

Biological interpretation of selection footprints

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Overview









GENOMENV: Biological interpretation of selection footprints



- Main goals
- Identify the main functions and biological pathways in which the genes are involved
- Confirm the biological importance of the genes that have been identified as being positively selected
- Strengthen the credibility of the positive selection model by the development of a sound scenario



Course outline

1. From regions under selection to biological interpretation

- Annotation of candidate genes using systems biology tools
 - Functional enrichment analysis: Gene ontology, pathway analysis
 - · Gene networks analysis
- Inferring the main selective pressures
- Interpretation and story-telling
- Including phenotypes and environmental covariates
- Identifying candidate mutations and prioritizing candidate genes

2. Interpretation of selection footprints: some examples

- Phenotype-free approaches
 - Manual functional annotation: Senepol cattle breed
 - Functional annotation using systems biology tools: french dairy cattle breeds, West-African cattle breeds, European bison/cattle
- Association with covariates
 - Phenotypes: dairy traits in French cattle breeds





1. From regions under selection to biological interpretation









From regions to candidate genes

- Identification of chromosomal regions under selection
 some arbitrary criteria
 - Sliding windows (e.g. 1Mb, 500kb overlap)
 - Nb of SNPs with score>significant threshold
 - If several overlapping regions under selection: merging regions or choosing region with the highest peak and the highest proportion of significant SNPs
- Mapping of regions on corresponding genome assembly on ucsc or ensembl (if available)
- List of all genes within regions (Refseq and Other species Refseq)
- Criteria to identify one or a few candidate genes per region
 - e.g. proximity to the peak (e.g. 15-100kb from gene boundaries)



From candidate genes to biological interpretation

- List of candidate genes with/without associated score
 - Short list
 - Long list
- Misleading gene names
- Possibly hundreds of papers describing gene functions

How to find biological sense?







Annotation using systems biology tools

- Systems biology
- Tools used to interpret results of transcriptomic and proteomic experiments
- e.g.: list of differentially expressed genes between different experimental conditions, different tissues





Annotation using systems biology tools: Functional enrichment analysis

- Is the list of candidate genes statistically enriched for some functions or some biological pathways?
- Needs:
 - Shared vocabulary: Gene Ontology, Biological pathways
 - Annotation of genes : association between terms and genes or gene products.
 - Statistical tests of enrichment



Annotation using systems biology tools Functional enrichment analysis: Gene Ontology

- Gene ontology (GO): hierarchical vocabulary of terms describing genes and protein
- Three GO represented by a root ontology term
 - cellular component referring to the place in the cell (e.g. nucleus, ribosome)
 - biological process referring to a biological objective (e.g. signal transduction, alphaglucoside transport)
 - molecular function describing activities (e.g. catalytic activity, Toll receptor binding)
- Developed by a consortium (geneontology.org)



- => collaborative effort
- Widely used biological ressource
- 41 species
- Generic terms / Some species-specific terms



Annotation using systems biology tools Functional enrichment analysis: Gene Ontology

GO structure:

- directed acyclic graph
- types of relationships of child to parent: "is a" or "part of"
- each term has defined relationships to one or more other terms.



geneontology.org



Annotation using systems biology tools Functional enrichment analysis: Gene Ontology

However some (commercial) tools developped their own ontology

e.g.: Ingenuity Pathway Analysis (IPA) => IPA ontology three main categories of function

- Diseases and disorders
- Molecular and cellular functions
- Physiological system development and function





Annotation using systems biology tools Functional enrichment analysis: Biological pathway

Biological pathway

- Biochemical engines responsible for the transduction of signals into output responses
- A series of actions among molecules in a cell that leads to a certain product or a change in the cell.

Pathway building

- Process of identifying and integrating the molecules entities, interactions, and associated annotations
- Contribute to the knowledgebase.
- Can have either a data-driven or a knowledge-driven objective



Annotation using systems biology tools Functional enrichment analysis: Biological pathway

INGENUITY

PATHWAY ANALYSIS

- Pathway databases
 - e.g. consortium effort







- Integration of several data sources
 - E.g.



commercial database





- Which functional category or biological pathway is more prevalent in the gene list than expected by chance?
- Tests
 - Fisher's exact test (one-tailed, right)
 - Gene Set Enrichment Analysis
 - Correction for multiple testing



Fisher's exact test (one-tailed, right)



Modified Fisher's exact test (one-tailed, right): e.g. EASE score More conservative: uses s_i-1 instead of s_i



Gene Set Enrichment analysis (GSEA) Subramanian et al, PNAS, 2005

- Determines if an *a priori* defined set of genes are statistically significant (presumably concordantly different) between two biological states.
 - > Sets of genes can be those within a pathway, biological process, etc.
 - > Statistical significance determined by permutation (shuffling of the data)
- Strategy
 - > The genes are ordered on the basis of the parameter from the statistical test
 - For each gene set compute an enrichment score (ES) is computed (i.e a measure of how relevant or associated a biological process is for discerning the difference between the two biological states)
 - > Create a running sum of a normalized Kolmogorov-Smirnov (non-parametric test) statistic.
 - Permute the class labels a large nb of times, each time recording the maximum ES over all gene sets.
 - > Compare the observed ES score to the distribution of the ES scores from the permuted data.
 - > Test the hypothesis that no gene set is associated with the class distinction



- Gene Set Enrichment analysis (GSEA)
 - A user-supplied ranked list of genes could also be used.
 - It determines whether a priori defined sets of genes show statistically significant enrichment at either end of the ranking.
 - A statistically significant enrichment indicates that the biological activity (e.g., biomolecular pathway) characterized by the gene set is correlated with the usersupplied ranking



 Which functional category or biological pathway is more prevalent in the gene list than expected by chance?

- Fisher's exact test (parametric)
- Gene Set Enrichment Analysis (non-parametric)
- Multiple testing correction:
 - Bonferroni correction or false discovery rate (FDR)



Functional enrichment analysis Tools

- •DAVID
- •EASE
- •AmiGO
- •GeneGo MetaCore
- GOMiner
- BiNGO & ClueGO integrated with Cytoscape
- sigPathway & GOStat (R/Bioconductor based)
- FuncAssociate
- FatiGO
- GOEAST
- TopGO
- Gene Set Analysis (GSA)
- GSEAPreranked

- KEGG
- Wikipathway
- Reactome
- Genemania in Cytoscape
- PANTHER
- InnateDB
- STRING
- Ingenuity Pathway Analysis



Functional enrichment analysis Tools

DAVID Bioinformatics Resources 6.7 National Institute of Allergy and Infectious Diseases (NIAID), NIH

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Shortcut to DAVID Tools

Functional Annotation

Gene-annotation enrichment analysis, functional annotation clustering , BioCarta &

KEGG pathway mapping, gene-disease

Gene Functional Classification

ovide a rapid means to reduce large lists

of genes into functionally related groups of

genes to help unravel the biological content

Convert list of gene ID/accessions to others of your choice with the most comprehensive gene ID mapping repository. The ambiguous accessions in the list can also be determined

Gene Name Batch Viewer

Display gene names for a given gene list; Display gene names for a growing of this your Search functionally related genes within your list or not in your list; Deep links to enriched detailed information. <u>More</u>

association, homologue match, ID translation, literature match and more

aptured by high throughput

emi-automatically. More

Gene ID Conversion

echnologies. More

Recommending: A paper published in Nature Protocols describes step-by-step procedure to use DAVID!

Welcome to DAVID 6.7

2003 - 2016

The Database for Annotation, Visualization and Integrated Discovery (DAVID) v6.7 is an update to the sixth version of our original web-accessible programs. DAVID now provides a comprehensive set of functional annotation tools for investigators to understand biological meaning behind large list of genes. For any given gene list, DAVID tools are able to:

- Identify enriched biological themes, particularly GO terms
- Discover enriched functional-related gene groups
- Cluster redundant annotation terms
- Visualize genes on BioCarta & KEGG pathway maps
- S Display related many-genes-to-many-terms on 2-D
- view.

DATABASE

- Search for other functionally related genes not in the list
- List interacting proteins
- Explore gene names in batch
- Link gene-disease associations
- Highlight protein functional domains and motifs
- Redirect to related literatures
- Convert gene identifiers from one type to another. And more



What's Important in DAVID?

- · Current (v 6.7) release note
- · New requirement to cite DAVID
- · IDs of Affy Exon and Gene arrays supported

Search

- · Novel Classification Algorithms
- Pre-built Affvmetrix and Illumina backgrounds
- User's customized gene background
- · Enhanced calculating speed

Statistics of DAVID

DAVID Bioinformatic Besources Citations



- > 21,000 Citations
- Average Daily Usage: ~2,600 gene lists/sublists from ~800 unique researchers.
- Average Annual Usage: ~1,000,000 gene lists/sublists from >5,000 research institutes world-wide



Functional enrichment analysis Limits

- Function enrichment analysis (GO)
 - Structure of GO: difficult to determine which level of hierarchy is most responsible for statistical enrichment
 - The most enriched terms are often broad functional categories, not very informative

Pathway analysis

- The majority of genes have not been assigned to a pathway
- Bias towards well-studied signalling pathways
- Informative about information already known

Statistical tests

- Choice of the reference gene set
- **Species of interest:** some tools are species specific





 Biological processes are mainly controlled by complex networks of molecular interactions

Network

- graph in which an entity (molecule or metabolite)
 is represented by a **node** and entities' relationships
 by **edges** between nodes
- Not restricted to one type of nodes of edges
- Based on publicly available data such as experimentally-supported interactions (e.g. protein-protein interactions: PPI)







• Different type of interactions:

- physical (e.g. PPI, protein-DNA)
- regulatory (e.g. miRNA-mRNA)
- biochemicals interactions (e.g.phosphorylation)

• Experimentally-validated interation data obtained by:

- Primary database: Text-mining from peer-reviewed litterature
 - Manual curation from peer-reviewed litterature
- Meta-databases :integration of different sources
- Some databases included *in silico*-predicted interactions
- Limited overlap between interaction information of primary databases
- Need integartion of information provided by several primary databases



Gene network analysis Molecular interactions databases

- Need integration of information provided by several primary databases: e.g.
 - Web service PSICQUICK (integrated in Cytoscape and in Bioconductor/R)

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Click on the links below to display The clustering will be enabled with	the results for each selected service. less than 5000 interactions.			
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- Commercial: IPA, manual curation by experts
- The meaning of an edge could vary as it integrates several different types of interaction



Gene network analysis A complementary approach to enrichment analysis methods

- More data-driven
- Interactome (molecular interactions within a biological system) available for several species
- Less biased towards well-studied pathways
- Additional possible integration of information associated with nodes or edges
- Software development for network visualization





- Integration of different type of interactions
 - The meaning of an edge could vary.

• Level of confidence associated with interactions

- Some technics used to predict PPI in large-scale studies are associated with a high false positive rate (e.g. Yeast-2 Hybrid)
- For more focused studies, biases toward well-studied biological processes
- Interactions are context-specific
- Species



Gene network analysis Tools

- Cytoscape (open source):
 - open source software platform for visualizing complex networks and integrating these with any type of attribute data.

a lot of Apps (plugins) available (most of them freely)
 Cytoscape Apps store:
 http://apps.cytoscape.org/apps/all









Cytoscape interface







Ingenuity Pathway Analysis (commercial)



Links between tools



INGENUITY[°]

PATHWAY ANALYSIS

Gene network analysis Properties and features of gene networks

Hubs

i.e. High degree nodes

- > Node degree: number of interactions/edges that a node has
- Important for the network structure
- Central
- Less targeted by selection



- Bottlenecks
 - i.e. Nodes with a high betweenness centrality
 - > nodes which are the crossroads of many shortest paths
 - > Distance between nodes: minimum number of steps between them
 - i.e. Major bridges



Gene network analysis Properties and features of gene networks

Modules

- i.e. group of molecules that preferentially interact which each other (sub-networks)
- Tend to be enriched for common biological functions or diseases







- Tools to identify gene networks properties
 - Network Analyst
 - Cytoscape Apps:
 - CytoHubba
 - jActiveModules
- Cannot be done with some tools (e.g. IPA)
 - Export network in Cytoscape format (need sometimes a specific licence)


Gene network and selection

- Position of selected genes selected within PPI network
 - Kim et al, 2007
 - > Hubs tend to be constrained and under negative selection
 - Observed positive selection at the network periphery
 - Qian et al, 2014
 - > Signal of selection tend to be underrepresented for genes with fewer neighbors
 - > Tend to enrich at subcentral position of the PPI network
 - > Molecules of high centrality could be under strong evolutionary constraints
 - > Molecules at periphery may not contribute enough to phenotypic effect

- Interactions between recently selected genes (Qian et al, 2014)
 - Closer interaction among genes under selection consistent with the effect of coselection



Inferring the main selective pressure

 Gene network reconstruction and identification of the main functions targeted by selection (IPA, Cytoscape)





Interpretation and over-interpretation

Goals of functional annotation of selection footprints

- Confirm the biological importance of the genes that have been identified as being positively selected.
- Stengthen the credibility of the positive selection model by a development of a sound scenario

Sometimes a misleading approach prone to storytelling

- The majority of genes in a genome have important biological functions
- easy to find connections between key genes using GO annotation and litterature mining tools
- Ex: Pavlidis et al, MBE, 2012



Integrating additional information: **QTL** location

QTL databases



Sheep QTL

Cattle QTL

Chicken QTL

Horse QTL

Pig QTL

 \otimes

There are 1,173 QTLs from 114 publications curated into the database. Those QTLs represent 204 different traits (see data summary for most recent updates).

Release 29 (Apr 29, 2016)







CattleQTLdb

Browse the Cattle QTLdb

Option 1: by Chromosomes

Chromosome 1	Chromosome 11	Chromosome 21
Chromosome 2	Chromosome 12	E Chromosome 22
Chromosome 3	Chromosome 13	Chromosome 23
Chromosome 4	E Chromosome 14	E Chromosome 24
Chromosome 5	Chromosome 15	Chromosome 25
Chromosome 6	E Chromosome 16	E Chromosome 26
E Chromosome 7	Chromosome 17	E Chromosome 27
E Chromosome 8	E Chromosome 18	E Chromosome 28
E Chromosome 9	E Chromosome 19	E Chromosome 29
Chromosome 10	Chromosome 20	Chromosome X

Option Inc. The WOLLS AND

2

3

6

m 2 : by trait classes	17
	18
Health Traits	19
Meat and Carcass Traits	20
Milk Traits	22
Production Traits	23
Reproduction Traits	24 25
Exterior Traits	26
	27
	28
	29



Browse Search View Maps

Cattle QTLdb at a Glimpse

Counts by Chromosomes

Chromsome

14

15

х

QTLs Found

1867

1865 1485

5264

2489

3937

1758 917

1390

2212 887

1540 3790

914

1077

1367 1351

2149

739 823

544

3163

556

684

Association with phenotypes and environmental covariates





Association with phenotypes and environmental covariates

- Identifying loci underlying local adaptation using correlations between allele frequencies and ecological population variables or phenotypes
 - Multi-dimensional methods (PCA, Laloë et al, in prep)
 - Bayesian model-based approaches (e.g. Baypass; Gautier, 2015)
 - LFMM (Frichot et al, MBE, 2013)

e.g Mediterranean cattle breeds (GALIMED project): SNPs contribution to the genetic variability according to mean temperature



marker



Identifying and annotating candidate mutations and prioritizing candidate genes





Identifying candidate mutations

- Individual whole genome sequencing
 - Alignment on reference genome
 - Variant calling
 - Calculation of allelic frequencies
 - Variant annotation

Comprehensive list of variants in the genome with estimation of their frequency





• Exhaustive list of variants (SNP and indels)

• Location of the variant / genes position in reference sequence



 Estimation of variant consequences on transcription and protein structure





- Tools
- e.g.





Use sequence ontology

- i.e. set of terms and relationships used to describe the features and attributes of biological sequence
- Initially developed by the GO consortium



- Integrate different information available on variants, e.g.
 - SIFT
 - GERP scores



- Use sequence ontology
 - i.e. set of terms and relationships used to describe the features and attributes of biological sequence
 - Initially developed by the GO consortium



• Relationship





Sequence ontology

sequence_attribute
 sequence_collection
 sequence_comparison
 sequence_feature

• Obsolete Terms

• Relationship

- sequence_attribute
 feature_attribute
 polymer_attribute
 nucleic acid
 - peptidyl
 - synthetic_sequence
 - topology_attribute
 - [•]sequence_location
 - organelle_sequence
 plasmid_location

 - [⊡]variant_quality
 - coding_variant_quality
 variant_frequency
 - variant_origin
 - [⊕] variant_phenotype

sequence_collection
 contig_collection
 genome
 peptide_collection
 variant_collection

	· _	÷
ŀ	³ sequence_comparison	
	"no_variation	
	🖹 sequence_variant	
	Functional_variant	
	"dominant_negative_variant	
	"gain_of_function_variant	
	"lethal_variant	
	"loss_of_function_variant	
	"loss_of_heterozygosity	
	"null_mutation	
	transcript_function_variant	
	"level_of_transcript_variar	
	transcript_processing_va	
	transcript_stability_variar	
	transcription_variant	
	Translational_product_functi	
	🖹 structural_variant	
	copy_number_change	
	feature_ablation	
	feature_amplification	
	[⊕] feature_fusion	
	feature_translocation	
	[.] feature_variant	

he Sequence Ontology Project

Home Browser Wiki GFF3 Resources About Request A Term Site Map

sequence_feature iunction Chromosome breakpoint clone insert end clone_insert_start deletion junction exon_junction insertion_site "polyA_site restriction_enzyme_cleavage_ "splice_junction "trans_splice_junction region [⊕]biological region biomaterial region experimental_feature topologically defined region sequence alteration . ⊡ UPD "copy_number_variation • deletion indel insertion • inversion "structural_alteration • substitution



Sequence ontology in VEP (ensembl)





- Prediction of regulatory variant effect on transcription/expression
 - e.g. variant within TFBS
 - Integrated in VEP for some species
 - Partial annotation in regulatory regions for other species need other tools: e.g. Lasagna, mrSNP
- Prediction of variant effect on protein structure:
 - SIFT: applies on non-synonymous polymorphism and predicts whether an amino acid substitution affects protein function.
 - based on the degree of conservation of amino acid residues in sequence alignments derived from closely related sequences.
 - > Tolerated/Deleterious (high, low confidence)



Prediction of variant effect in the whole sequence

- Genomic Evolutionary Rate Profiling (GERP) :

a method for producing position-specific estimates of evolutionary constraint using maximum likelihood evolutionary rate estimation and discovering "constrained elements" that is indicative of a putative functional element.

- > Based on multiple sequence alignments and phylogenic tree
- Constraint intensity quantified in terms of a "rejected substitutions" (RS) score, i.e. the number of substitutions expected under neutrality minus the number of substitutions "observed" at the position.
- Positive scores represent a substitution deficit and thus indicate that a site may be under evolutionary constraint. Negative scores indicate that a site is probably evolving neutrally





Different categories based on impact on protein level or protein structure, e.g.:

Strong impact:

- Non synonymous mutation deleterious (SIFT)
- 5'UTR within TFBS
- 3'UTR within miRNA binding site
- In splice region
- With positive GERP score

Weakest impact

- 5'UTR not in TFBS
- 3'UTR not in miRNA binding site
- Downstream
- Upstream
- Intronic
- Non synonymous mutation tolerated





- Criteria
 - Proximity to the position with the highest score (peak)
 - Number of variants in categories with strong effect (after variant annotation)
 - Allelic frequencies
 - mutation fixed (freq=1)
 - Mutation with high frequency
- Functional annotation of these new list of candidate genes using systems biology tools





2. Some examples of biological interpretation





Course outline

1. From regions under selection to biological interpretation

- Annotation of candidate genes using systems biology tools
 - Functional enrichment analysis: Gene ontology, pathway analysis
 - Gene networks analysis
- Inferring the main selective pressures
- Interpretation and story-telling
- Including phenotypes and environmental covariates
- Identifying candidate mutations and prioritizing candidate genes

2. Interpretation of selection footprints: some examples

Phenotype-free approaches

- Manual functional annotation: Senepol cattle breed
- Functional annotation using systems biology tools: french dairy cattle breeds, West-African cattle breeds, European bison/cattle
- Association with covariates
 - Phenotypes: dairy traits in French cattle breeds



Manual functional annotation The exemple of the Senepol cattle breed

- Senepol: a European taurine breed with a small proportion of zebu ancestry.
- Living in tropical area (Caribbean, St Croix island)
- Identification of selection footprints
 - 153 individuals genotyped on 47,365 SNPs
 - Tests based on the extent of haplotype homozygosity: iHS (Voight et al, 2007) and Rsb (Tang et al, 2007)
 Package rehh, Gautier & Vitalis, 2012
 - 1Mb sliding windows (0.5Mb overlap)
 - Selection of regions with at least 2 SNPs exceeding the significant threshold (P<0.0001)





Flori et al, 2012, PLoS One

Manual functional annotation The exemple of the Senepol cattle breed

• Only four regions under selection

Region	ВТА	Position (Mb)	Peak position (Mb)	iHS _{SEN}	Rsb _{EUT/SEN}	Rsb _{ZEB/SEN}	Nb of significant SNPs	Gene closest to the maximum
#1	1	52.6-53.6	52.9	3.984	NS	3.984	2	No gene found
# 2	1	2.4-3.4	3.2	NS	4.161	NS	6	TIAM1
#3	1	4.7-5.8	5.5	NS	4.538	NS	8	GRIK1
#4	20	38.6-39.6	39.5	4.055	4.576	4.055	3–6	RAI14

BTA1 =>polled locus BTA20 =>*slick* locus involved in a short and sleek hair coat and in thermotolerance/adaptation to tropical conditions

Identification of candidate gene: boundaries located less than 25kb from the peak position



Manual functional annotation The exemple of the Senepol cattle breed

- Identification of the mutation responsible of the slick phenotype within the PRLR (Littlejohn et al, 2014)
- PRLR is located in the region #4 found under selection in Senepol







The three main French dairy cattle breeds =>artificial selection



Breed	Herd-Book creation	Population size (2002)	Ne (généalogie)	Milk productio n <i>(kg)</i>	Content (°/oo fat/Protein)
MON	1872	1,799,200	34	7,441	38.8 / 32.5
NOR	1883	2,106,000	61	6,595	44.2 / 36.0
HOL	1922	11,535,378	42	8,628	40.9 / 31.6

Increase in milk production but decline in reproductive performances



Identification of selection footprints

- 2,803 bulls genotyped on 42,486 SNPs
- Differentiation (Fst) between breeds (Nicholson et al, 2002)
- Empirical approach
 - combining information between closely related SNPs
 - Smoothed individual SNP Fst values over each chromosome for observed and simulated datasets
 - Calculation of local q-values
- **13 regions under selection** (qvalue<0.05)
- Some regions contain genes with causal variants with a strong effect on milk production trait (GHR) or coloration (MC1R)



Regions under selection

BTA	Start-End (peak position) in Mb	$F_{\rm S7}$ at the peak position (qvalue)	candidate gene	Breeds within which region is also significant
3	57.084-58.505 (58.343)	0.375 (0.0298)	CCCBL2	
4	78.833-80.43 (79.701)	0.667 (0.0298)	NUDCD3	
5	20.301-23.091 (21.02)	0.483 (0.0298)	na	NOR, HOL
5	97.803-100.826 (98.26)	0.557 (0.0298)	PIK3C2G	NOR, HOL
5	108.461-109.236 (109.182)	0.403 (0.0401)	CD163	
5	110.286-111.861 (111.552)	0.46 (0.0435)	ANO2	
6	37.433-38.756 (37.963)	0.566 (0.0298)	LAP3/LCORL	MON
6	66.599-66.935 (66.809)	0.165 (0.0435)	na	
6	68.938-76.32 (72.024)	0.616 (0)	PDGFRA	NOR
14	22.02-25.567 (22.634)	0.591 (0)	na	MON, NOR
18	12.987-14.058 (13.36)	0.632 (0)	MC1R	MON, HOL
20	31.964-33.757 (32.277)	0.523 (0.0298)	GHR	
26	22.137-23.191 (22.983)	0.509 (0.0298)	C10ORF76	
	BTA 3 4 5 5 5 5 6 6 6 6 6 14 18 20 26	BTA Start-End (peak position) in Mb 3 57.084-58.505 (58.343) 4 78.833-80.43 (79.701) 5 20.301-23.091 (21.02) 5 97.803-100.826 (98.26) 5 108.461-109.236 (109.182) 5 102.866-111.861 (111.552) 6 37.433-38.756 (37.963) 6 65.99-66.935 (66.809) 6 68.938-76.32 (72.024) 14 22.02-25.567 (22.634) 18 12.987-14.058 (13.36) 20 31.964-33.757 (32.277) 26 22.137-23.191 (22.983)	BTA Start-End (peak position) in Mb <i>F_{sT}</i> at the peak position (qvalue) 3 57.084-58.505 (58.343) 0.375 (0.0298) 4 78.833-80.43 (79.701) 0.667 (0.0298) 5 20.301-23.091 (21.02) 0.483 (0.0298) 5 97.803-100.826 (98.26) 0.557 (0.0298) 5 108.461-109.236 (109.182) 0.403 (0.0401) 5 108.461-11.861 (111.552) 0.466 (0.0435) 6 37.433-38.756 (37.963) 0.566 (0.0298) 6 65.599-66.935 (66.809) 0.165 (0.0435) 6 68.938-76.32 (72.024) 0.616 (0) 14 22.02-25.567 (22.634) 0.591 (0) 18 12.987-14.058 (13.36) 0.523 (0.0298) 20 31.964-33.757 (32.277) 0.523 (0.0298) 26 22.137-23.191 (22.983) 0.509 (0.0298)	BTA Start-End (peak position) in Mb F _{S7} at the peak position (qvalue) candidate gene 3 57.084-58.505 (58.343) 0.375 (0.0298) CCCBL2 4 78.833-80.43 (79.701) 0.667 (0.0298) NUDCD3 5 20.301-23.091 (21.02) 0.483 (0.0298) na 5 97.803-100.826 (98.26) 0.557 (0.0298) PIK3C2G 5 108.461-109.236 (109.182) 0.403 (0.0401) CD163 5 108.461-109.236 (109.182) 0.466 (0.0435) ANO2 6 37.433-38.756 (37.963) 0.566 (0.0298) LAP3/LCORL 6 65.99-66.935 (66.809) 0.165 (0.0435) na 6 65.99-66.935 (66.809) 0.161 (0) na 6 68.938-76.32 (72.024) 0.591 (0) na 14 2.02-25.567 (22.634) 0.632 (0) na 18 1.987-14.058 (13.36) 0.632 (0.0298) GHR 20 31.964-33.757 (32.277) 0.523 (0.0298) GHR 21 2.137-23.191 (22.983) 0.509 (0.0298) C100RF76



- Some regions contain genes with causal variants with a strong effect on milk production trait (GHR) or coloration (MC1R)
- Some regions contain loci involved in morphological traits



Holstein



7 eligible genes/8







14 eligible genes/19







Montbeliarde



13 eligible genes /16

- Significant networks contain different genes under selection in the three breeds
- But genes under selection in the each breed are involved in the same biological pathway
- Genome Plasticity of genome response to the same selective pressure



- Global gene network
 - contains genes under selection in at least one breed
- Central role of somatotropic and gonadotropic axes in response to artificial selection
- Illustrates the antagonism between milk production and reproduction





Course outline

1. From regions under selection to biological interpretation

- Annotation of candidate genes using systems biology tools
 - Functional enrichment analysis: Gene ontology, pathway analysis
 - Gene networks analysis
- Inferring the main selective pressures
- Interpretation and story-telling
- Including phenotypes and environmental covariates
- Identifying candidate mutations and prioritizing candidate genes

2. Interpretation of selection footprints: some examples

Phenotype-free approaches

- Manual functional annotation: Senepol cattle breed
- Functional annotation using systems biology tools: French dairy cattle breeds, West-African cattle breeds, European bison/cattle
- Association with covariates
 - Phenotypes: dairy traits in French cattle breeds



Annotation using systems biology tools The exemple of the West-African cattle breeds

• West-African cattle: models of adaptation to tropical conditions



Gautier et al, 2009, BMC Genomics



Exemple of the West-African cattle breeds : annotation using systems biology tools

- Identification of selection footprints
 - 342 individuals genotyped on 36320 SNPs
 - Differentiation (Fst, Bayesian model) between 9 West-African populations
 - Decision rule to identify no-neutral loci based on BF, expressed in deciban unit (dB_i=10log₁₀(BF_i))
 - Smoothed individual SNP BF values over each chromosome
 - Permutations to estimate local p-values
 - Correction of the local p-values by computing q-value
- **53 regions under selection** (at the 5% local FDR)



Exemple of the West-African cattle breeds : annotation using systems biology tools

SNP annotation using Transmap Refseq

- 46,598 Refseq ID anchored to the bovine genome assembly
- A SNP is considered representative of a gene if localized within gene boundaries extended by 15kb upstream and downstream
- Annotation of the Refseq using IPA => 7,177 different genes
- Identification of 42 candidate genes
- Functional analysis using IPA
- Network analysis using IPA and functional annotation of each significant network
 - Each network contains at most 35 molecules



Exemple of the West-African cattle breeds : annotation using systems biology tools

A. Network N (obtained merging N1, N2 et N4)

22 genes under selection

B. Network N3 9 genes under selection





INGENUIT

PATHWAY ANALYSIS



Exemple of the West-African cattle breeds :

annotation using systems biology tools

Gautier *et al*, 2009, BMC Genomics Flori *et al*, 2009, PLoS One Flori *et al*, 2014, Mol. Ecol



Climatic conditions

Drought and food shortage

Breeders'choices (coat color, horn)

Pathogens



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Annotation using systems biology tools The example of the European bison

- Wisent: the largest European herbivore, emblematic of the continent wildlife
- Identification of selection footprints between

the European bison and cattle

- Using individual whole-genome sequences (10X, 100nt paired-end)
- Calculating K_a/K_s (Kimura, 1983; *KaKs calculator*, Zhang et al, 2006):
 - > the K_a/K_s ratio is an indicator of selective pressure acting on a protein-coding gene.
 - i.e. ratio of the number of non-synonymous substitutions per non-synonymous site (K_a) to the number of synonymous substitution per synonymous site (K_s), in a given period of time.
- Genes with a K_a/K_s ratio above 1 are evolving under positive selection



Gautier et al, under revision





Annotation using systems biology tools The example of the European bison

- 873 transcripts under selection => 450 genes ready for functional and network analyses
- **70% of the genes participated to a global network** corresponding to six significant interconnected networks (by up to 12 common molecules)








Functional analysis

Main functional categories	Diseases and Biological Functions	p-value	Number of molecules
Physiological System Development and Function	Nervous System Development and Function	1.26x10 ⁻¹⁶ - 3.05x10 ⁻⁰²	54
	Hematological System Development and Function	1.88x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	39
	Immune Cell Trafficking	1.88x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	34
	Embryonic development	2.07x10 ⁻⁰³ - 3.05x10 ⁻⁰²	15
	Organ development	2.07x10 ⁻⁰³ - 3.05x10 ⁻⁰²	11
Diseases and Disorders	Inflammatory Response	1.88x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	35
	Infectious Disease	2.70x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	14
	Connective Tissue Disordes	3.29x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	12
	Organismal Injury and Abnormalities	3.29x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	208
	Skeletal and Muscular Disorders	3.29x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	11
Molecular and Cellular Functions	Cell Death and Survival	2.96x10 ⁻⁰⁵ - 3.05x10 ⁻⁰²	33
	Cellular Compromise	2.96x10 ⁻⁰⁵ - 3.05x10 ⁻⁰²	24
	Cell Morphology	1.10x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	22
	Cellular Assembly and Organization	1.10x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	17
	Cell-To-Cell Signaling and Interaction	1.17x10 ⁻⁰⁴ - 3.05x10 ⁻⁰²	94

Most significant category: Nervous System Development and Function

INGENUITY PATHWAY ANALYSIS





 Many genes are related to several functions

=>pleiotropic role

Some genes underlie obvious distinctive features between wisent and cattle







 Some genes underlie obvious distinctive features between wisent and cattle

Key functions

Presumed adaptation

- Hair development and thermogenesis
- Olfactory and taste receptor genes
- > Genes involved in immune reponse
- Genes involved in lipid metabolism and mammary gland development
- Genes involved in key processes of nervous system and in several by-product phenotypes of domestication

-> to food recourses (Earest babitet

=> to cold climatic conditions (wisent)

- => to different pathogen exposures (wildlife/domestication)
- => artificial selection in cattle (dairy traits)
- =>to domestication/wildlife



- Numerous genes involved in the domestication syndrom
- Main modified traits in domestication syndrom
 - Docility

Selection

By-products phenotypes of domestication

- Juvenile behavior
 - Depigmentation (white patches)
 - Floppy and reduced ears
 - Reduced muzzle and jaws
 - Smaller teeth
 - Shape of the tail (curly)
 - More frequent oestrus



Belyaev and Trut, 1989; Trut et al, 2009



• Our analysis give an empirical support to the unified explanation of the domestication syndrome in mammals proposed by Wilkins et al, 2014





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Association with covariates The example of the French cattle breeds

Gautier and Flori, in prep

Identification of regions under selection associated with dairy traits

- Dairy traits: average milk production, lactation length, fat and protein content
- 13 breeds
- 50K SNP chip
- Baypass (Gautier, 2015):
 - > Auxiliary covariate model

Correlation plot based on Ω





Association with covariates The example of the French cattle breeds

Identification of regions under selection associated with dairy traits





Association with covariates The example of the French cattle breeds

Identification of regions under selection associated with dairy traits

- Using Refgene file from UMD3.1 assembly on ucsc (14627 Refseq ID)
- Annotation of SNPs using the Refgene file: a SNP is representative of a gene if it is located within gene boundaries +/- 15kb
 =>94492 single ID
- BFmc>20 =>decisive evidence
 - » Milk production: NUDC3, LAP3, LCORL
 - » Fat content: 24 RefSeq
 - » Protein content 5 Refseq
- Choice of a less stringent criteria would be more informative (e.g. BFmc>10)





Conclusion

- Functional and network analysis might be powerful to find a biological interpretation of footprints of selection
- Taking special care to avoid storytelling
 - Criteria to define a candidate region and candidate genes (e.g. significant threshold, nb of significant SNP/gene or region, distance from peak)
 - » If too stringent criteria => loss of information
 - Integrating other information
 - » Sequencing data=>exhaustive list of variants that can be annotated and allelic frequencies
 - » QTL information
 - » Using association methods with population environmental covariates or phenotypes





Some references

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- Khatri et al, PLoS Computational biology, 2012. Ten years of pathway analysis: current approaches and outstanding challenges
- **Pavlidis et al, MBE, 2012.** A critical assessment of storytelling: Gene ontology categories and the importance of validating genomic scans.

