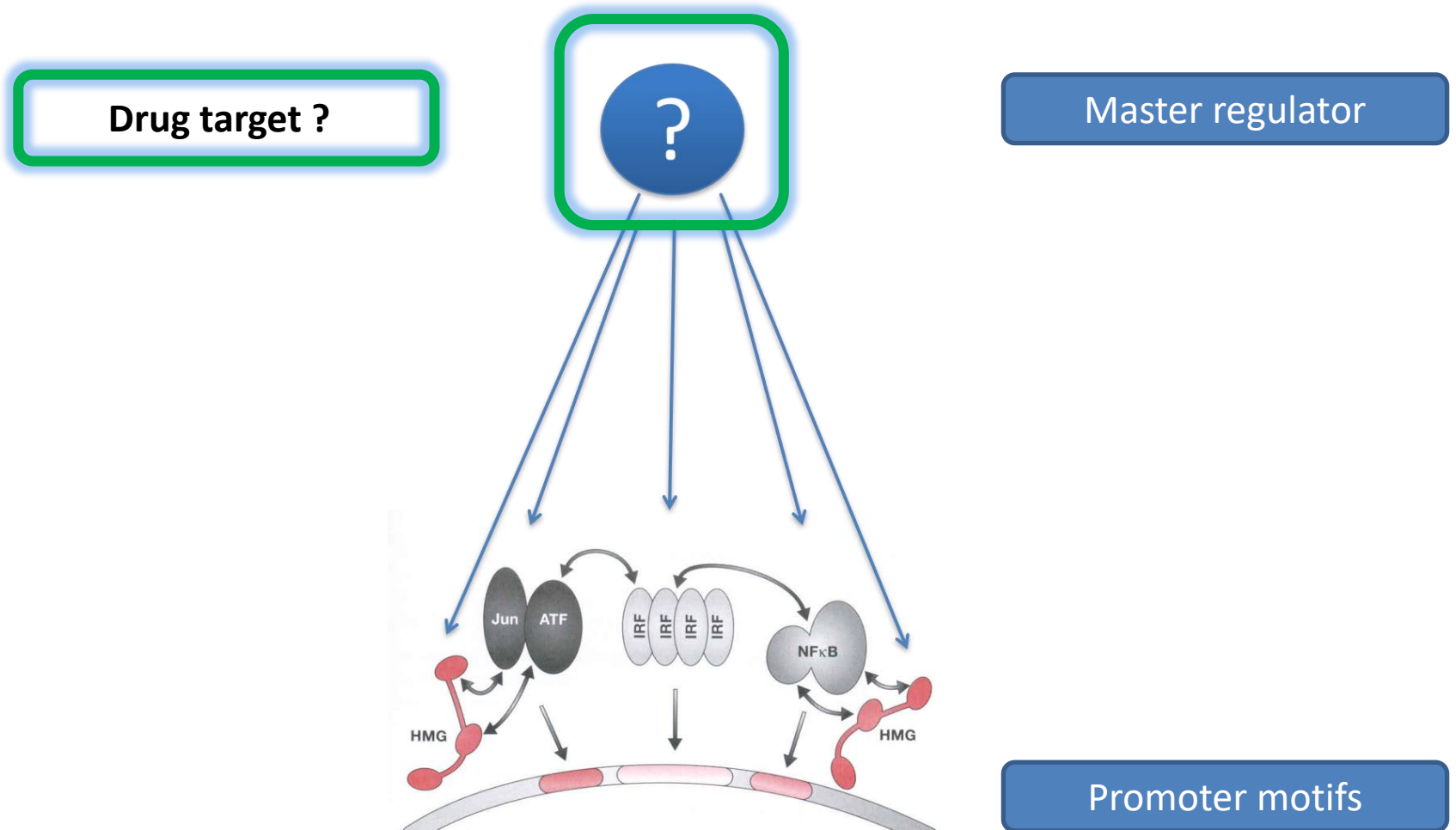


# Targets

Day 7

# Upstream analysis – From motifs in promoters to drug targets

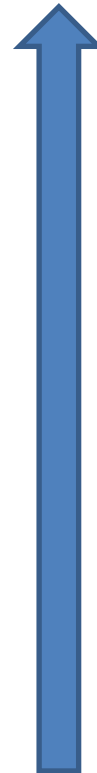


# Upstream Analysis Concept

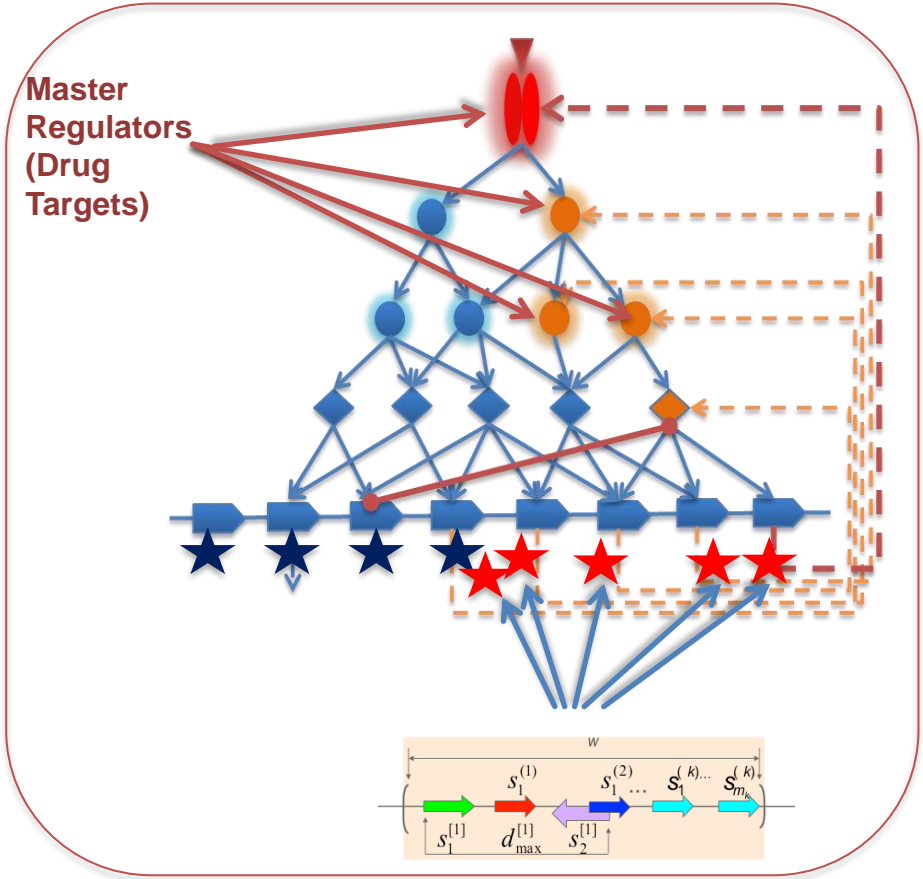
The concept of upstream analysis includes two steps, promoter analysis and network analysis.

Promoter analysis is done as TFBS search using proprietary F-Match and CMA algorithms together with the unique collection of [TRANSFAC](#)<sup>®</sup> matrices.

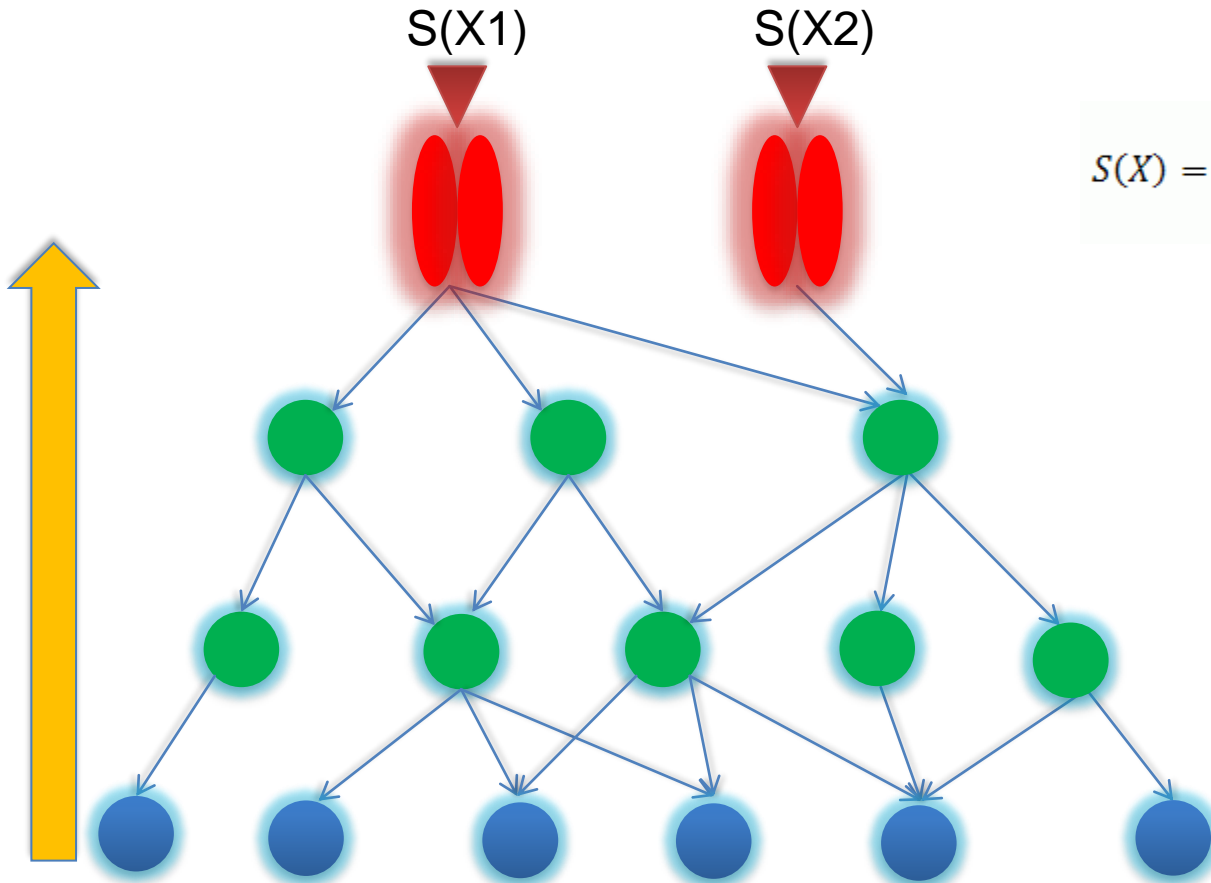
Network analysis is done with a proprietary graph-analyzing algorithm and the unique collection of [TRANSPATH](#)<sup>®</sup> reactions. As a result, the master regulators in the networks are identified.



Upstream



# Search for master regulators



$$S(X) = \sum_{r=1}^R \frac{M(X,r)}{M_{max}(r)} \cdot \frac{1}{1 + pN(X,r)/N_{max}(r)}$$

Where:

**R** - Max radius (input parameter)

**p** - Penalty (input parameter)

**N(X,r)** - total number of molecules reachable from key molecule X within the radius r.

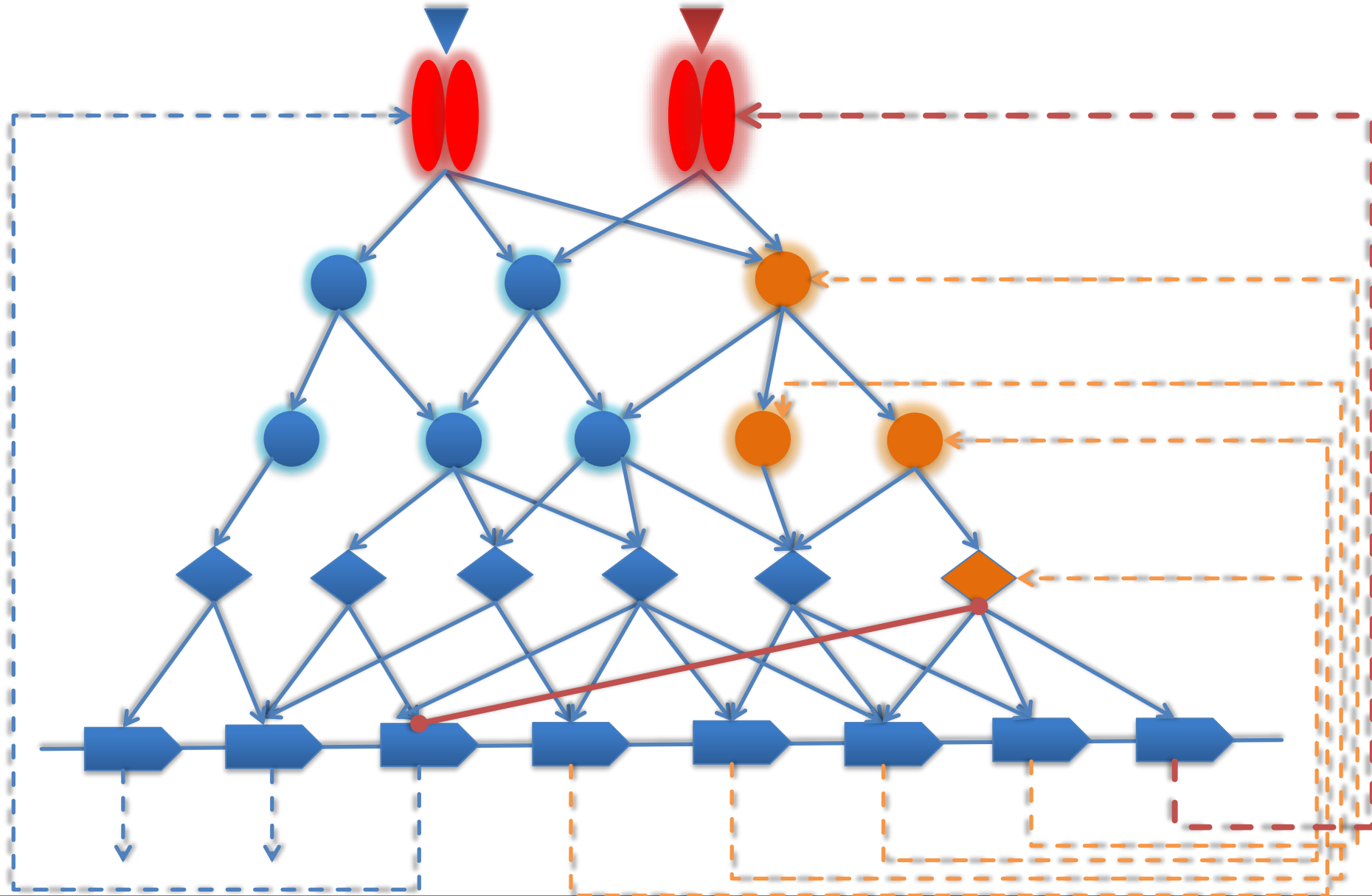
**N<sub>max</sub>(r)** - maximal value of N(X,r) over all key molecules X found for this radius.

**M(X,r)** - sum of w(X) for all hits reachable from key molecule X within the radius r, where w(X) - weight of hit X.

**M<sub>max</sub>(r)** - maximal value of M(X,r) over all key molecules X found for this radius.

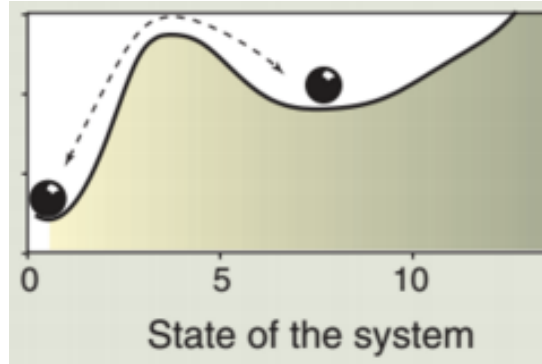
Kel, A., Voss, N., Jauregui, R., Kel-Margoulis, O. and Wingender, E.: Beyond microarrays: Find key transcription factors controlling signal transduction pathways *BMC Bioinformatics* 7(Suppl. 2), S13 (2006). [Link](#)

# Walking pathways

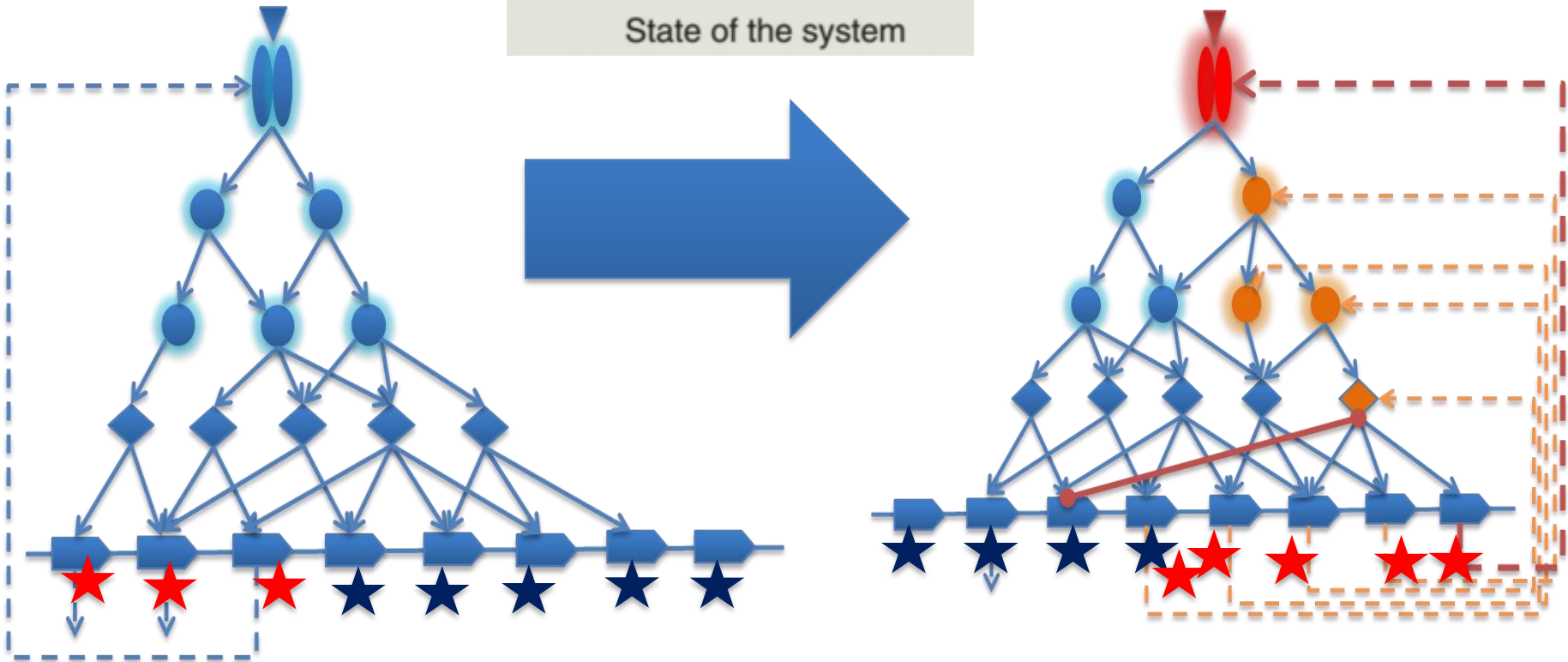


# „Walking pathways“

## Healthy



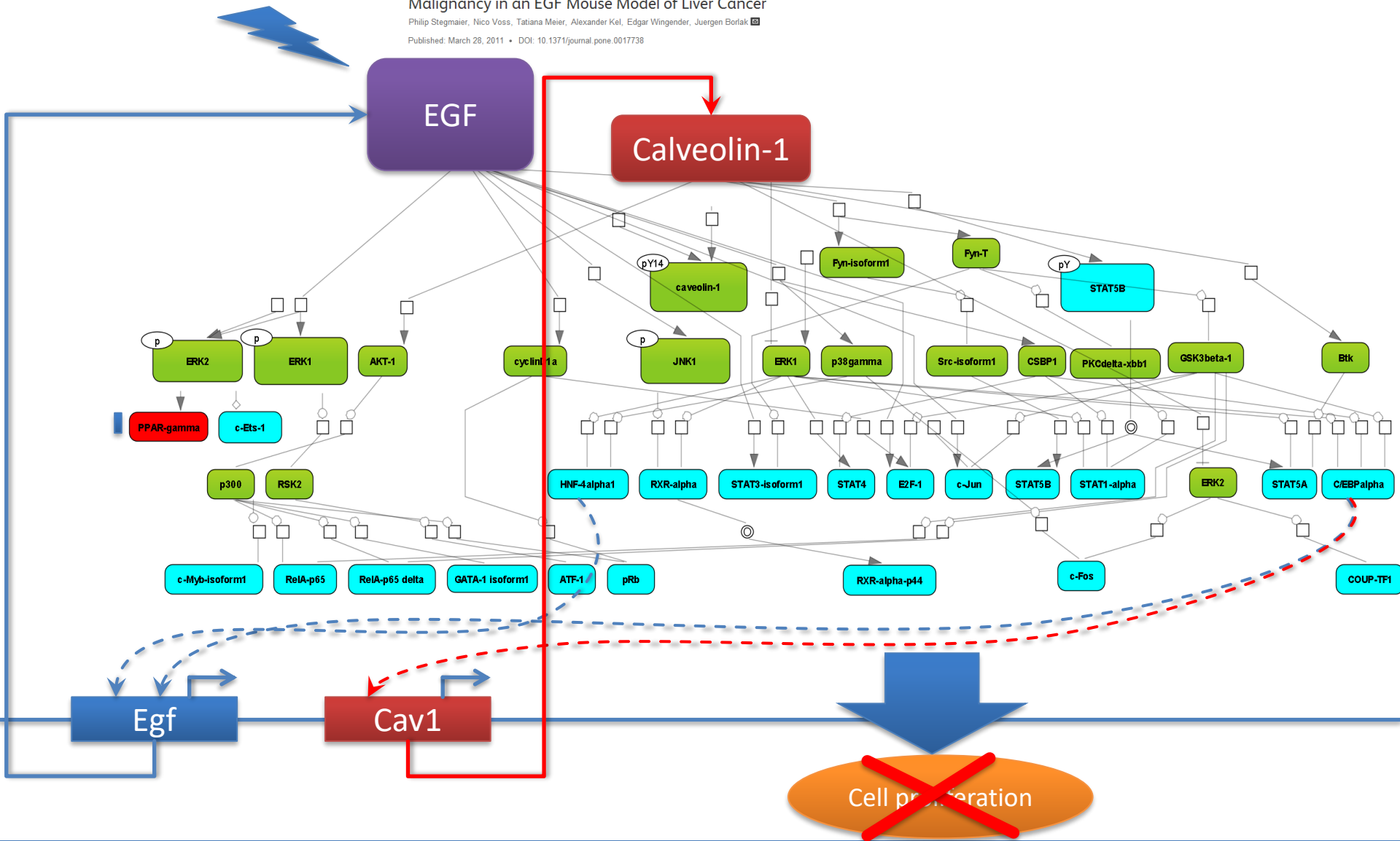
## Disease



# Advanced Computational Biology Methods Identify Molecular Switches for Malignancy in an EGF Mouse Model of Liver Cancer

Philip Stegmaier, Nico Voss, Tatiana Meier, Alexander Kel, Edgar Wingender, Juergen Borlak

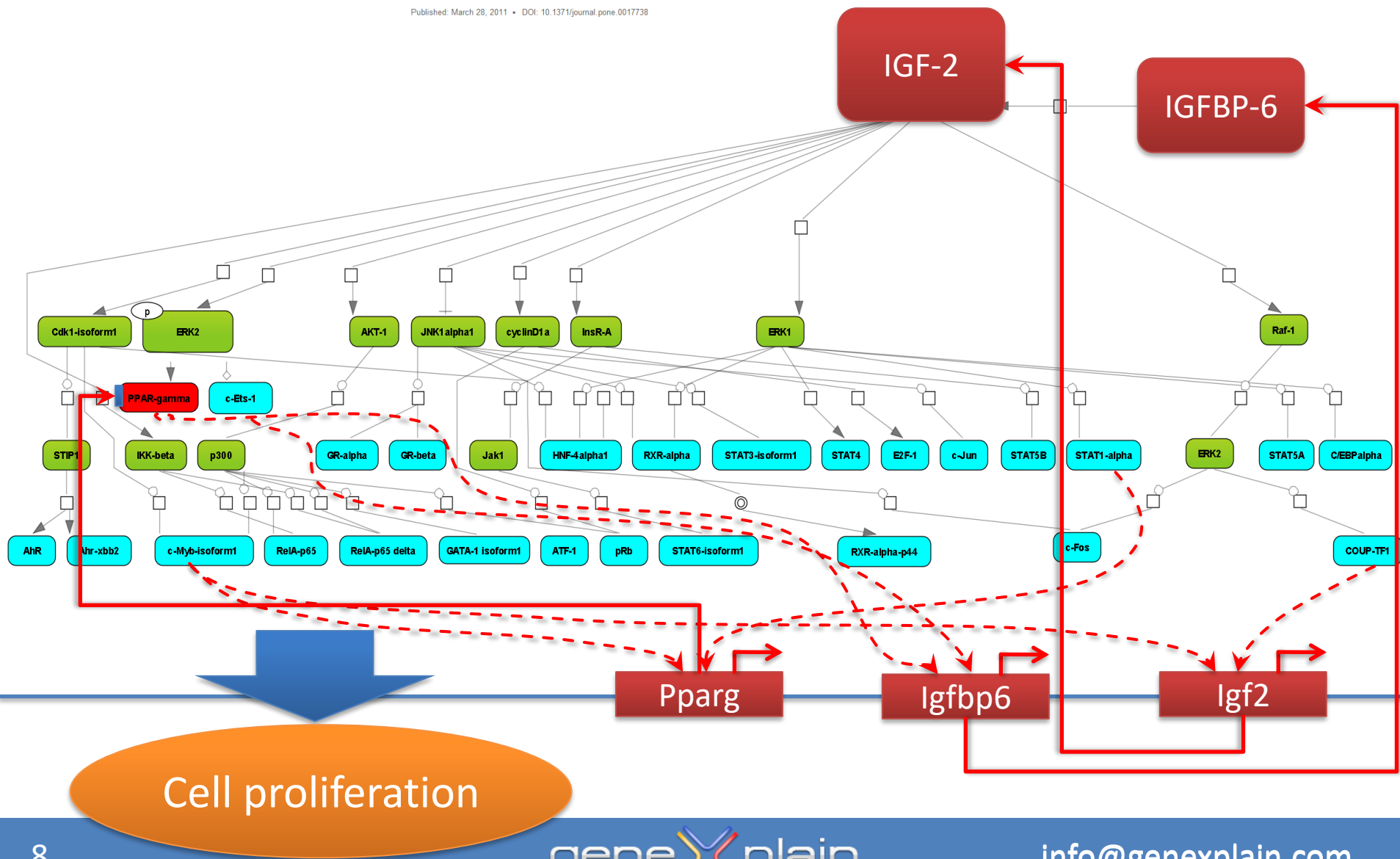
Published: March 28, 2011 • DOI: 10.1371/journal.pone.0017738



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Published: March 28, 2011 • DOI: 10.1371/journal.pone.0017738





# Introduction into Genome Enhancer

RSCI training

Day 7

# Genome Enhancer

Welcome to the new era of precision medicine

Release 2.2

Please sign in

Sign in

Register

[I forgot my password](#)

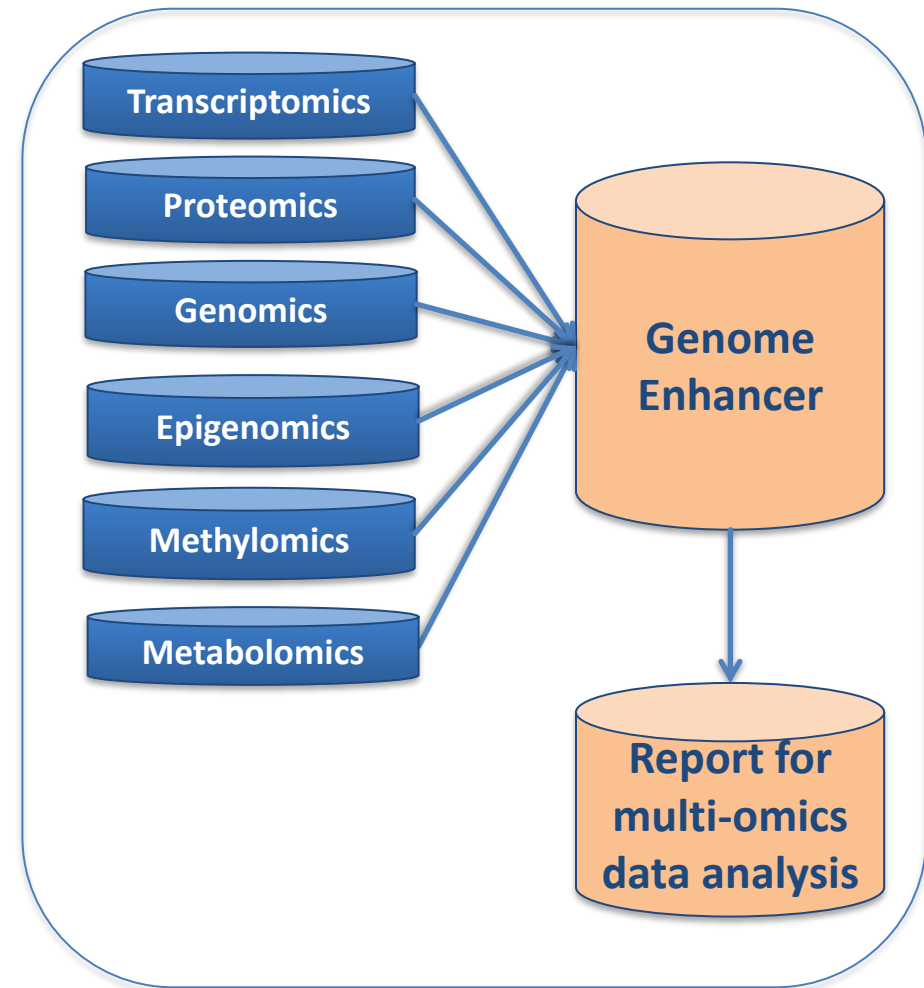
<https://ge.genexplain.com/>

## Multi-omics data analysis

GE provides a unique possibility for multi-omics analysis.

GE accepts input of several different human omics data at the same time to be analyzed together.

The report is generated for multi-omics data analysis.



# Input data formats

As input, GE accepts a wide range of data formats, starting with different raw formats

<https://genexplain.com/genome-enhancer/#section3>

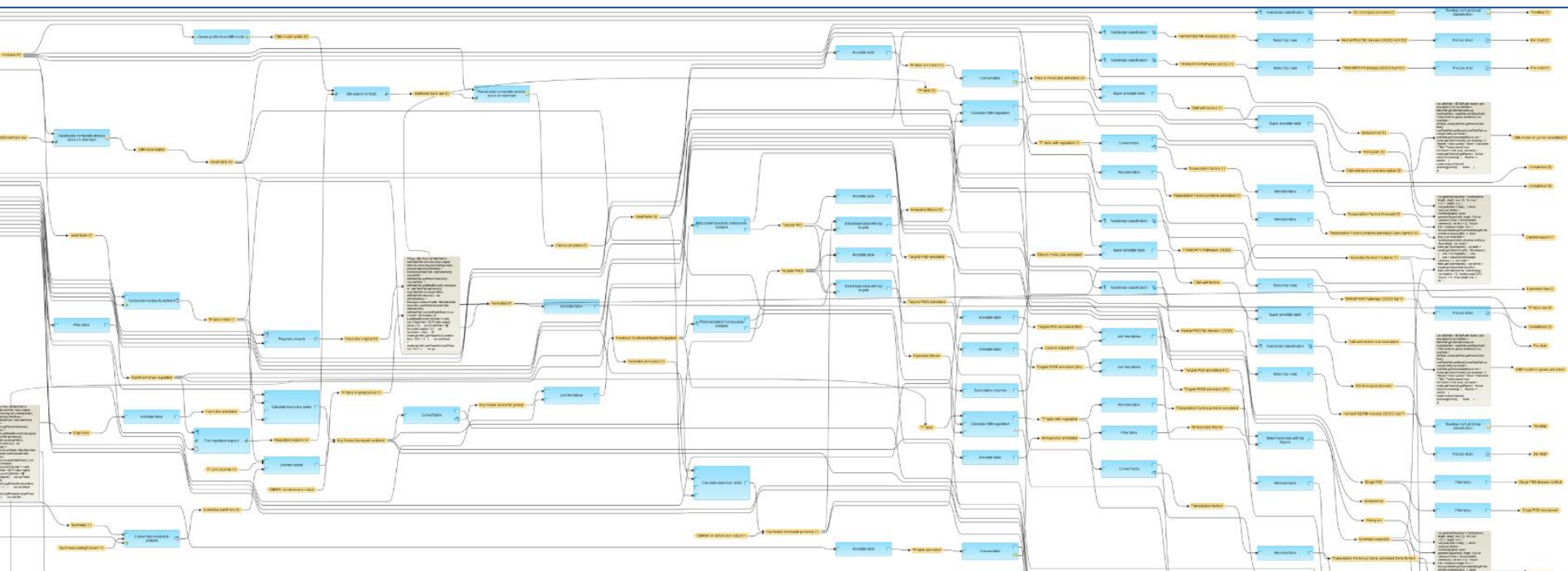
and going to pre-calculated data as Excel tables, csv, and other tabulated formats.

Researchers are welcome to import the data formats they already have in their hands.

<b>Transcriptomics (RNA-seq, microarrays)</b>	<b>Genomics</b>
*.txt, *.csv, *.xls (table with gene identifiers)	*.vcf
*.CEL (affymetrix)	*.txt, *.csv, *.xls (table data with SNP identifiers, rs*)
*.txt (special agilent format)	*.fastq
*.txt (special illumina format)	<b>Proteomics</b>
*.fastq	*.txt, *.csv, *.xls (table with protein identifiers)
<b>Epigenomics (ChIP-seq)</b>	<b>Metabolomics</b>
*.fastq	*.txt, *.csv, *.xls (table with the list of metabolites from chebi database, e.g. CHEBI:57316)
*.bam (hg38 only)	<b>Files of one data format can be uploaded in a .zip archive</b>
*.bed (hg38 only)	
*.txt (table with illumina methylation probe ids, cg*)	

# Automatically generated workflow (pipeline)

Depending on the submitted omics data and file format, GE automatically generates a pipeline. An example illustrates a fragment of a pipeline for transcriptomics data analysis.

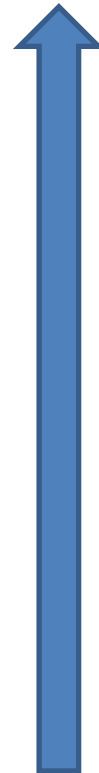


# Upstream Analysis Concept

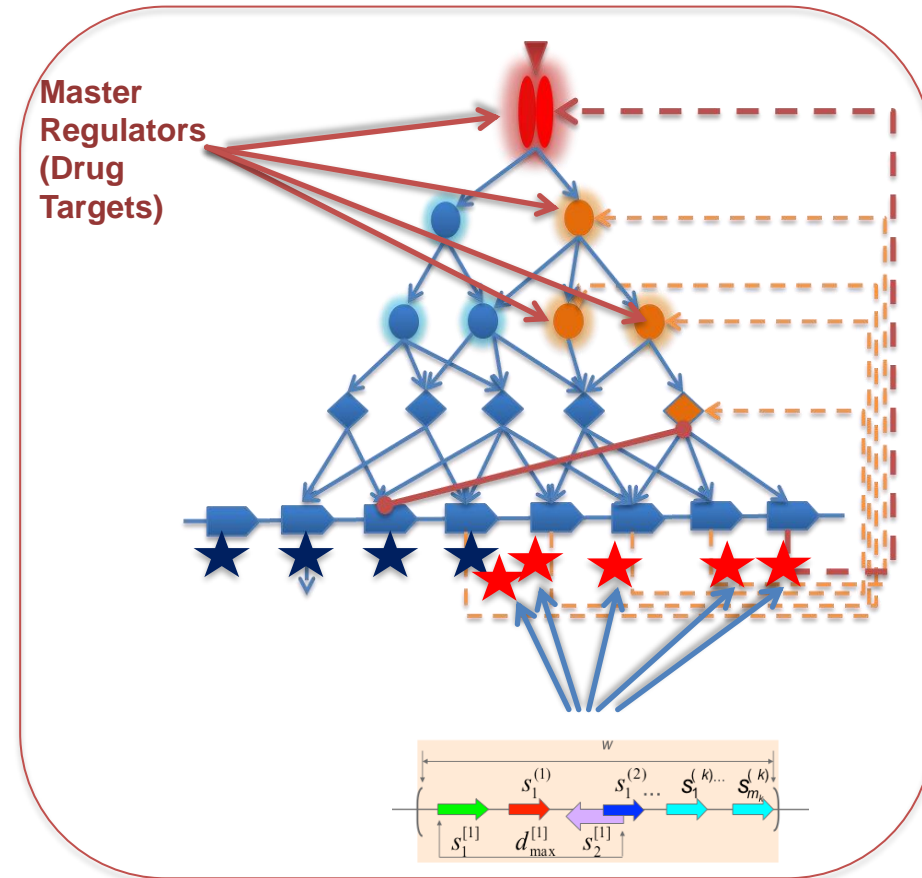
The concept of upstream analysis includes two steps, promoter analysis and network analysis.

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Network analysis is done with a proprietary graph-analyzing algorithm and the unique collection of [TRANSPATH](#)<sup>®</sup> reactions. As a result, the master regulators in the networks are identified.



Upstream



# Genomic variations in gene regulatory regions (regulatory SNPs)

GE includes a unique algorithm for analysis of genomic variations in gene regulatory regions, also known as regulatory SNPs.

Changes of TFBS scores resulting from particular genomic variations are part of the report generated by GE.

Table 3. Mutations revealed in genes carrying SNP variations

[See full table →](#)

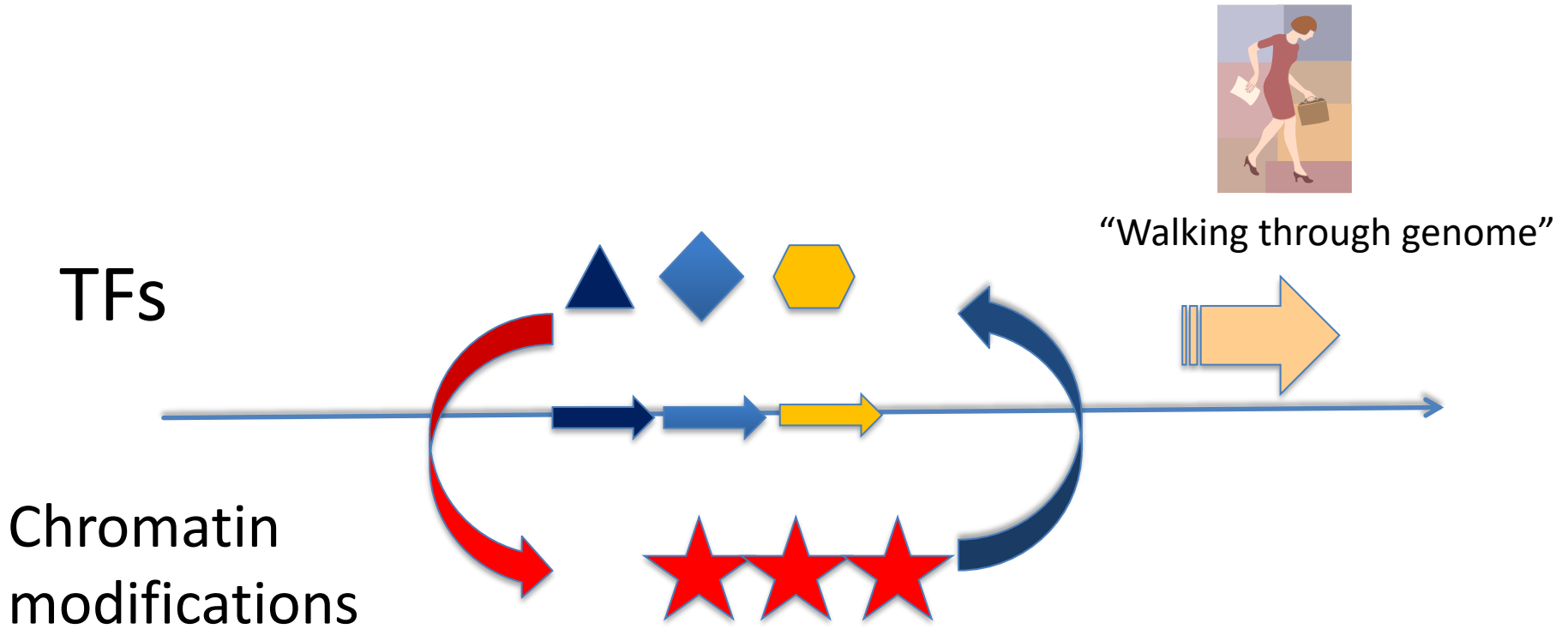
ID	Gene symbol	Gene schematic representation	Number of variations
ENSG00000196735	HLA-DQA1		81
ENSG00000130164	LDLR		46
ENSG00000161888	SPC24		30
ENSG00000165029	ABCA1		30
ENSG00000128918	ALDH1A2		26
ENSG00000166035	LIPC		24
ENSG00000169174	PCSK9		20
ENSG00000136573	BLK		19
ENSG00000175445	LPL		18
ENSG00000084674	APOB		16

Table 4. PWMs whose sites were lost or gained due to mutations in genes carrying SNP variations

[See full table →](#)

ID	P-value (gains)	P-value (losses)	yesCount (gains)	yesCount (losses)
V\$BTEB2_Q3_01	1.79E-11	1.01E-12	14	21
V\$SNAI2_01	1.79E-11	2.11E-12	14	21
V\$STAT4_01	3.37E-14	6.82E-13	17	24
V\$TFE_Q6	3.37E-14	1.26E-13	17	25
V\$AML1_Q4	2.62E-16	5.58E-12	20	22
V\$AP1_Q2	2.62E-16	5.58E-12	20	22
V\$CETS1_Q6	2.62E-16	5.58E-12	20	22
V\$LEF1_Q5	2.62E-16	3.6E-13	20	24
V\$LEF1_Q5_01	2.62E-16	3.6E-13	20	24
V\$AHR_Q6	1.3E-16	3.32E-11	21	22
V\$AML1_Q1	1.3E-16	3.32E-11	21	22

# Epigenetic “walking”





# Druggability of the Master Regulators (drug targets)

GE calculates, in a unique way, and provides a report section for the druggability of the identified master regulators (drug targets).

Additionally, the clinical trials for the suggested drugs are included in the report as well.

These data are included based on [HumanPSD](#) database.

ID	Gene symbol	Gene description	Druggability score
<a href="#">ENSG00000125538</a>	IL1B	interleukin 1 beta	11
<a href="#">ENSG00000097007</a>	ABL1	ABL proto-oncogene 1, non-receptor tyrosine kinase	8
<a href="#">ENSG00000109320</a>	NFKB1	nuclear factor kappa B subunit 1	8
<a href="#">ENSG00000136244</a>	IL6	interleukin 6	6
<a href="#">ENSG00000197122</a>	SRC	SRC proto-oncogene, non-receptor tyrosine kinase	6
<a href="#">ENSG00000259207</a>	ITGB3	integrin subunit beta 3	4
<a href="#">ENSG00000112062</a>	MAPK14	mitogen-activated protein kinase 14	3
<a href="#">ENSG00000140992</a>	PDPK1	3-phosphoinositide dependent protein kinase 1	3
<a href="#">ENSG00000143322</a>	ABL2	ABL proto-oncogene 2, non-receptor tyrosine kinase	3
<a href="#">ENSG00000103653</a>	CSK	c-src tyrosine kinase	2

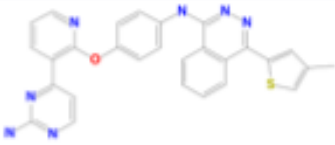
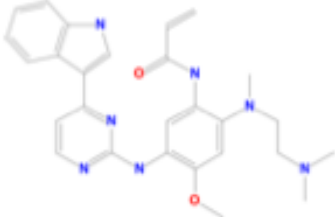
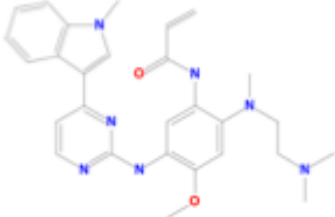
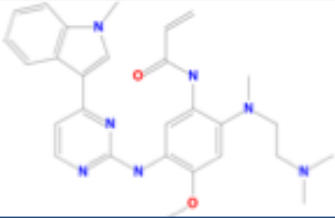
ID	Name	Target names	Target activity score	NA	Phase 1	Phase 2	Phase 3	Phase 4
<a href="#">DB00482</a>	Celecoxib	PDPK1	1	Inflammation, Adenoma, Adenomatous Polyposis Coli, Adenomatous Polyps, Alzheimer Disease, Antiphospholipid Syndrome, Arthritis...	Inflammation, Apnea, Brain Abscess, Breast Neoplasms, Carcinoma, Carcinoma, Non-Small-Cell Lung, Carcinoma, Small Cell...	Adenocarcinoma, Adenoma, Adenomatous Polyposis Coli, Adenomatous Polyps, Amyotrophic Lateral Sclerosis, Angiomyoma, Apnea...	Adenoma, Adenomatous Polyposis Coli, Adenomatous Polyps, Alzheimer Disease, Anterior Cruciate Ligament Injuries, Arthritis, Gouty...	Inflammation, Inflammatory Bowel Diseases, Adenoma, Adenomatous Polyposis Coli, Adenomatous Polyps, Ankle Injuries, Arteriosclerosis...
<a href="#">DB01017</a>	Minocycline	IL1B	1	Acne Vulgaris, Acute Kidney Injury, Alopecia, Angelman Syndrome, Anxiety,	Inflammation, Acne Vulgaris, Acute Kidney Injury, Affect, Alcohol Drinking...	Inflammation, Acne Vulgaris, Alcohol Drinking, Alopecia, Alzheimer Disease	Acne Vulgaris, Affect, Alopecia, Amphetamine-Related Disorders, Amyotrophic	Acne Vulgaris, Affect, Alopecia, Autistic Disorder, Bacterial Infections

## Calculations done by the cheminformatics software PASS are part of the report

GE report includes suggestions for drug re-purposing (off label drug applications), and even for new chemical substances, to target identified master regulators.

This part is based on the calculations done with the Cheminformatics software [PASS](#).

Integration of bioinformatics and Cheminformatics approaches is the next specific and unique GE feature.

Name	Structure	Target names	Target activity score
AMG 900		ADAM17, MAP4K4, SETD7, PINK1, ABL1, PTPN6, MAPK14...	11.67
AZ5104		ADAM17, MAP4K4, PINK1, ABL1, MAPK14, IRAK2, SGK1...	8.52
Osimertinib		ADAM17, MAP4K4, PINK1, ABL1, MAPK14, IRAK2, SGK1...	7.83
AZD9291		ADAM17, MAP4K4, PINK1, ABL1, MAPK14, IRAK2, SGK1...	7.83

# The Report is generated automatically and can be downloaded

As a result of GE run, a report is automatically generated, which can be downloaded in a comfortable format, as PDF file or HTML page.

All primary results, including all intermediate results, are available as tables or graphs, depending on data type, and can be downloaded in a number of formats.

## Sequence and Pathway analysis

### PLK1 and TRIM22 are promising druggable targets for treating colorectal neoplasms that control activity of TP53, AR and ESR2 transcription factor on promoters of genes carrying sequence variations in colon tissue

Demo User  
geneXplain GmbH  
info@genexplain.com  
Data received on 24/01/2020 ; Run on 20/02/2020 ; Report generated on 20/02/2020  
Genome Enhancer release 1.9 (TRANSFAC®, TRANSPATH® and



#### Abstract

In the present study we applied the software package "GenoTissue". The study is done in the context of colorectal neoplasms network that governs the studied pathological process. In the genes activities in the pathological state. The activities of these second step of analysis. After a subsequent druggability check for the analyzed pathology. At the end the pipeline comes to compounds with the potential to interact with selected drug targets.

From the data set analyzed in this study, we found the following variations: TP53, AR and ESR2. The subsequent network analysis and druggable molecular targets. Finally, the following drugs Coenzyme A, 4-(4-METHYLPYPERAZIN-1-YL)-N-[5-(2-THIENYL)RING OF Pentamannosyl 6-Phosphate, Adenosine-5'-Monophosphate

#### 1. Introduction

Recording "omics" data to measure gene activities, protein expression, metabolites, etc. in the pathological state of an affected organism or tissue. Increasingly large "multiomics" datasets. Still the challenge remains how to state different from the norm. The disease-causing mechanism as a result of a genetic or epigenetic alterations influencing the networks can help identify potential master regulators of the point to ways how to block a pathological regulatory cascade. Stop the pathological process and cure the disease.

Conventional approaches of statistical "omics" data analysis phenomena and therefore contribute little to the understanding method [1-4] applied here has been devised to provide a comprehensive overview of the process and its regulation. It comprises two major steps: (1) analysing promoters and enhancers involved in their regulation and, thus, important for the process and identifying master regulators at the top of such pathways. TF binding site identification algorithms Match [7] and ChIA [8] and special graph search algorithms [10] implemented in the software.

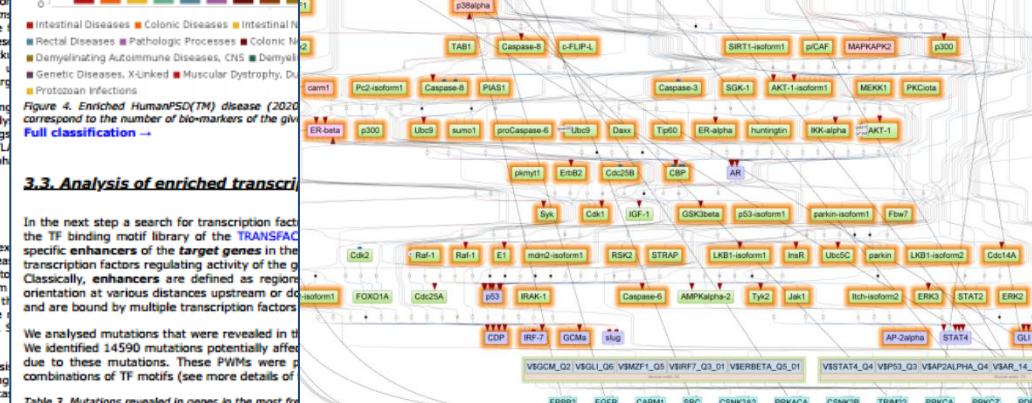
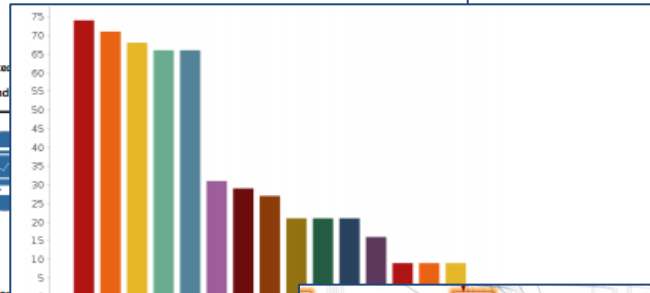


Figure 4. Enriched HumanP50(TM) disease (2020) correspond to the number of bio-markers of the given transcription factors.

#### 3.3. Analysis of enriched transcription factors

In the next step a search for transcription factors binding to the promoters of the target genes in the transcription factors regulating activity of the target genes. Classically, enhancers are defined as regions of DNA that are bound by multiple transcription factors.

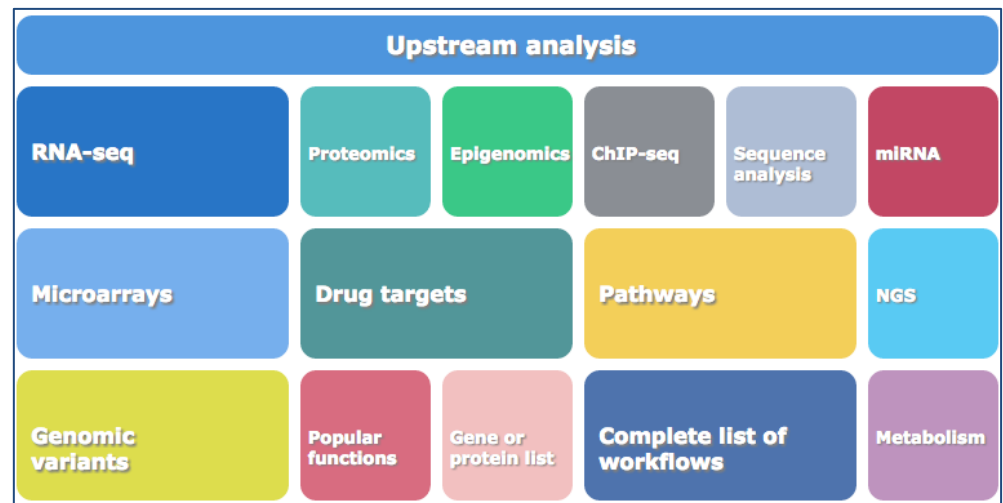
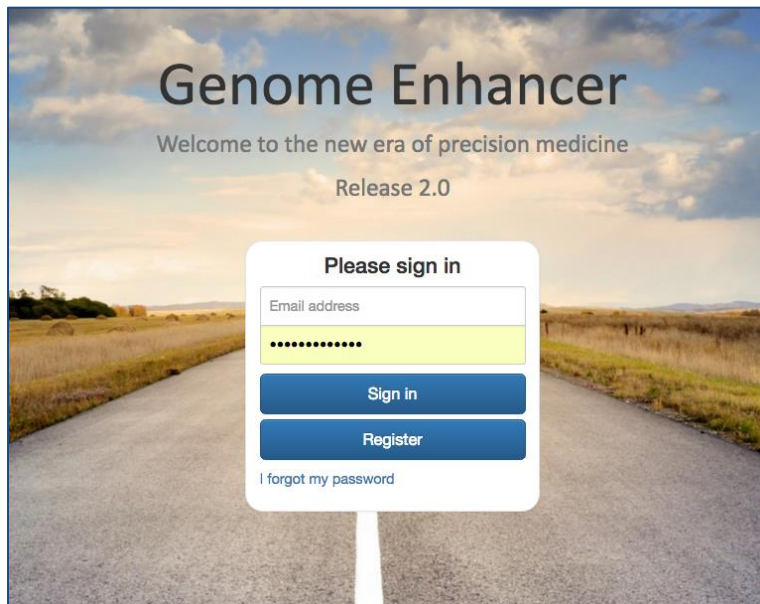
We analysed mutations that were revealed in the target genes. We identified 14590 mutations potentially affected due to these mutations. These PWMs were compared with combinations of TF motifs (see more details of the analysis in the full report).

Table 3. Mutations revealed in genes in the most frequent TF motifs.

ID	Gene symbol	Gene schematic representation	Number of variations
ENSG00000132570	PCBE2		172
ENSG00000242086	LINC00969		147
ENSG00000248923	MTND5P11		126
ENSG00000234745	HLA-B		122
ENSG00000154237	LRRK1		117
ENSG00000259755	RP11-505E24.2		111
ENSG00000230021	RFS-857K21.4		104
ENSG00000067057	PFKP		92
ENSG00000247627	MTND4P12		91
ENSG00000281344	HELLPAR		88

# Genome Enhancer Expert

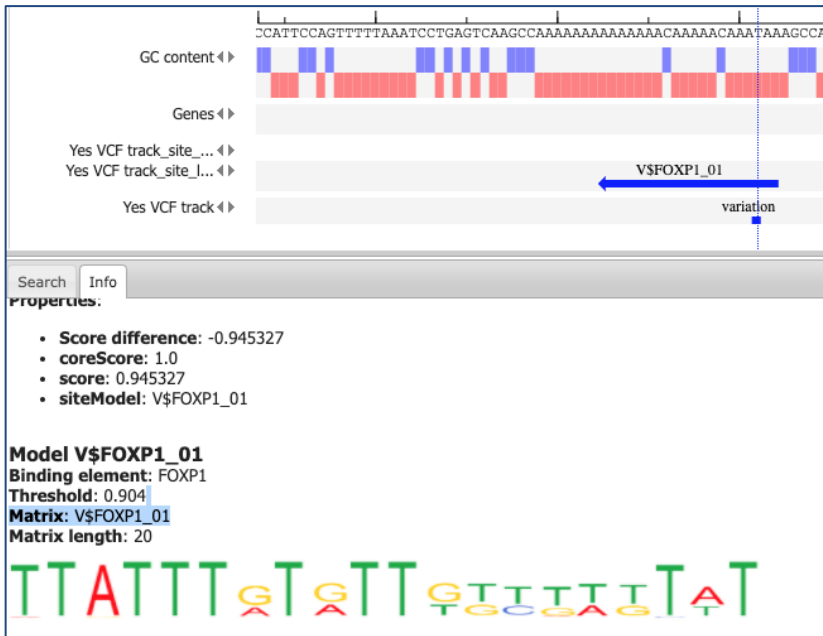
**Synergism between the automatic pipeline  
of Genome Enhancer and  
a comprehensive bioinformatics toolbox  
of the geneXplain platform**



# Genome Enhancer Expert

## More functionalities for the Genome Enhancer users:

- ✓ Post-processing the results of the automatic GE pipeline
- ✓ Integrated databases and analysis tools
- ✓ Ready-made workflows for an easy start
- ✓ Genomes for several different organisms
- ✓ Java API access
- ✓ Simulation engine inside



Databases Data Analyses

Site analysis

- Apply CMA model to tracks
- Change profile cutoffs
- Cluster track
- Compare TFBS mutations
- Compute profile thresholds
- Construct IPS CisModule
- Construct composite modules
- Construct con...
- Construct con...
- Construct con...
- Construct con...
- Continue CMA

Workflows

- Common
- GTRD
- HumanPSD
- TRANSFAC
  - Analyze SNP list (TRANSFAC(R))
  - Analyze any DNA sequence (TRANSFAC(R))
  - Analyze any DNA sequence for site
  - Analyze promoters (TRANSFAC(R))
  - ChIP-Seq - Identify TF binding sites

Simulation plot

Quantity or concentration

Time