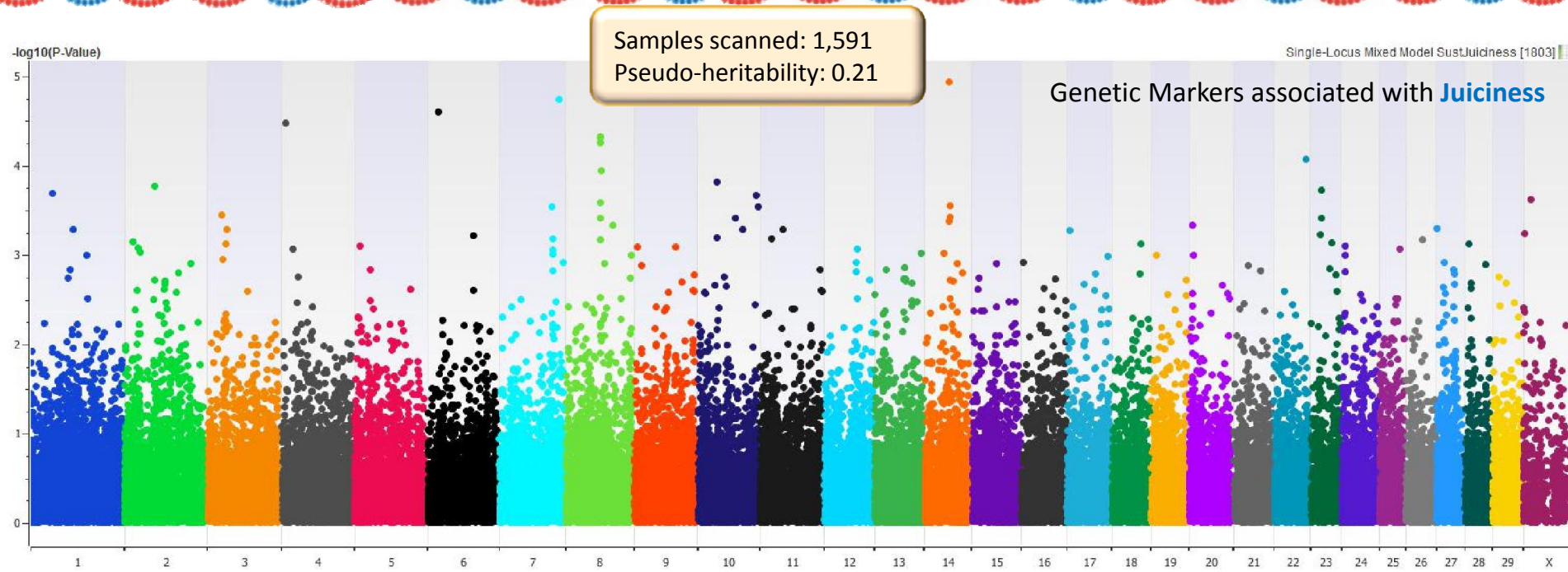
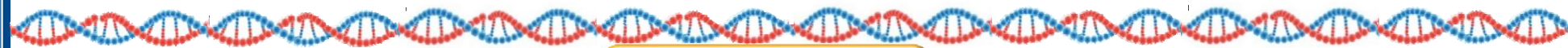
351 features read (complete)

(Chr5: 102,700,496 - 119,627,930)

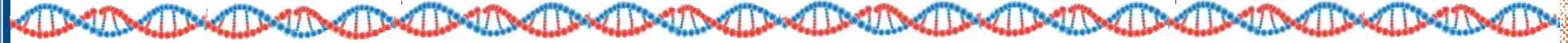
GWAS on other "tenderness" traits



GWAS on other palatability traits



Palatability of Beef - goals

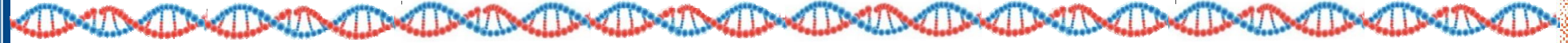


1 Genome-wide association (GWAS)

2 Genomic Prediction

3 GWAS signals underlying biology

Genomic Prediction - WBSF



Compute Genomic BLUP (GBLUP)

Computations
Compute GBLUP (Genomic Best Linear Unbiased Predictors) of additive genetic merits by sample and of allele substitution effects (ASE) by marker.

Impute missing genotypic data as:
 Homozygous major allele Numerically as average value

Correct For Gender
Choose Sex Column: _____ Select Column
Chromosome that is hemizygous for males: X

Use Pre-Computed Genomic Relationship Matrix
GBLUP Genomic Relationship Matrix Select Sheet

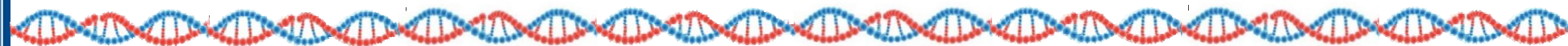
NOTE: If no pre-computed genomic relationship matrix spreadsheet is selected, a genomic relationship matrix will be computed from the genotype data and used for this analysis.

Correct for Additional Covariates
NUMCONTG
Add Columns
Remove Selected
Clear List

Missing Phenotypes
 Predict random effects for samples with missing phenotypes
 Drop samples with missing phenotypes

OK Cancel Help

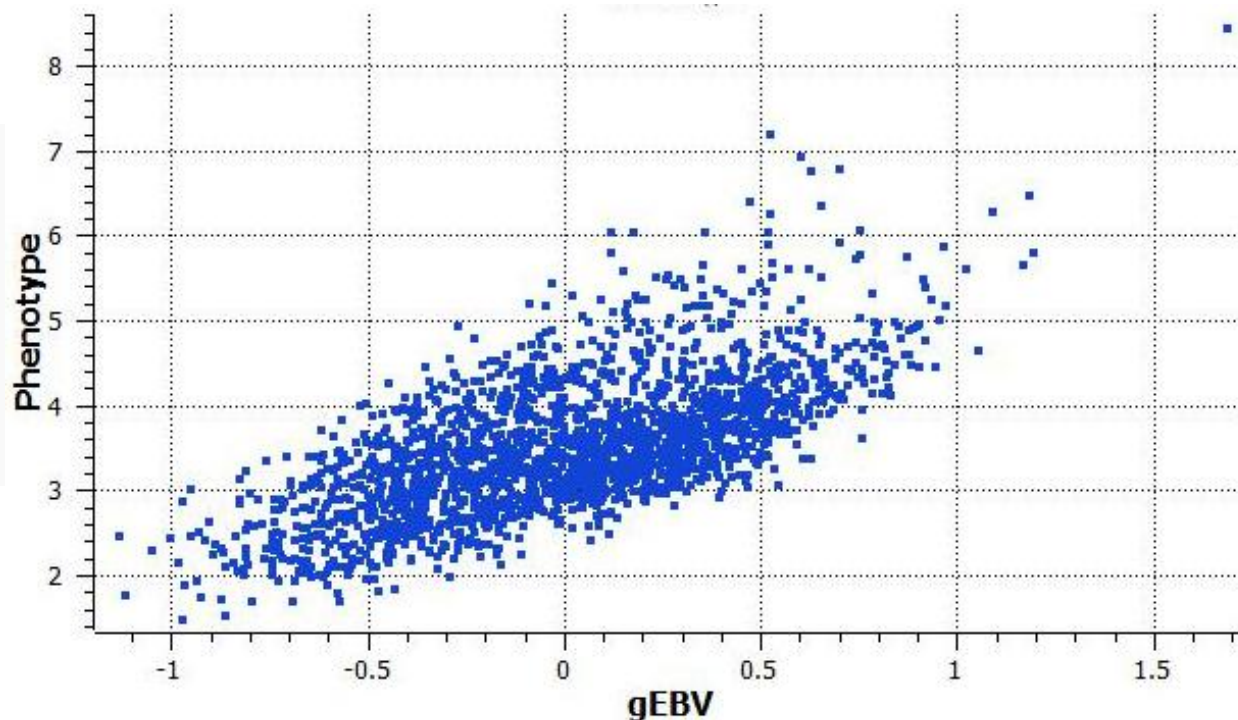
GBLUP Prediction - WBSF



Samples scanned: 2076
Markers scanned: 51218

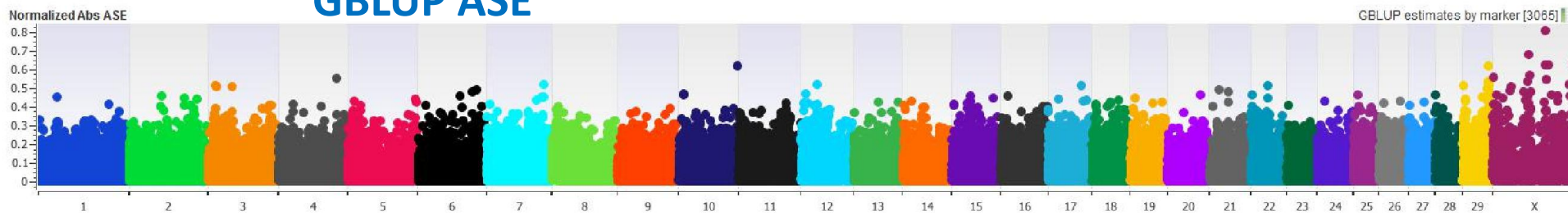
Markers analyzed: 51218
(Monomorphic and all-heterozygous markers included above: 77)

Pseudo-heritability: 0.387245
Vg: 0.223871
Ve: 0.35424
Proportion of genetic variance: 0.380596

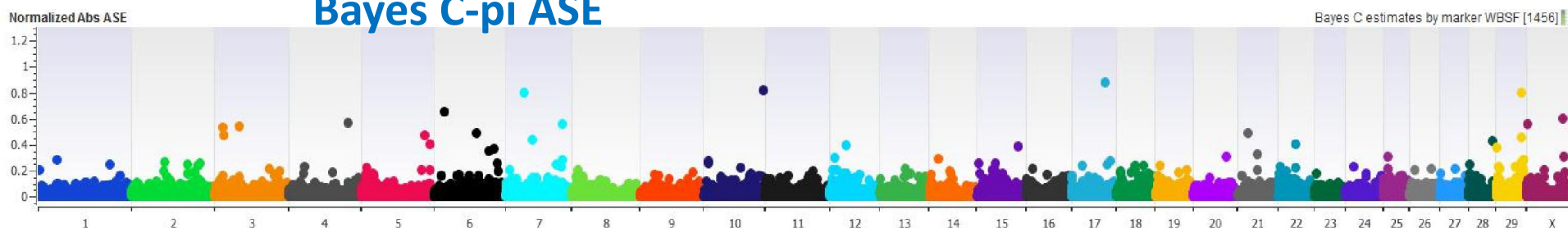


ASE vs GWAS results

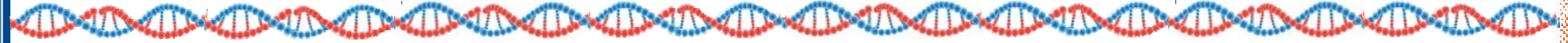
GBLUP ASE



Bayes C-pi ASE

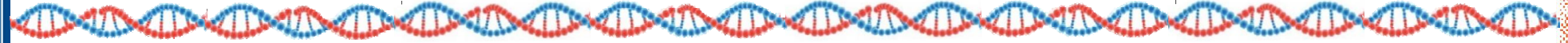


Genomic Prediction



- **Accuracy** of DGV - key to successful application of genomic selection
- Cross validation - assess performance of **prediction**
- SVS: five-fold cross-validation for evaluation of predictive performance of GBLUP/BayesC for **WBSF**

5-fold cross-validation - WBSF



K-Fold Cross Validation (for Genomic Prediction)

Computations
Perform k-fold cross validation on GBLUP and Bayes C/C-pi

Method(s)

- Genomic Best Linear Unbiased Predictors (GBLUP)
- Bayes C-pi
- Bayes C

Bayesian Options

Number of Iterations:

Burn-in:

Thinning:

Initial Pi (for Bayes C this will be the fixed value)

Correct For Gender

Choose Sex Column:

Chromosome that is hemizygous for males:

Use Pre-Computed Genomic Relationship Matrix

Correct for Additional Covariates

NUMCONTG

Impute Missing Genotypic Data As:

Homozygous major allele Numerically as average value

Stratify Folds by

kmeangrLet

K-Fold Options

Number of Folds

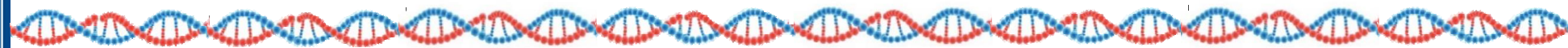
Number of Iterations

Spreadsheet Options

Delete intermediate spreadsheets with results for each fold?

NOTE: If no pre-computed genomic relationship matrix spreadsheet is selected, a genomic relationship matrix will be computed from the genotype data and used for this analysis.

Cross-validation - WBSF



GBLUP Summary Statistics - Iteration 1

Summary Statistics

Overall

Pearson's Product-Moment Correlation Coefficient: 0.4097671933
 Residual Sum of Squares: 1034.7494367615
 Total Sum of Squares: 1242.2059729345
 R-Squared: 0.1670065518
 Root Mean Square Error: 0.7059987380
 Mean Absolute Error: 0.5424411481

Fold 1

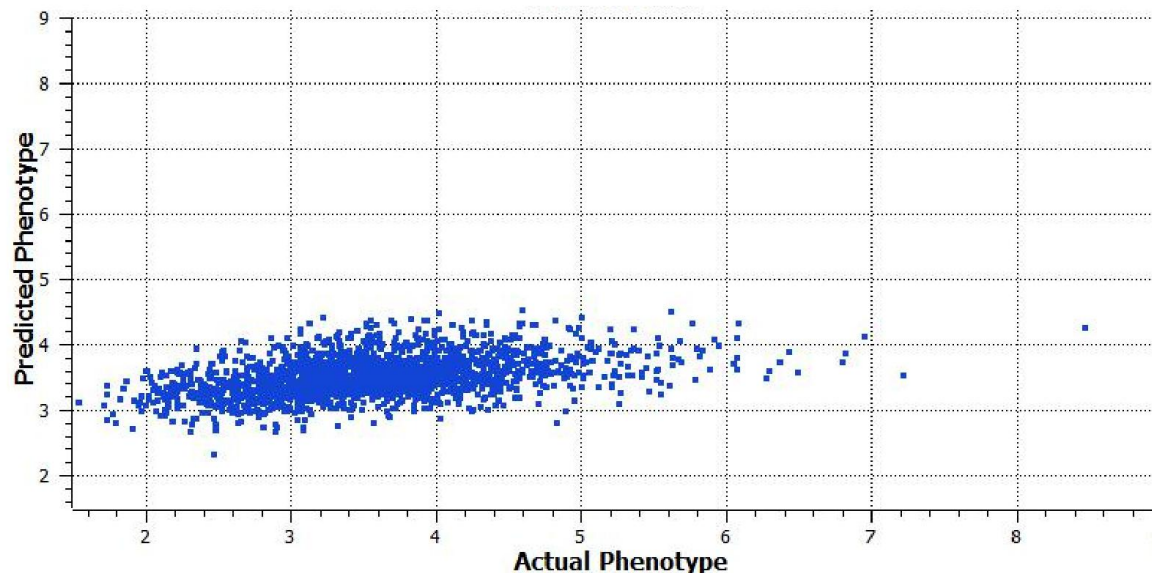
Pearson's Product-Moment Correlation Coefficient: 0.4015587129
 Residual Sum of Squares: 198.6182286880
 Total Sum of Squares: 236.5105178469
 R-Squared: 0.1602139706
 Root Mean Square Error: 0.6893208444
 Mean Absolute Error: 0.5236755179

Fold 2

Pearson's Product-Moment Correlation Coefficient: 0.4448843677
 Residual Sum of Squares: 195.7270701054
 Total Sum of Squares: 241.8604776538
 R-Squared: 0.1907438867
 Root Mean Square Error: 0.6859283962
 Mean Absolute Error: 0.5213248258

Fold 3

Pearson's Product-Moment Correlation Coefficient: 0.3887937612
 Residual Sum of Squares: 228.8890867620
 Total Sum of Squares: 267.1873564337
 R-Squared: 0.1433386302
 Root Mean Square Error: 0.7426573692
 Mean Absolute Error: 0.5805400467



GBLUP - ASE - Iteration 1 [2141]

Unsort	R	1	R	2	R	3
Map	Marker	Allele substitution effect (ASE)	Absolute value of ASE	Normalized Abs ASE		
1	Hapmap43437-BTA-101873	-0.000368841373186229	0.000368841373186229	26.0085477384702		
2	ARS-BFGL-NGS-16466	0.000173913631369154	0.000173913631369154	12.2633774643051		
3	ARS-BFGL-NGS-19289	0.000107101214065017	0.000107101214065017	7.55215450695027		
4	Hapmap34944-BES1_Contig627_1906	6.57063929251151e-005	6.57063929251151e-005	4.6332325529346		
5	BTA-07251-no-rs	-1.05547432562128e-005	1.05547432562128e-005	0.744259087518099		
6	ARS-BFGL-NGS-98142	2.37371645487265e-005	2.37371645487265e-005	1.67380674247128		
7	Hapmap53946-rs29015852	-0.000650784019124274	0.000650784019124274	45.8895028033669		
8	ARS-BFGL-NGS-114208	-0.000455347554385593	0.000455347554385593	32.1084603484919		
9	ARS-BFGL-NGS-66449	-0.000191744079108326	0.000191744079108326	13.5206769023169		
10	ARS-BFGL-BAC-32770	4.8994743326343e-005	4.8994743326343e-005	3.45482425067834		
11	ARS-BFGL-NGS-65067	-0.000101175402819359	0.000101175402819359	7.13430077394715		
12	ARS-BFGL-BAC-31497	1.87061319691707e-005	1.87061319691707e-005	1.31904759522911		

Accuracy of gEBV

- Genetic correlation between gEBV and phenotype.



- Bivariate animal model in **Wombat**

```
===== Version 12-06-2015 ===== **KM** =====
Program WOMBAT: Estimates of covariance components
=====

Genetic Correlation gEBV and WBSF

Analysis type      : "MUV 2"
Data file          : "gEBV-wbsf.dat"
Pedigree file      : "PedFile.dat"
Parameter file     : "rgGBVwbsf.par"

No. of traits      =      2   gEBV   WBSF
No. of records     =   4152   2076   2076
No. of parameters  =      6
Maximum log L      =                2008.335

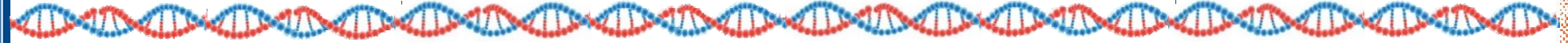
***** Estimates for RE 1 "animal" *****
      No. of levels      =   4121
      Covariance structure =   NRM
      Order of fit       =      2

Covariance matrix
  1   0.69448E-01
  2   0.36425E-01  0.55370E-01
Eigenvalues of covariance matrix
Value      0.10      0.03
(%)       79.72     20.28
Matrix of correlations and variance ratios
  1   0.9857
  2   0.5874   0.1519

Covariances & correlations & approx. sampling errors
  4 COVS A 1 1  0.6944E-01  0.5609E-02  vrat  0.986  0.055
  5 COVS A 1 2  0.3642E-01  0.7192E-02  corr  0.587  0.107
  6 COVS A 2 2  0.5536E-01  0.1644E-01  vrat  0.152  0.044
```

Accuracy:
0.59

Palatability of Beef - goals



1 Genome-wide association (GWAS)

2 Genomic Prediction

3 **GWAS signals → underlying biology**

Palatability trait



- **8-10 individual traits (sub-phenotypes):**

- 50,000 SNP effects for each one on >2,000 animals

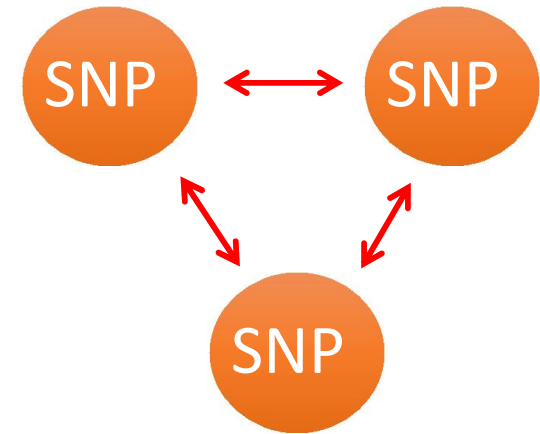
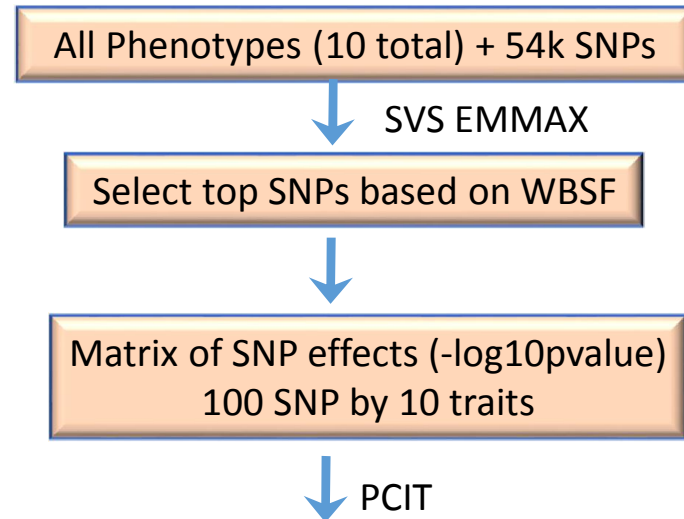
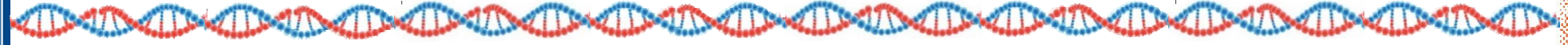
- **Systems biology: integration of data sets**

- Holistic view of the system – key players can emerge

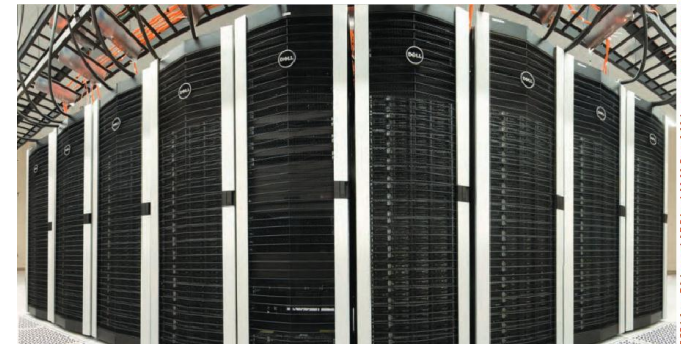
Gene network theory: use SNP association data to guide the inference of gene regulatory networks

Association Weight Matrix (Fortes 2010)
multivariate view of GWAS using PCIT (partial correlation and information theory Reverter & Chan 2008)

Network Analysis (Palatability)



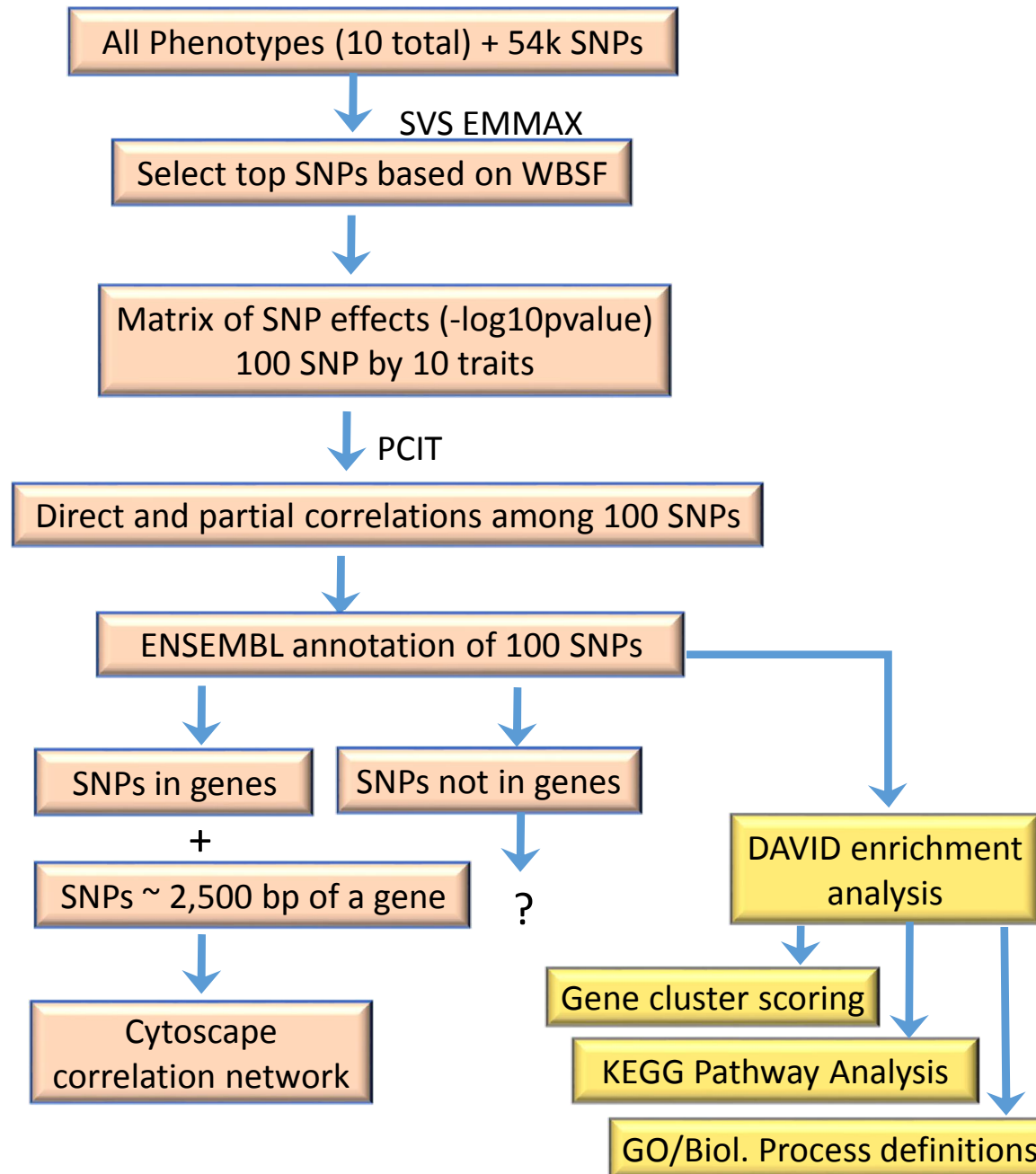
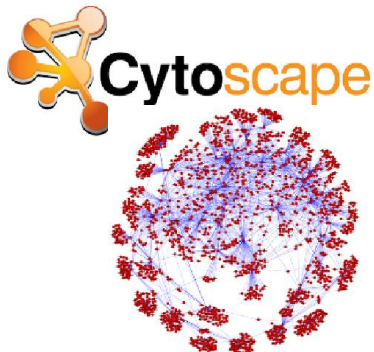
- **Partial Correlation Information Theory** *(Reverter and Chan, 2008)*
- Significant SNPs correlated among multiple sub-phenotypes are important for regulating the overall phenotype
 - Optimized to handle SNPs and genomic windows



HiPerGator
The University of Florida Supercomputer for Research

Network Analysis (Palatability)

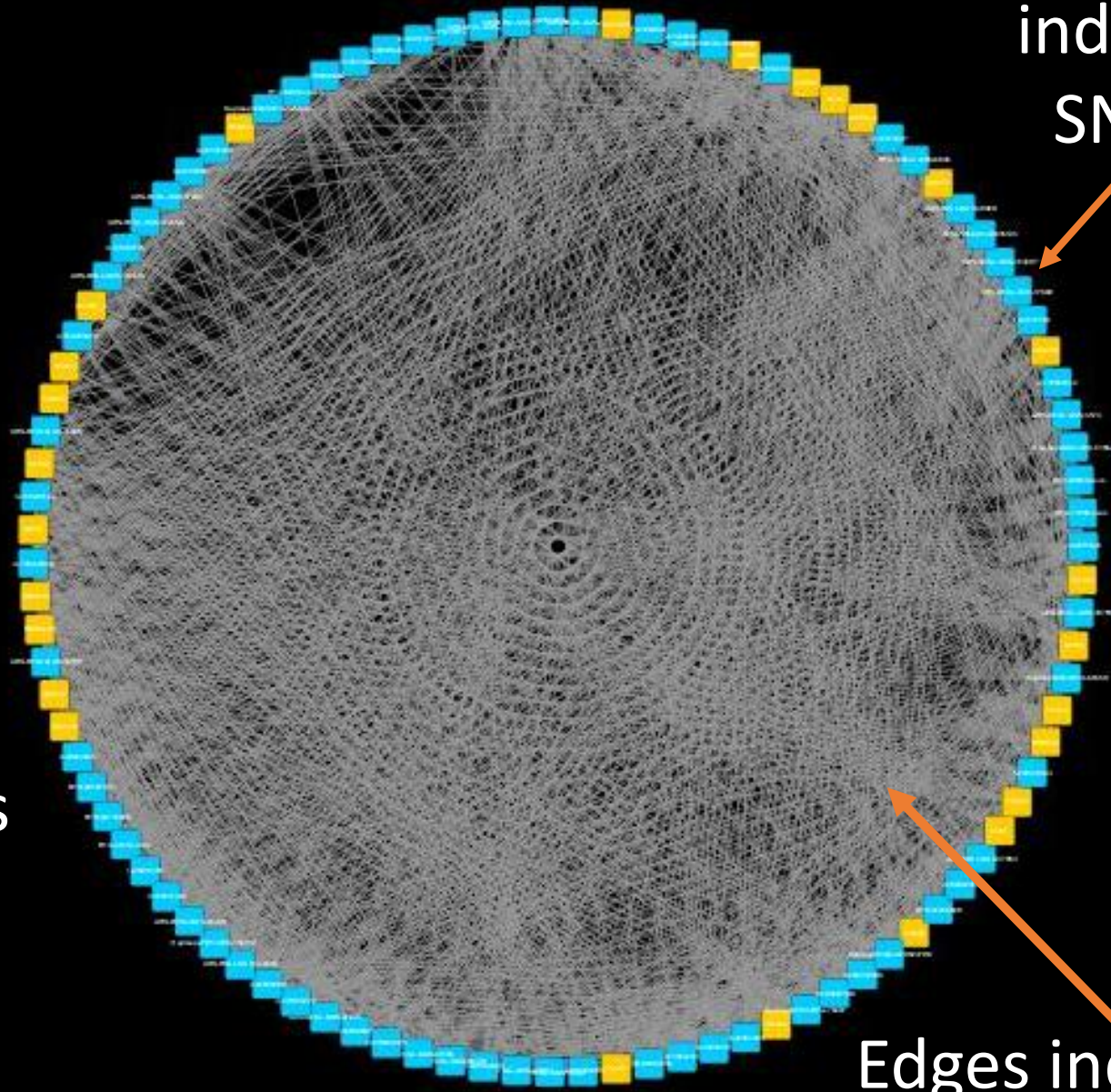
- **Networking and Clustering**
- Direct correlations visualized in **Cytoscape**
- Clustering analysis - **MCODE** - highly connected genes
- Enriched **GO** terms were identified using **DAVID** and visualized using **REViGO**



Network Visualization

Reference
phenotype -
**Warner-Bratzler
Shear Force**
Most significant
SNPs

• **99** SNPs



Nodes
indicate
SNP's

Edges indicate
correlations

Network Visualization

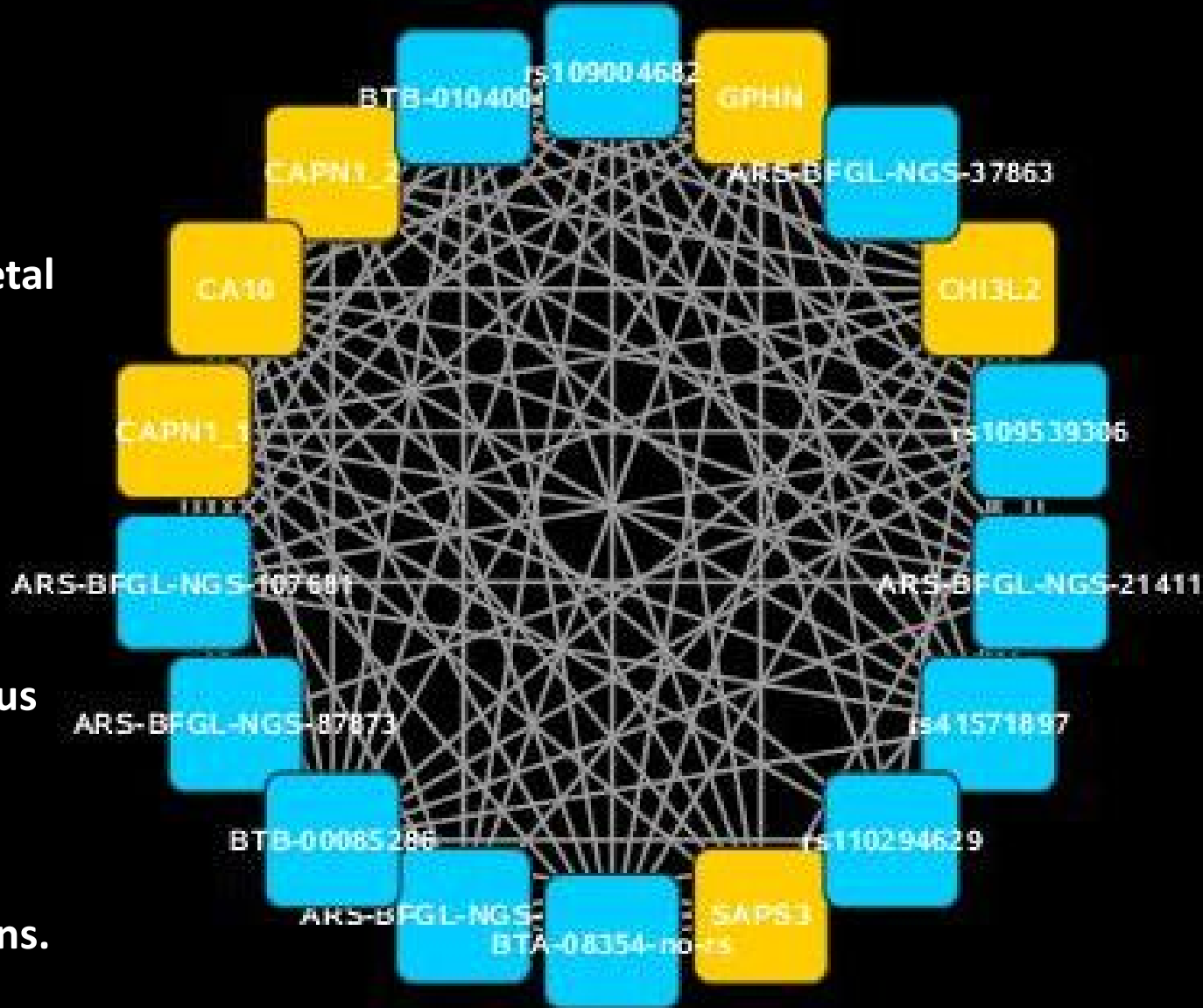
SAPS3: modulates protein phosphatase catalytic subunits

CAPN1: modulates proteolysis of cytoskeletal remodeling and signal transduction.

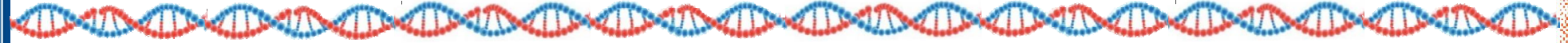
CHI3L2: Involved in cartilage biogenesis.

CA10: Catalyzes reversible hydration of carbon dioxide in various processes.

GPHN: Involved in membrane protein-cytoskeleton interactions.



Conclusions



- Palatability or eating satisfaction - important for the long-term **sustainability** of beef industry
- Collection of **palatability** phenotypes on large numbers of animals is still problematic
 - Findings focused on QTL detection rather than genomic prediction
- New methods and approaches to move from SNP signals closer to **functional variants**

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Questions?