

# A new era in transcription factor research

**Edgar Wingender** 



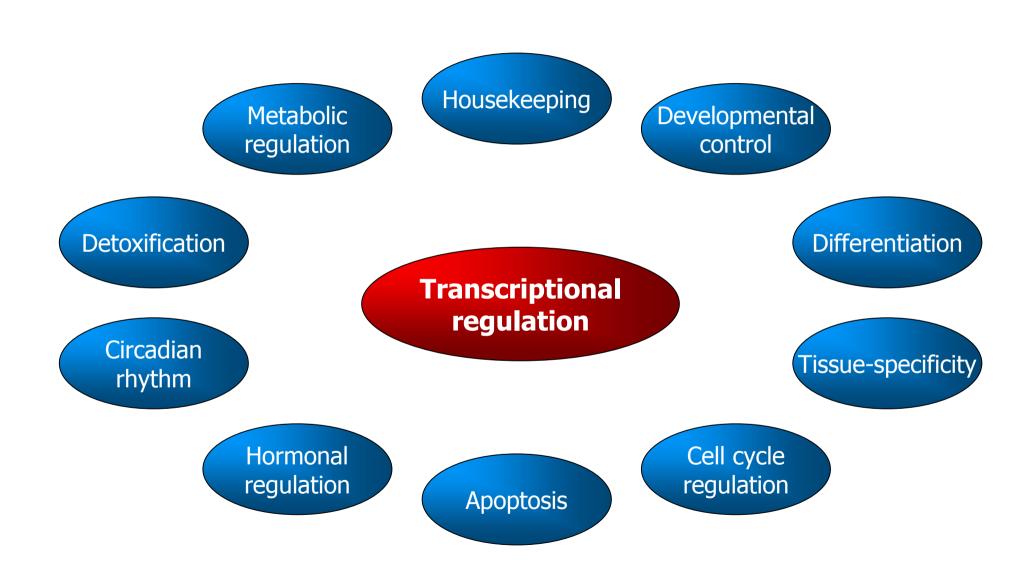
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## **Impact of transcriptional regulation**







## The goals of transcription factor research:

- (1) To understand how the gene-specificity of transcriptional regulation is achieved
- (2) To develop genome-wide maps of transcription factor binding sites (TFBSs)
- (3) To enable prediction of new TFBSs
- (4) To comprehend the complex structure of regulatory genome regions (promoters, enhancers, etc.)
- (5) To predict the DNA-binding specificity of new transcription factors (TFs)
- (6) To construct system-wide transcription networks
- (7) To understand transcriptional dysregulation under disease conditions
- (8) To render transcriptional regulation amenable for targeted alterations



## The goals of transcription factor research:

(1) To understand how the gene-specificity of transcriptional regulation is achieved

**Biology** 

- (2) To develop genome-wide maps of transcription factor binding sites (TFBSs)
- (3) To enable prediction of new TFBSs

**Bioinformatics** 

- (4) To comprehend the complex structure of regulatory genome regions (promoters, enhancers, etc.)
- (5) To predict the DNA-binding specificity of new transcription factors (TFs)
- (6) To construct system-wide transcription networks

**Systems biology** 

- (7) To understand transcriptional dysregulation under disease conditions
- (8) To render transcriptional regulation amenable for targeted alterations

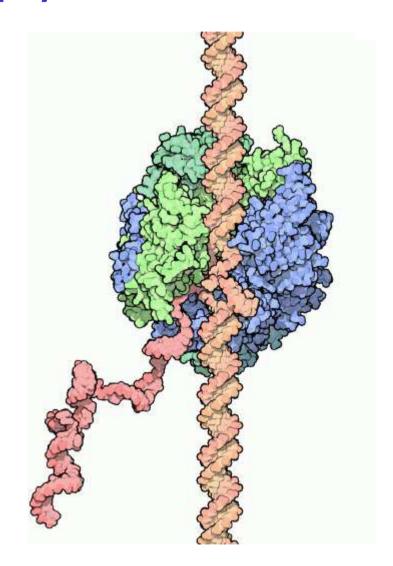
**Synthetic biology** 



## **The DNA-dependent RNA-polymerases**

RNA polymerase II from *S. cerevisiae* 

PDB entry 1I6H





## **The DNA-dependent RNA-polymerases**

In *E. coli*:

RNA polymerase (RNAP): all genes 5 subunits

In Eukaryotes:

RNA polymerase I (A): 45S-rRNA 7-14 subunits

RNA polymerase II (B): mRNA 12 subunits

RNA polymerase III (C): tRNA, 5S-rRNA 10 subunits

RNA polymerase IV: siRNA in plants



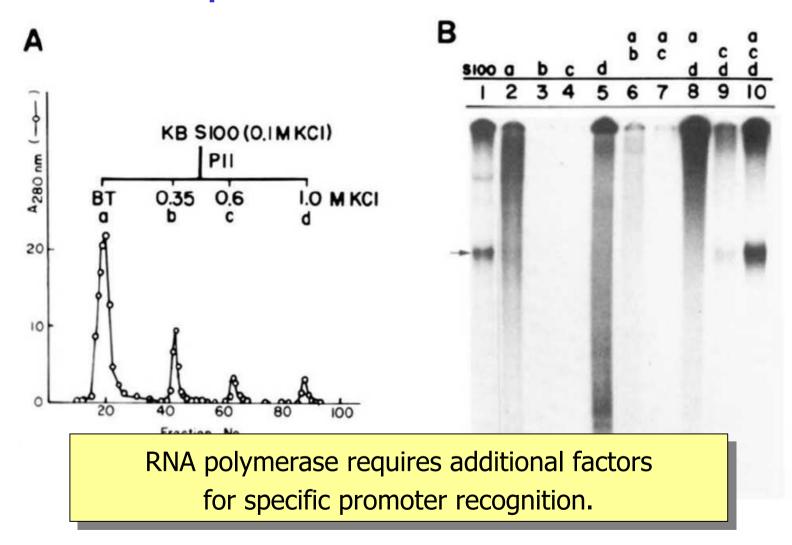
## **The DNA-dependent RNA-polymerases**

## Goal #1:

Understand how the specificity of eukaryotic transcriptional regulation is achieved?



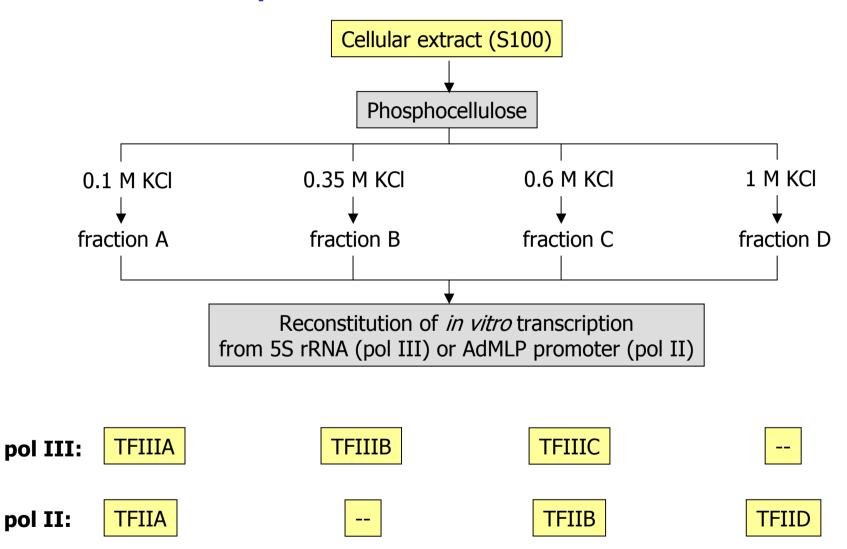
## **General transcription factors**



Multiple factors required for accurate initiation of transcription by purified RNA polymerase II. Matsui T, Segall J, Weil PA, Roeder RG. J Biol Chem. 1980 Dec 25;255(24):11992-6. PMID: 7440580



## **General transcription factors**



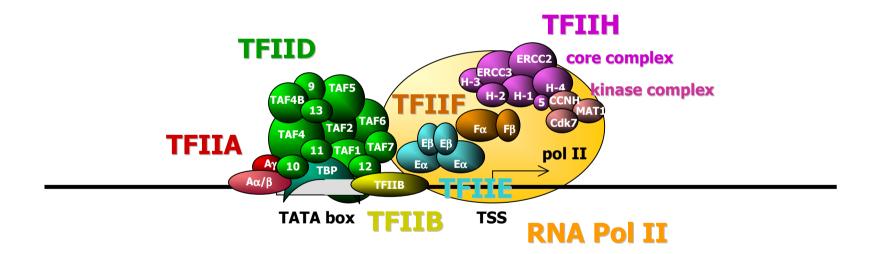
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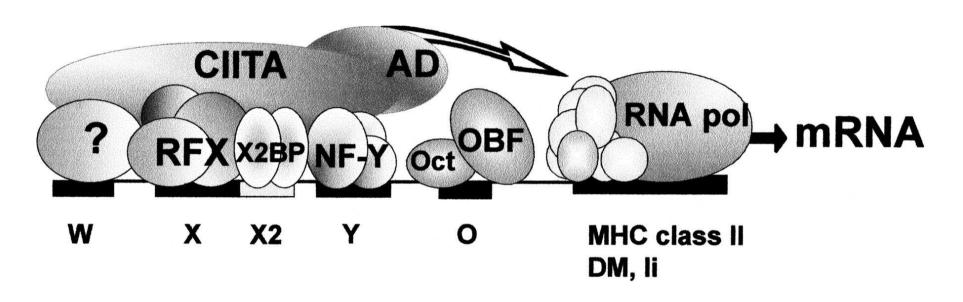
## **General transcription factors**

The pre-initiation transcription complex





#### **Enhanceosome**



In addition to the assembly of general transcription factors, "upstream" factors are required. Many of them are sequencespecific DNA-binding proteins.

Their task is to provide a favorable chromatin structure and/or to facilitate the assembly of the general transcription factors.



#### **Definition**

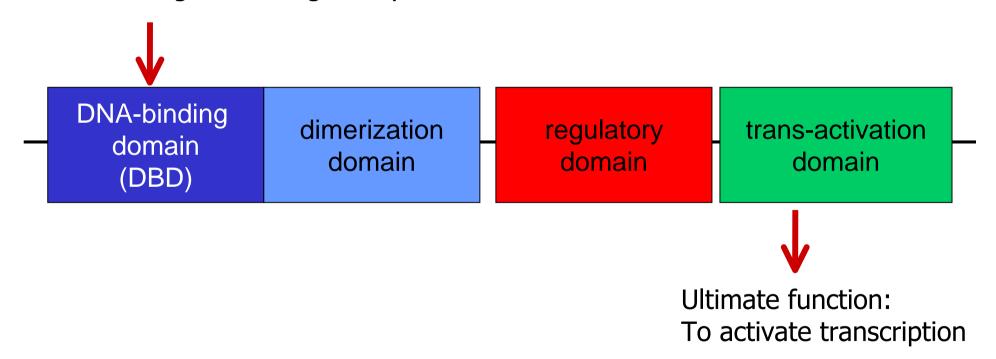
## What is a transcription factor?

A transcription factor is a protein that regulates transcription by specific interaction with DNA or, after nuclear translocation, by stoichiometric interaction with a protein that can be assembled into a sequence-specific DNA-protein complex.



## **Modular structure of a transcription factor**

Primary function:
To recognize *cis*-regulatory elements





## Goal #2:

To generate a comprehensive map of genome-bound proteins.



## The first compilation

Volume 16 Number 5 1988	Nucleic Acids Research
Compilation of transcription regulating proteins	
Edgar Wingender	
Gesellschaft für Biotechnologische Forschung mbH, Mascheroder Weg	I, D-3300 Braunschweig, FRG
Received November 28, 1987; Revised and Accepted January 28, 1988	

Wingender E. Compilation of transcription regulating proteins. Nucleic Acids Res. 1988 Mar 25;16(5):1879-902. PMID: 3282223



## The first compilation

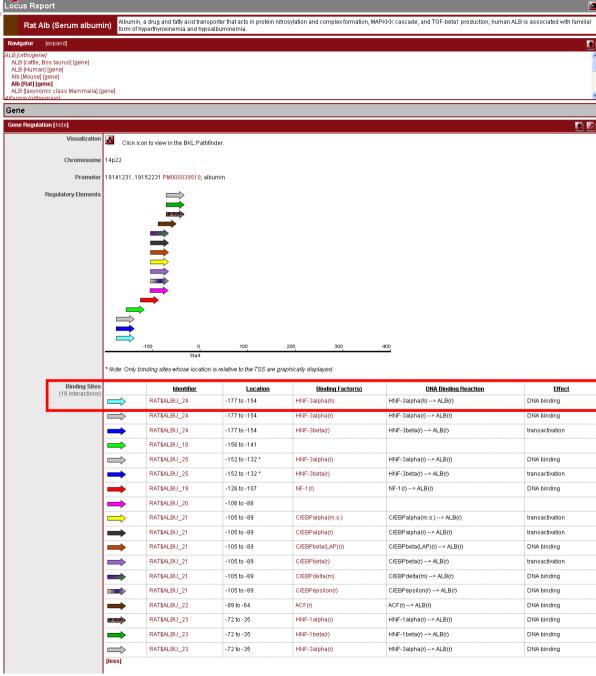
gene, gene product	species/tissue(a)	protein interacting region (b)	ethod (c)	sequence motif	factor bound	ref.
<b>c</b> -actin	chicken/rat myocytes /rat non-myocytes	-83 to -78	3, 4b 3, 4b		myocspec.f. distinct factor	1 1
β-actin	rat/HeLa		3 (compet. with c-fos)	small, dyad symmetry	SRF ?	2
actin SC	Drosophila	-38 to +34	1a	TATAAAA	B factor	3
actin (cytoskeletal)	Xenopus laevis /HeLa	-94 to -75 (SRE)	1a, 4a, 4b	AAGATgcCCAtATtTGGcgATCTT	SRF (?)	4
Ad2MLP (adenovirus 2 major late prom.)	adenovirus/HeLa	-68 to -49 -63 to -52 -50 to -10 -40 to +35 -34 to -22	1a, 3, 4a, 4b 1b 1a 1a 1b	TGTAGGCCACGTGACCGG GGCCACGTGACC TATAAAA TATAAAA TATAAAA	UEF, USF, MLTF USF ? TFIID TFIID	5-10 9 8 9
Adh (alcohol dehydrogenase) —distal prom. —proxim.prom.	Drosophila	- 85 to - 47 -269 to -229 -151 to -105 - 98 to - 77	ia ia ia ia	BCTGCtGCTGCatcCGTCGaCGTCG TACTAA (4x) GCAGCGCTGCCGTCGccggctgaGCAGC GAGATCGCGTAACSGTAGATAA	Adf-1 Adf-1 ?	11 11 11
Adh1 (alcohol dehydrogenase)	maize	-190 to -186 -145 to -138 (after ind.) -120 to -117 (after ind.) -108 to -100	4a (in vivo) 4a (in vivo) 4a (in vivo) 4a (in vivo)	CCACG CCCGG CSTGG CCCACAGGC		12 12 12 12
ADH2 (alcohol dehydrogenase)	yeast	-257 to -216	deletions	22 bp dyad symmetry	ADR1	13
albumin	rat/liver	-156 to -141 -126 to -107 -105 to - 89 - 72 to - 35	ia ia ia ia	GCAAGGGATTTAGTTA TTTTTBGCAAGGAT ATTTTGTAAT TGGTTAATGATCTACAGTTATTGGTTA	NF-1	14 14 14 14
A-HuLV (amphotrop. murine leukemia virus)	/F9, PCC4	-87 to -59	1a, 3	CCAAT	EPBF	15
aP2 (adimocuta P2)	mouse/adipocytes	-124 to -108	1a, 3	AACATBACTCAGAGGAA	c-fos ?	16

Wingender E. Compilation of transcription regulating proteins. Nucleic Acids Res. 1988 Mar 25;16(5):1879-902. PMID: 3282223



## The present status

A BKL Locus Report (rat albumin)





## The preser Site Report



R26428

Site: RAT\$ALBU 24 - ALB (albumin)

#### Binding Site Information [hide] Identifier RAT\$ALBU 24 Gene Rat Alb (albumin) Region promoter Sequence agettcagaTGGCAAACATACgca Export FASTA Sequence Type | DNA Reference Point for Sequence Start Element | Site X Element Range from -177 to -154 External Identifiers | EMBL/GenBank/DDBJ: \$82890; (327:350) Element Mapped to Promoter(s) Binding Factors | HNF-3beta(r) | Quality:1 HNF-3alpha(r) Quality:2 HNF-3alpha(h) Quality:5 Factor Source rec(rat-COS-7); Rat; recombinant expression; rat factor has been expressed in COS-7 HepG2: Human: Method functional analysis, direct gel shift, gel shift competition Comments | Positive regulatory element [1];

## A Site entry

#### References [hide] (1 entry)



[1] PMID 10024498. Hsiang, C. H., Marten, N. W., Straus, D. S. Upstream region of rat serum albumin gene promoter contributes to promoter activity; presence of functional binding site for hepatocyte nuclear factor-3. Biochem J 338 241-9. (1999)



## The present status

#### Number of Entries

	Compilation 1988	TRANSFAC 2009.1ª
Transcription factors	145	12,183 <sup>b</sup>
Binding sites	464	24,745 <sup>c</sup>
Factor-site links	361	33,513
Genes	122	36,317
ChIP-chip fragments		155,306
Matrices		885
References	209	20,072

New high-throughput technologies quickly populate the genomewide map.

Additional 130,000 ChIP-Seq fragments coming up with the Summer release.

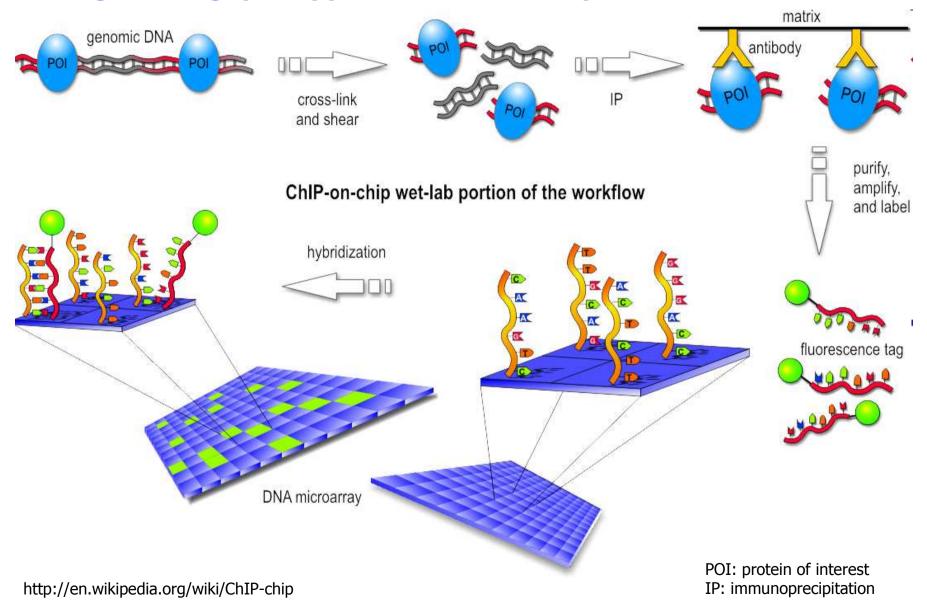
<sup>a</sup> as of March 31, 2009

b including 237 miRNAs

<sup>&</sup>lt;sup>c</sup> including 577 miRNA target sites

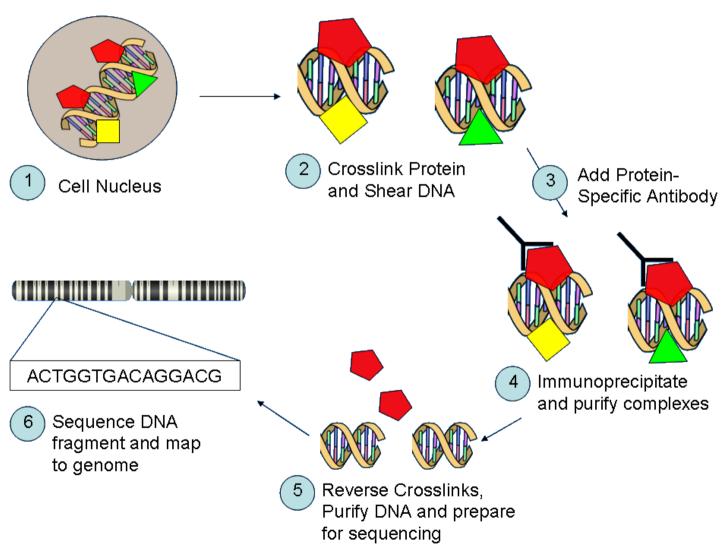


## **High throughput approaches: ChIP-chip**





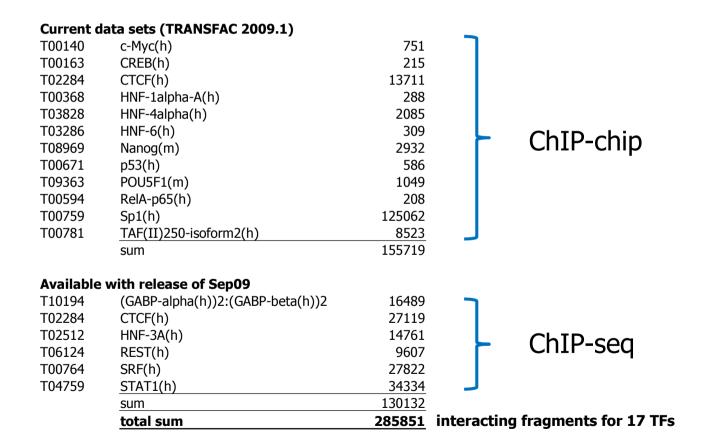
## High throughput approaches: ChIP-seq





## The present status

Additional 130,000 ChIP-Seq fragments coming up with the Summer release ...





## The present status

... and more coming up until the end of 2009, totaling ~790,000 genome fragments interacting with 40 different TFs

Most of the fragments come from the international ENCODE project.

## Further upcoming datasets for human TFs: c-Fos 2939:

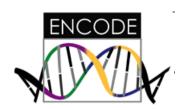
C-1 US	23331
с-Мус	27921
JunD	6352
NF-E2	13325
SREBP-1a	1147
Tcf4	52293
c-Jun	60574
GATA-1	11097
Max	27084
Pol2	56968
SREBP2	1145
ZNF263	32101
Sum	319398

#### **Further upcoming datasets for mouse TFs:**

c-Myc	3422
CTCF	39601
E2F1	20696
ESRRB	21644
Klf4	10873
NANOG	10343
n-Myc	7182
Oct4	376
P300	524
SmaD1	1126
Sox2	4526
Stat3	2546
Suz12	4215
TCFCP2l1	26908
Zfx	10336
FoxA2	11462
P300	2453
Sum	181617



### **The ENCODE Project**



- ENCyclopedia Of DNA Elements
- **Goal:** To compile a comprehensive parts list of functional elements in the human genome
- Pilot Project:
  - 2003-2007
  - Experiments focused on a limited set of genomic regions comprising about 1% of the human genome
- In September 2007, the ENCODE project was scaled up from pilot to productive phase, aiming at the coverage of the entire human genome



## **The ENCODE Project**

- Protein-coding and non-protein coding genes
  - Full-length coding sequence and variants
  - Transcriptional regulatory elements
  - All pseudogenes
- Global sequence features:
  - Methylation/CpG islands, sequence variation, evolutionary history of sequence blocks, and repetitive elements
  - Non-coding chromosomal elements:
    - Origins of replication, nuclease hypersensitive sites, matrix attachment sites, and histone modifications



## The ENCODE Project: experimental protocols applied

Feature Class	Experimental Technique(s)	Abbreviations	No. Exp. Data Points
Transcription	Tiling array, Integrated annotation	TxFrag, RxFrag, GENCODE	63,348,656
5' Ends of transcripts	Tag sequencing	GIS-PET, CAGE	864,964
Histone modifications	Tiling array	Histone nomenclature, RFBR	4,401,291
Chromatin structure	QT-PCR, Tiling array	DHS, FAIRE	15,318,324
Sequence- specific factors	Tiling array, tag sequencing, Promoter assays	STAGE, ChIP- Chip, ChIP-PET, RFBR	324,846,018
Replication	Tiling array	TR50	14,735,740
Computational analysis	Computational methods	CCI, RFBR Cluster	NA
Comparative sequence analysis	Genomic sequencing, multi- sequence alignments, computational analyses	CS	NA
Polymorphisms	Resequencing, copy no. variation	CNV	NA



## **The ENCODE Project: Status**

- Pilot project results (Nature 447, 799-816, 2007):
  - Transcription more complex than expected
  - Transcription Start Sites (TSS) more numerous than protein-coding genes
  - Regulatory information is distributed:
    - clustered across the genome, distribution near TSSs is symmetrical
  - Many distal DNaseI Hypersensitive Sites (DHSs)
  - Replication correlated with histone structure in a more detailed manner than previously known
- 9 Dec 2008 First ENCODE whole-genome data freeze completed



However, even with the new HTP technologies, we will not be able to investigate the complete "promotome" for:

- all TFs (~1800 TF genes in human) in
- all cell types (~300 human cell types),
- all organs (~14000 morphologically distinguishable structures)
   at
- all developmental stages (>24 in human) and
- under all environmental conditions  $(--> \infty)$ .

But what we can do is to learn about the rules behind the observations and to develop predictive models.



## Goal #3:

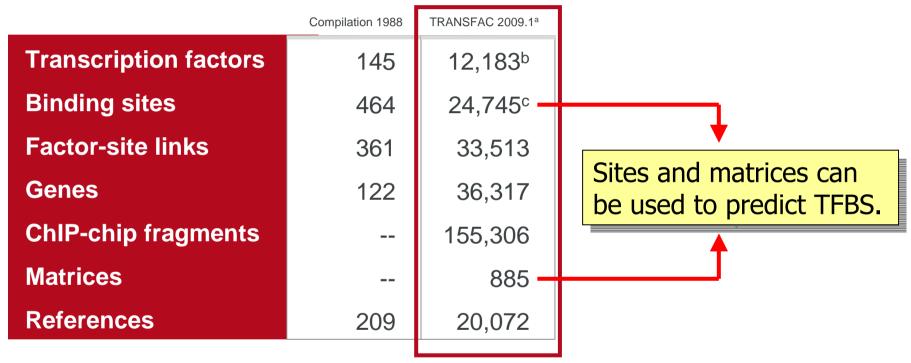
To enable prediction of new transcription factor binding sites.

#### **Prediction of TFBSs**



#### **Prediction-relevant contents in TRANSFAC**

#### Number of Entries



<sup>&</sup>lt;sup>a</sup> as of March 31, 2009

b including 237 miRNAs

<sup>&</sup>lt;sup>c</sup> including 577 miRNA target sites



#### **Matrix construction**

#### **TRANSFAC**

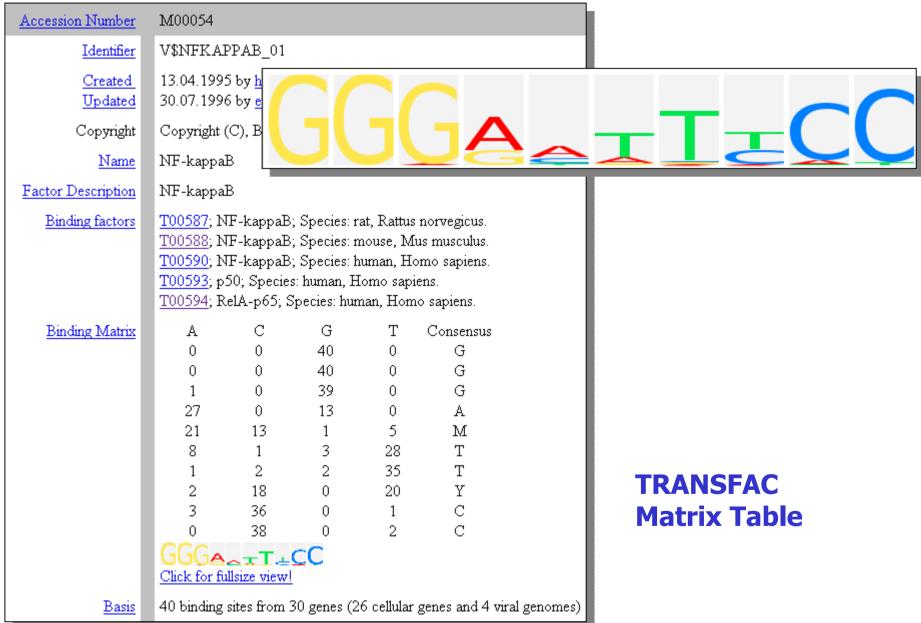
A 3 3 1 11 0 0 0 0 2 2 0 0 C 6 2 1 0 12 0 0 0 7 3 2 4 G 1 7 3 1 0 12 0 12 3 6 8 4 T 2 0 7 0 0 0 12 0 0 1 2 4



GCCCTACGTGCTGTCTCA
CAGGCAACGTGCAGCCGGA G
CAGTGCATACGTGGGCTCCA
CTTTGTGTGTACGTGCAGGAA
GAAATACGTGCGCTTTGTGTG
CGCGAGCGTACGTGCCTCAGG
CCCCCTCGGACGTGACTCGGACCAC
AGGGCCGGACGTGACTCGGACCAC
ACGCTGAGTGCGTGCGGACCCC
GCAGTACGTGCGGGACCCC
ACGCTGAGTGCGTGCGGGAC
CCCAGCCTACACGTGGGGTTC
GGAGCCCAGCGGACGTGCGGGAA

#### **Prediction of TFBSs**





Matys et al., Nucleic Acids Res. 34, D108-D110 (2006).



# TFBS detection with TRANSFAC matrices and the Match™ algorithm:

$$q = \left(\sum_{i=1}^{L} I(i) f_{i,b_i} - \sum_{i=1}^{L} I(i) f_i^{\min}\right) / \sum_{i=1}^{L} I(i) f_i^{\max}$$

with:  $b_i$ , nucleotide b found in the i-th position of test sequence,

 $f_{bi}$ , frequency of nucleotide b in the i-th position of the aligned training sequences,  $f_i^{min}$ , minimum frequency in position i,  $f_i^{max}$ , maximum frequency in position i, and

$$I(i) = \sum_{B \in \{A, T, G, C\}} f_{i,B} \ln(4f_{i,B}), \quad i = 1, 2, ..., L$$



## Validation of potential TFBSs by comparative genome analysis:

## General idea:

Genomic sites that are functionally important are under evolutionary pressure and, thus, are more conserved among related genomes than the genomic background ("phylogenetic footprinting").

This could add an independent criteria to the pattern-based prediction of TFBSs.



## **Validation of potential TFBSs by comparative genome analysis:**

#### However:

What has to be conserved, the sequence or the pattern?



## Validation of potential TFBSs by comparative genome analysis:

```
core sim/matrix sim
Sequence-only conservation:
                                        1.000 / 0.997
GGGGAATTTCC (NF-kB consensus):
** *****
GGAGAATTTCC (10/11 match, 91%): 0.713 / 0.834
** * ****
                                        0.426 / 0.672
GGAGTATTTCC (9/11 match, 82%):
                                         Matrix V$NFKB_Q6, M0194
Pattern-only conservation:
                                        0.991 / 0.983
AATGCCTGAGGCGCT (AP-2\alphaA):
   ***
         * * *
                  (6/15 match, 40%)
TTCGCCCCAGGGCGC (AP-2\alphaA):
                                        0.957 / 0.951
```

Matrix V\$AP2ALPHA\_02, M01045

#### **Prediction of TFBSs**



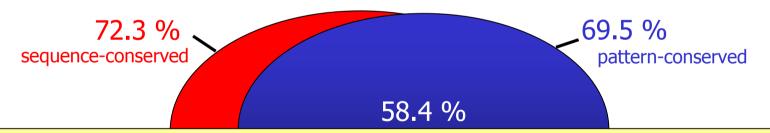
#### Validation of potential TFBSs by comparative genome analysis:

Human / rodent comparison for all corresponding TRANSFAC sites:

General sequence conservation of TFBS: 72.3 %

Background conservation in upstream sequences: 35.2 %

Pattern-conserved TFBS: 69.5 %



When comparing two genomes, sequence and pattern conservation are not identical.

sequence-only pattern-only conserved conserved

Sauer et al., Bioinformatics 22:403 (2006)



#### Validation of potential TFBSs by comparative genome analysis:

#### human c-jun gene

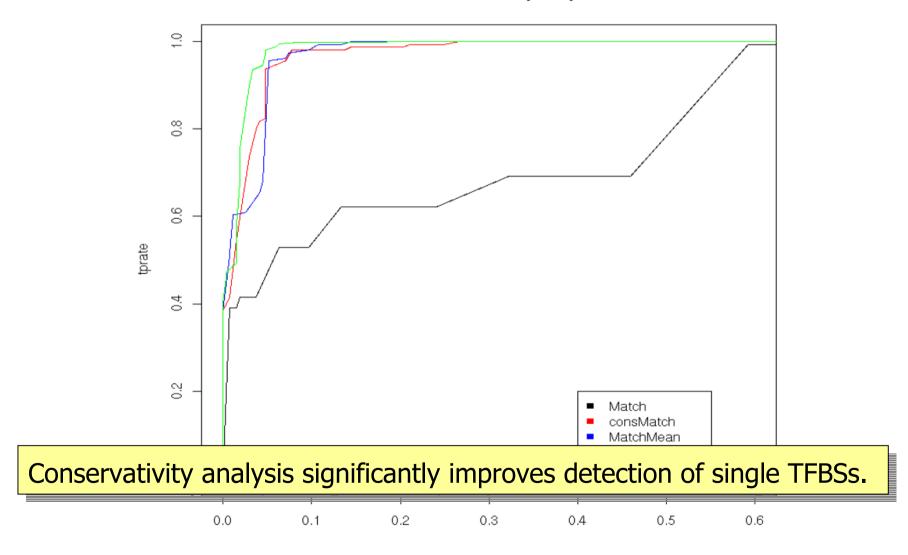
```
R00960 (AP-1
     : GCTGCG------CTCGAGAGAGCTCCGTGAGTGACCGCGACTTT @ 953/1171
COW
     doq
human : GCTGCG-----CACGAAGAGCCCCTCAGTGAGTGACCGCGACTTT @ 1001/1217
mouse : GCTGCT------CCCCGAGAGCCCCTCCGTGAGTGACCGCGACTTT @ 964/1173
     : GCTGCT-----TCCCGAGAGCGCTCCGTGAGTGACCGCGACTTT @ 957/1165
rat
     = 1051
              1061
                      1071
                              1081
                                       .091
                             M00925 (AP-1)
     : T--CAAAGCCGGGCGCGCGCG---AGCCCACACTAAGAGCGCGGGC @ 998/1171
COW
     human : T--CAAAGCCGGGTAGCGCGCG---AGTCGACAAGTAAGAGTGCGGGA @ 1046/1217
mouse: T--CAAAGCTCGCGCGCGGG---AGCCTACCAACGTGAGTGCTAGC @ 1009/1173
     : T--CAAAGCTCCGGATCGCGCGGG---AGCCAACCACGTGAGTGCAAGC @ 1002/1165
rat
              1111
                       1121
                                       1141
     = 1101
                               1131
```

Multiple genome comparison may better reveal conserved sites, and may pinpoint to "surrogate" sites.



## Validation of potential TFBSs by comparative genome analysis:

ROC-curve NFK(m774)



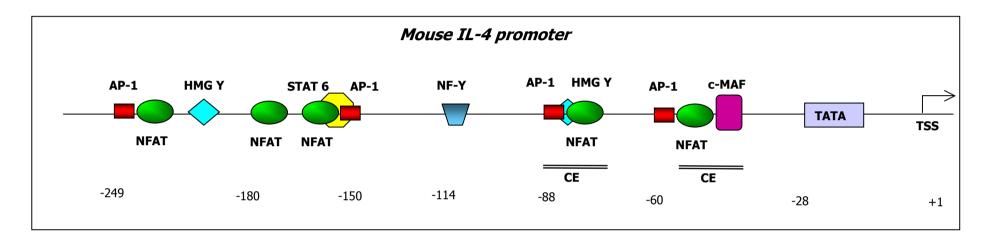


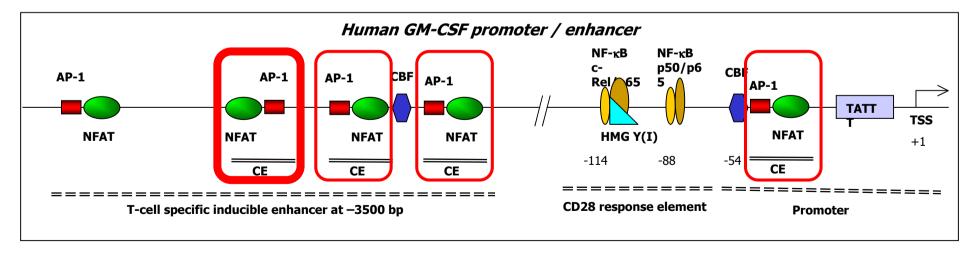
## Goal #4:

To reveal the complex structure of regulatory genome regions (promoters, enhancers, etc.).



#### **Promoters and enhancers: examples**







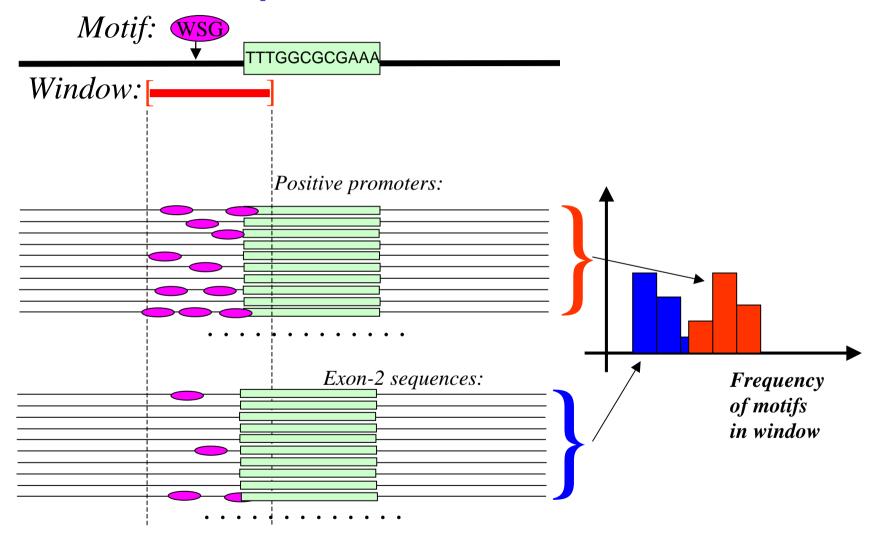
### **Local context analysis**

## Task:

Finding nucleotide patterns around specific TFBSs which may facilitate or impede TF binding.



## **Local context analysis**



Kel et al., Genome Biol. 2008, 9:R36.



#### **Local context analysis**

$$f(\lambda, w, S) = \frac{N(\lambda, w, S)}{t_2 - t_1}$$

#### Context Score:

$$d(S) = \beta + \sum_{i=0}^{k} \alpha_i \times f(\lambda_i, w_i, S)$$

 $N(\lambda, w, s)$ : Number of motifs  $\lambda$  in window  $w = [t_1, t_2]$  of sequence S

	N	Motif (λ)	Window $(w)^{1)}$	$\hat{f}^{\scriptscriptstyle Y}/\hat{f}^{\scriptscriptstyle N}$	Utility	$\alpha_{i}$
	1	MGCG	[27,34]	0.0048 / 0.0041 = 1.179	0.80	0.394
SS	2	TTT	[39,41]	0.0112 / 0.0032 = 3.536	0.75	0.9618
ve risti	3	CGSK	[17,38]	0.0851 / 0.0341 = 2.499	0.90	0.5353
Positive	4	HKCG	[13,16]	0.0675 / 0.0095 = 7.071	0.79	0.5904
Positive characteristics	5	<b>VDWW</b>	[17,46]	0.1233 / 0.0536 = 2.299	0.72	0.223
ch	6	DWTT	[21,26]	0.0337 / 0.0000	0.80	0.5036
	7	GSDM	[3,69]	0.0980 / 0.0559 = 1.754	0.82	0.595
tics						
Negative characteristics	8	vws	[7,66]	0.1258 / 0.1932 = 0.651	0.91	-0.095
ega	9	HSWY	[26,65]	0.0413 / 0.0813 = 0.508	0.79	-0.2297
n har	10	VTV	[19,34]	0.0427 / 0.1354 = 0.315	0.71	-0.261
	11	BAY	[7,65]	0.0274 / 0.0614 = 0.447	0.78	-0.566
						$\beta = -5.6767$

E2F binding sites start at pos. 31; Y and N: positive and negative sequence set; "utility": -1 < U < +1, calculated from the average difference, distribution overlap, normality of the distribution, *bootstrap* test, etc.



#### **Global context analysis**

## Task:

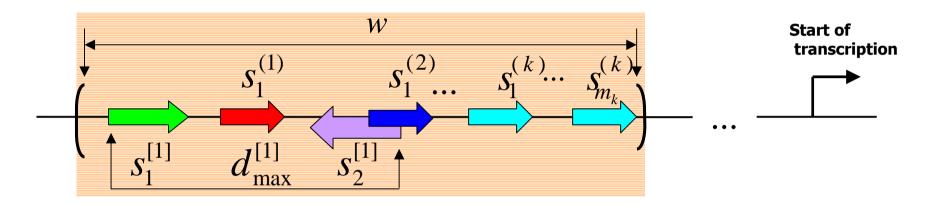
Finding (a) combination(s) of TFBS ("Composite Modules") that is/are characteristic for a given set of promoters.



### **Composite module analyzer (CMA)**

Kel et al., Bioinformatics 22, 1190-1197 (2006).

#### Mathematical model (v.2)

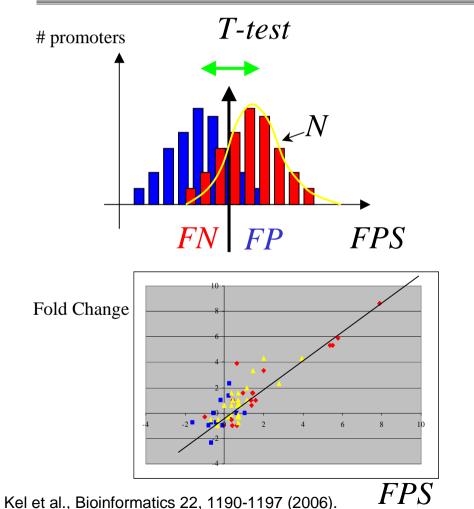


A CM contains <u>single elements</u> as well as <u>composite elements</u> (site pairs)



#### Fitness function of the Genetic Regression Algorithm (GRA)

$$F = \alpha \cdot \mathbf{I} + \beta \cdot \mathbf{I} - FN + (1 - \beta) \cdot \mathbf{I} - FP + \gamma \cdot \mathbf{I} + \delta \cdot \mathbf{N} - \mu \cdot k$$



R – linear regression

*FN* – false negatives

**FP** – false positives

T – T-test (difference between mean values)

N – normal likeness

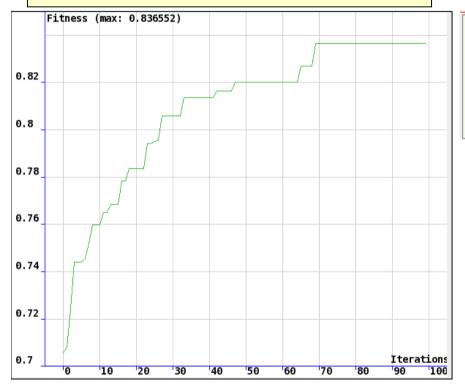
k – number of free parameters



### **Composite module analyzer (CMA)**

#### Promoter model generation

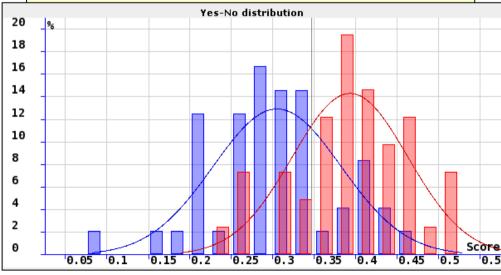
# Increase of Fitness function with number of iterations



#### Composition of the promoter model

V\$OCT1 05 01		<u>V\$SP1_01</u>
<u>V\$AP2_Q3</u>		<u>V\$EGR2_01</u>
<u>V\$HIF1_Q5</u>	or	<u>V\$FOXM1_01</u>
V\$USF_Q6		<u>V\$CREB_Q2_01</u>
<- <u>V\$USF_C</u> -> [330] <- <u>V\$AR_01</u> ->		<- <u>V\$GR_Q6</u> -> [330] <- <u>V\$EGR_Q6</u> ->

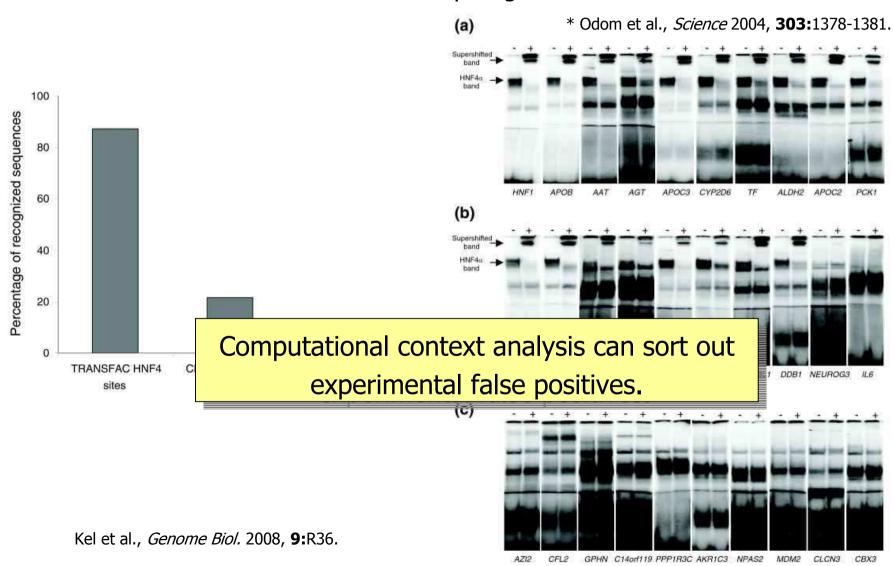
# Sequences of YES and NO sets are well separated by the selected promoter model





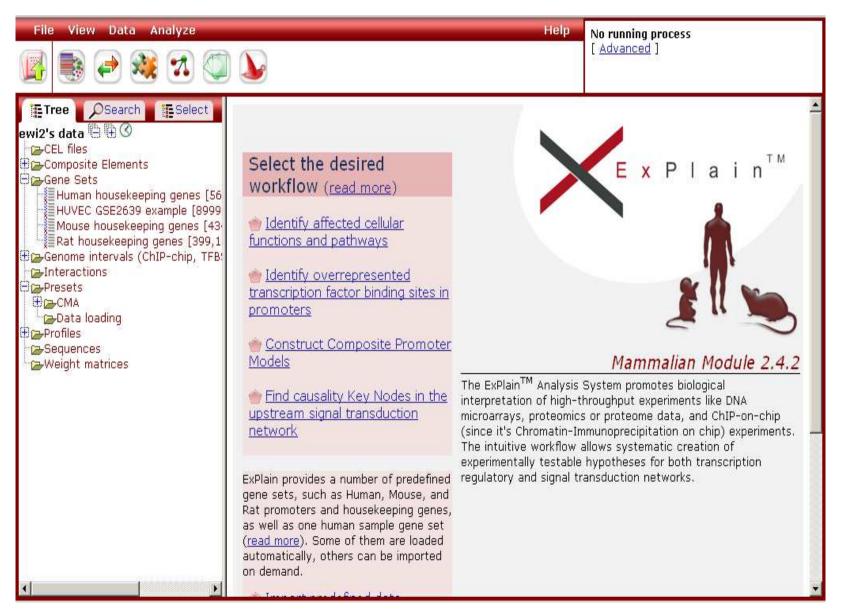
### **Composite module analyzer (CMA)**

HNF- $4\alpha$  sites of ChIP-chip fragments\* revisited:



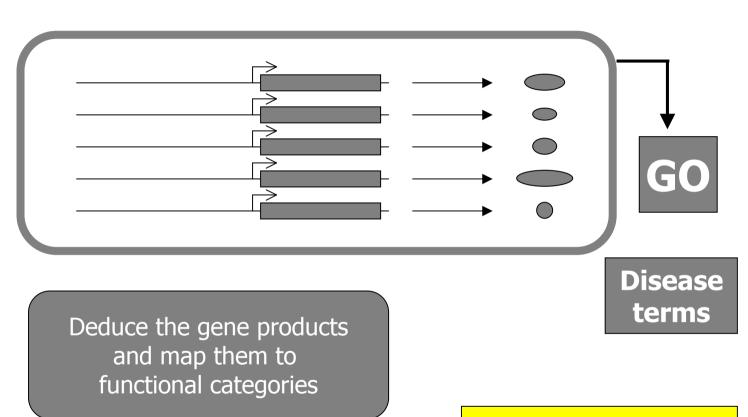


### **The ExPlain™ System**





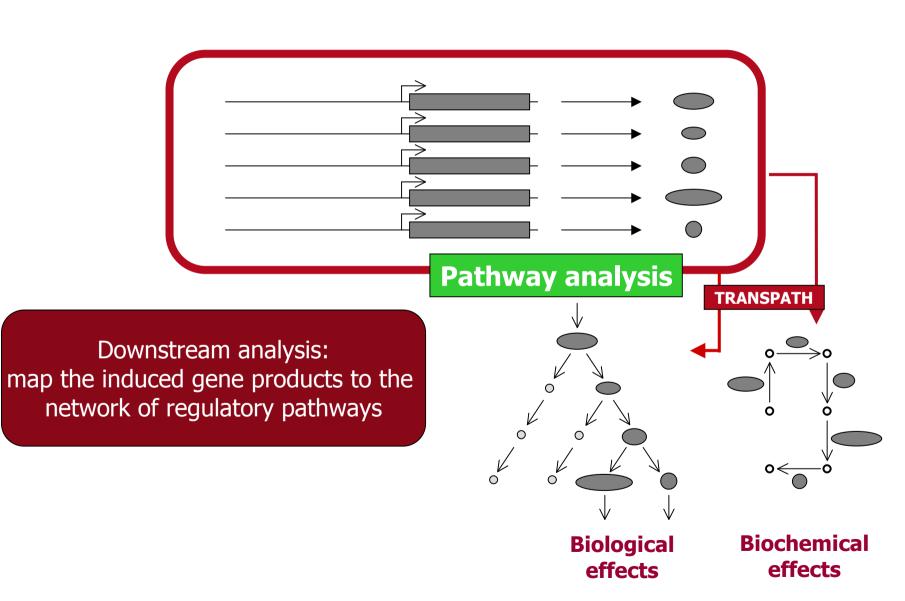
### **The ExPlain™ System**



**Functional analysis** 



### **The ExPlain™ System**





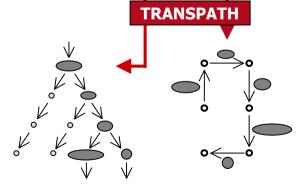
**The ExPlain™ System** 

Pathway analysis

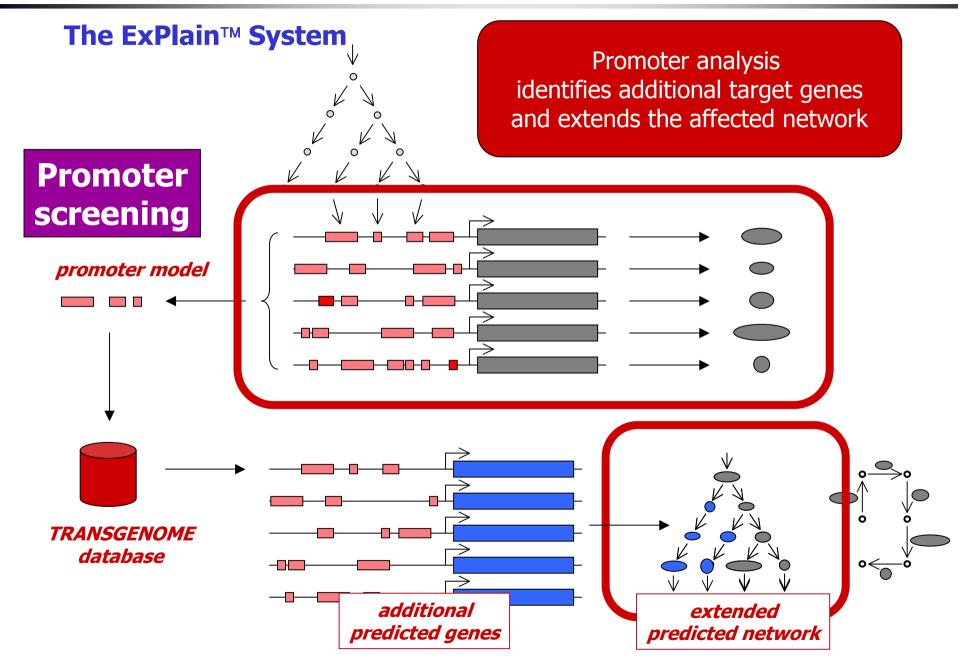
**Promoter analysis** 

Identification of new targets

Reasoning
of experimental findings:
promoter analysis of induced genes
connected to network mapping

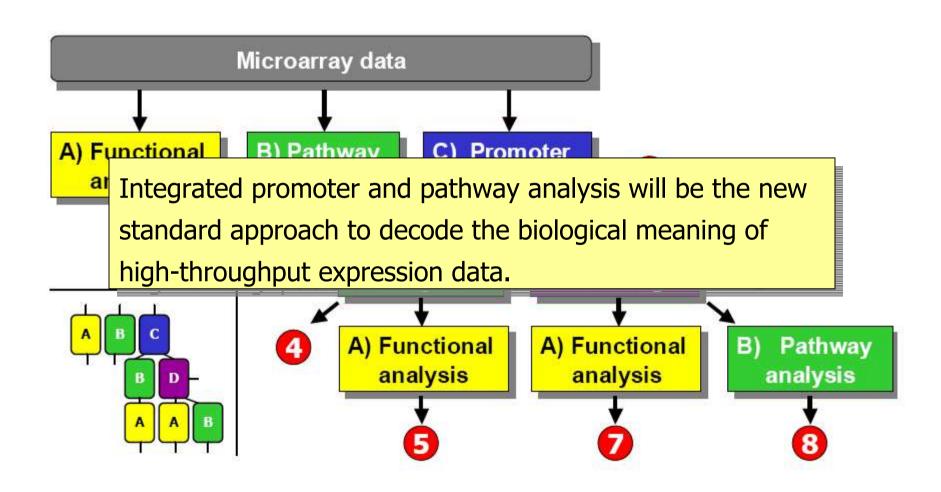








#### The ExPlain™ System: workflow example





## Goal #5:

Deciphering the Protein-DNA interaction code in order to predict the DNA-binding specificity of new transcription factors.



# The first compilation: The FACTOR table

The zinc-finger structure: The eukaryotic alternative to the bacterial helix-turnhelix motif?

Wingender E. Compilation of transcription regulating proteins. Nucleic Acids Res. 1988 Mar 25;16(5):1879-902. PMID: 3282223

factor	source	<u>qene</u>	M. W. (kDa) finger str	synonyms. quivalents	<u>ref.</u>
60K protein	soybean	lectin	60		104
a4 protein	munan (HeLa)	€0 (HSV) ≪4	2x 170	ICP4, Vav175	84, 176 84
Adf-1	Drosophila	adh			11
ADR1	yeast	ADH2	151 / 2 CH		13, 177
— AP=1 — —	husan	collagenase IL-2 (?) metallothionein IIA polyoma virus (?) rat stromelysin (?) SV40 enhancer			21, 110, 178 21
AP-2	HeLa	BPV GH metallothionein IIA c-myc MHC class I H-2K <sup>m</sup> SV40 enhancer	50; 52	KBF1 (mouse) ?	18, 110 18 18 18 18 18, 110, 162
AP-3	HeLa	enhancer (SV40)			110, 162
B factor	Orosophila •	actin 5C histone H3 histone H4		TFIID (HeLa) ?	3
CBF	sea urchin	histone H2B-1			64
CBP	mouse human rat rat	e-globin hsp70 ? LTR (MSV) tk (MSV)		CTF ?	51 79 125 125
CDF	sea urchin	histone H2B-1			64
CCAAT-binding factor	human sea urchin mouse	hsp70 histone H2B-1 α <sub>2</sub> (I) collagen		CTF	78 64 20
CCAAT box bind.protein	mouse, rat, human(HeLa)	€ <sub>A</sub> globin	64 ?	CCAAT-bind.f.; CTF;CBP ?	51



## The first compilation: Zinc finger alignments

	gene. gene product (**) (ref.)	position	finger sequences
	TFILLA	1	M & E K A L
	(210, 211)	7	PVVYKRYI <u>C</u> SFAD <u>C</u> GAAYNKNWKLQA <u>N</u> LCK <u>N</u>
	•	38	TGEK PFP <u>c</u> keeg <u>c</u> ekgftslhhltr <u>W</u> slt <u>W</u>
		68	TGEK N <i>f</i> T <u>C</u> DSDG <u>C</u> DLR <i>f</i> TTKANMKK <u>N</u> FNRF <u>N</u>
		99	NIKICVYV <u>C</u> HFEN <u>C</u> BKA <i>F</i> KKHNQLKV <u>X</u> QFS <u>X</u>
		130	TQQL PYE <u>Č</u> PHEG <u>Č</u> DKR <i>f</i> SLPSR <i>l</i> KR <u>N</u> EKV <u>N</u>
		160	AG YPČKKDDŠ <u>C</u> SFVGKTHTLYLK <u>H</u> VAEC <u>H</u>
		189	QD LAVĘ DVĘNRKFRHKDYŁRO <u>N</u> QKT <u>N</u>
		215	EKERTVYL <u>Č</u> PRDG <u>Č</u> DRSYTTAFN <i>L</i> RS <u>W</u> IQSF <u>W</u>
		247	EEQR PFV <u>ē</u> EHAG <u>ē</u> GKCFAMKKSLER <u>N</u> SVV <u>N</u>
	Xfin	103	SAKK SHI <u>C</u> S NTGKLFSCTAAVVR <u>H</u> QRM <u>N</u>
	(212)	131	QLQK SHHÇ PH <u>C</u> KKSFVQRSDFIK <u>M</u> QRT <u>M</u>
		159	TGER PYR <u>C</u> VE <u>C</u> RKKFTERSALVN <u>H</u> RRT <u>H</u>
		187	TBER PYT <u>C</u> LD <u>C</u> RKTFNRRSALTK <u>H</u> RRT <u>H</u>
		215	TEER PYR <u>C</u> SV <u>C</u> SKSFIQNSDLVK <u>H</u> LRT <u>H</u>
		243	TBEK PYE <u>Č</u> PL <u>Č</u> VKRFAESSALMK <u>H</u> KRT <u>H</u>
		271	STHR PFR <u>C</u> SE <u>C</u> SRSFTHNSDLTA <u>N</u> MRK <u>N</u>
Wingender E. Compilation of		299	TEFR
transcription regulating prot		320	NVASSPYSC SK CRKTFKRWKSFLN <u>K</u> QQT <u>M</u>
Nucleic Acids Res. 1988 Mar	•	349	SREK PYLČ SH ČNKGFIRNSDLVK NFRT N
25;16(5):1879-902.		377	TEER PYRC AE CHKEFIRKSDLYK ALRT H
PMID: 3282223		405	TEEK PEKC SH COKKETERSALAK NORT N
		433	TEEK PYKÉ SD ÉGKEFTORSHLIL MORI M
		464	THE BURK TI ARBTETOMERIUM MONU M



# The helix-turn-helix motif as paradigm of prokaryotic DNA-binding domains

Sauer RT, Yocum RR, Doolittle RF, Lewis M, Pabo CO (1982) Nature 298, 447-451:

Homology among DNA-binding proteins suggests use of a conserved super-secondary structure.

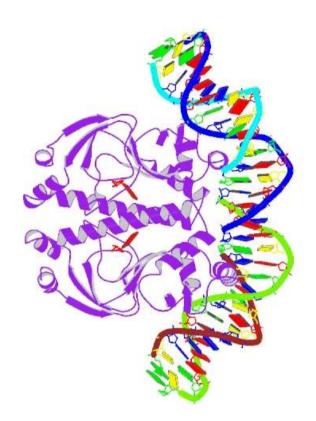
The amino acid sequences of the repressor and cro proteins of phages lambda, 434 and P22 are homologous, especially in a region in which repressor and lambda cro have a similar alpha-helix-turn-alpha-helix secondary structure. Model-building studies indicate that this structure is important in DNA binding, and we suggest it may be a common feature of many DNa-binding proteins.

PMID: 6400820



# The helix-turn-helix motif as paradigm of prokaryotic DNA-binding domains

PDB entry 1CGP



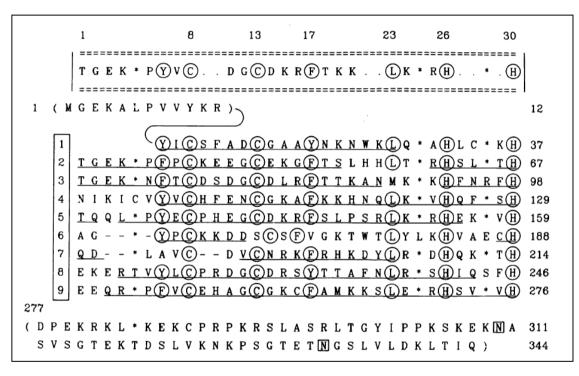
Schultz, S.C., Shields, G.C., Steitz, T.A. (1991) Crystal structure of a CAP-DNA complex: the DNA is bent by 90 degrees. Science **253**: 1001-1007

dating back to:

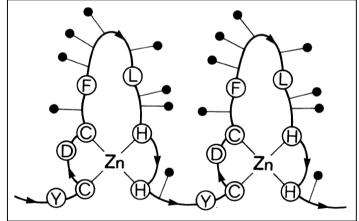
McKay, D.B., Steitz, T.A. (1981) Structure of catabolite gene activator protein at 2.9 A resolution suggests binding to left-handed B-DNA. Nature **290**:744-749.



# The zinc finger motif as general paradigm for eukaryotic DNA-binding domains?



9 Consecutive Cys<sub>2</sub>-His<sub>2</sub> motifs in TFIIIA identified, with a zinc finger structure proposed.



Figs. 3 and 4 in:

Miller J, McLachlan AD, Klug A. (1985) Repetitive zinc-binding domains in the protein transcription factor IIIA from Xenopus oocytes. EMBO J. **4**:1609-1614.

PMID: 4040853



# The zinc finger motif as general paradigm for eukaryotic DNA-binding domains?

PDB entry 1TF3



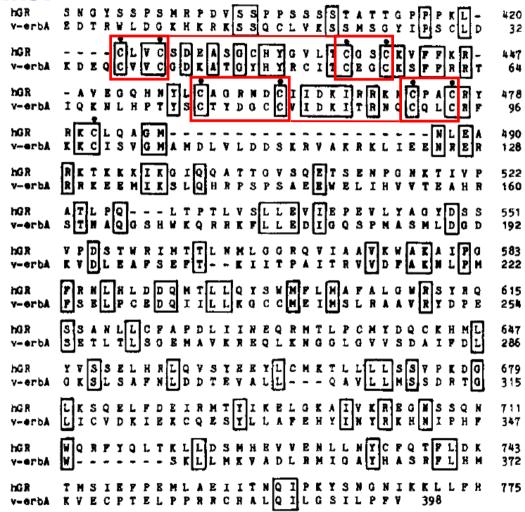
Foster, M.P., Wuttke, D.S., Radhakrishnan, I., Case, D.A., Gottesfeld, J.M., Wright, P.E. (1997) Domain packing and dynamics in the DNA complex of the N-terminal zinc fingers of TFIIIA. Nat.Struct.Biol. **4:** 605-608



# The zinc finger motif as general paradigm for eukaryotic DNA-binding domains?

#### **GR** zinc finger

Similarity with TFIIIA zinc finger structure suggested for the first time.

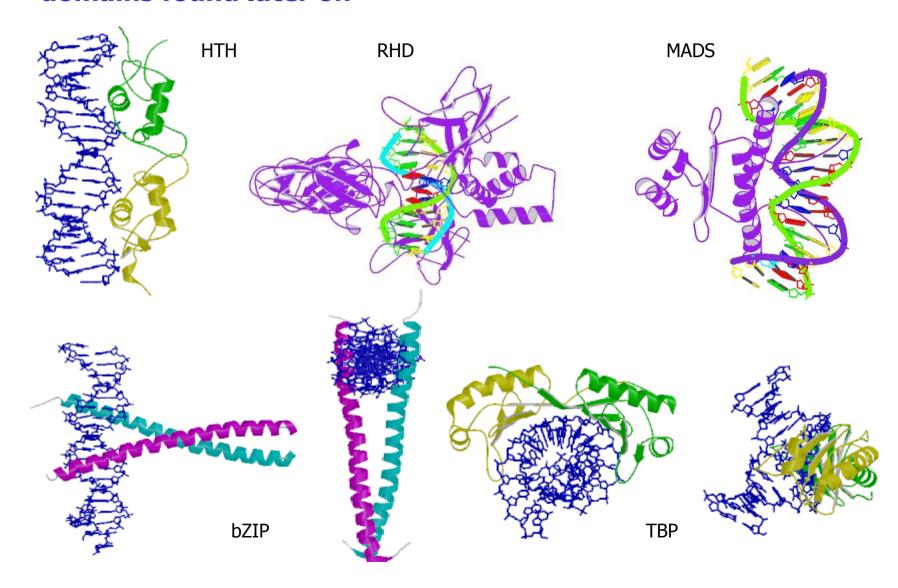


Weinberger C, Hollenberg SM, Ong ES, H 108 Q K 777

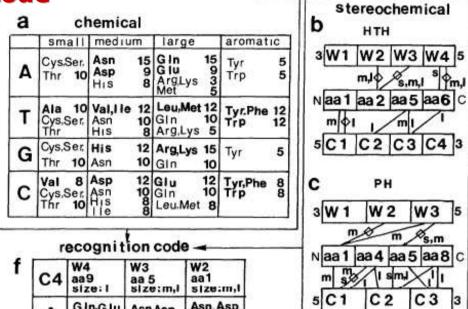
Identification of human glucocorticoid receptor complementary DNA clones by epitope selection. Science. 1985 May 10;228(4700):740-2. PMID: 2581314



# Many more structural classes of eukaryotic DNA-binding domains found later on



**Protein-DNA** recognition code?



Leu Met

The protein-DNA recognition code is class-specific.

Thus, a comprehensive TF classification is required.

Leu . Met

W3 W2 aa 3 size:m(I) size: I W4 aa6 size: I GIn,Glu Asn,Asp GIn.Glu (Gin.Glu) Val, He Leu,Met Leu, Met Leu.Met) Arg, Lys His Arg, Lys (Arg.Lys) Asp (Glu) Glu Leu, Met, Leu, Met, Leu,Met

Leu Met.

Suzuki & Yagi, PNAS 91:12357-12361, 1994



## **Transcription factor classification**

Table 1. Classification scheme for eukaryotic transcription factors

Level	Pattern of classification no.	group designatio n	description	example
1	N	superclass	general topology of DBD	zinc-coordinating domains
2	N.N	class	structural blueprint of the DBD	zinc-finger nuclear receptors
3	N.N.N	family	functional criteria such as protein-DNA-complex formation (DNA-binding specificity, multimerization behaviour) or biological effect	T <sub>3</sub> R/RAR (in contrast to steroid hormone receptors)
4	N.N.N.N	subfamily	mainly according to sequence similarity of the DBDs	RAR (retinoic acid receptor)
5	N.N.N.N.N	genus	according to factor gene	RAR- $\alpha$ , RAR- $\beta$
6	N.N.N.N.N	factor "species"	initiation/ splice/ processing variants	RAR-α1, RAR-α2



1	■ Basic Domains
Transcription f 🥦	★ Leucine zipper factors (bZIP)
1.2	→ Helix-loop-helix factors (bHLH)
1.3	→ Helix-loop-helix / leucine zipper factors (bHLH-ZIP)
1.6	⊞ bHSH
2	☐ Zinc-coordinating DNA-binding domains
2.1	Cys4 zinc finger of nuclear receptor type
2.2	diverse Cys4 zinc fingers
2.3	Cys2His2 zinc finger domain
2.4	⊕ Cys6 cysteine-zinc cluster
2.5	

DNA-binding domains of eukaryotic TFs can be classified in (at least) 4 superclasses and 30 classes.

4	☐ beta-Scaffold Factors with Minor Groove Contacts	
4.1		
4.2	± STAT	

More recent data, however, suggest that there may be up to 13 superclasses (mostly spreading out from previous superclass 4).

0	☐ Other Transcription Factors	
0.1	Copper fist proteins	
0.2	⊞ HMGI(Y)	
0.3		
0.4		
0.5		
176705L		

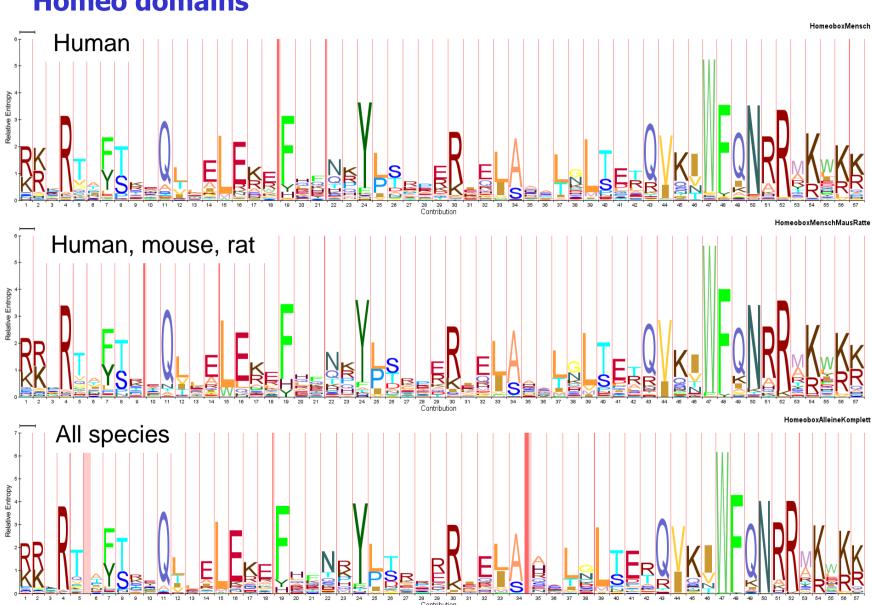


## **Transcription factor classification**

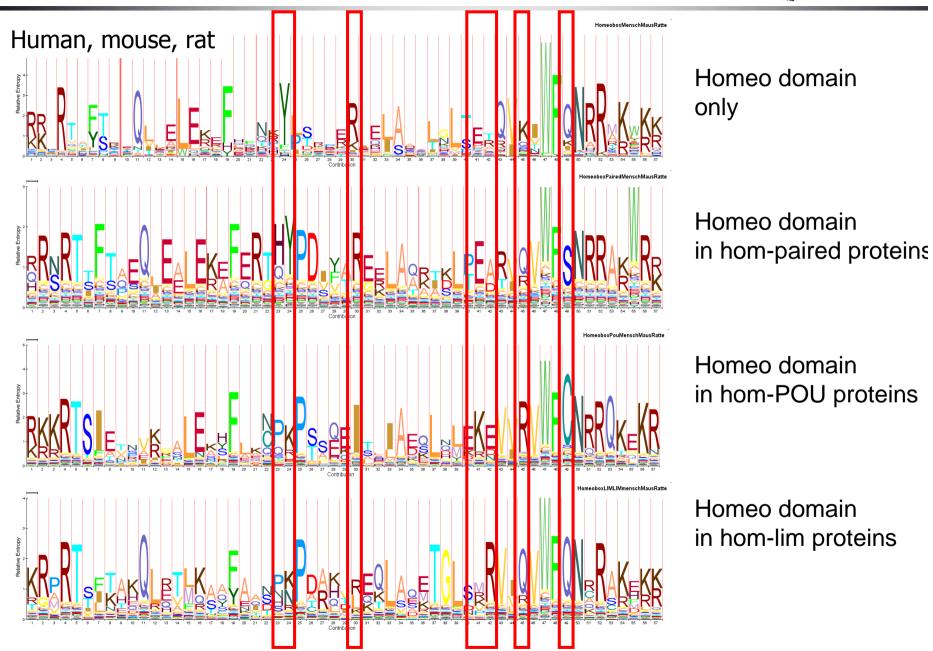
1	<u> Basic Domains</u>
2 2.1	☐ Zinc-coordinating DNA-binding domains
2.1	Cys4 zinc finger of nuclear receptor type
2.1.1	
2.1.1.1	⊟ Corticoid receptors (NR3C)
2.1.1.1.1	□ GR (NR3C1): GR GR GR GR GR GR GR
2.1.1.1.1.1	L GR-alpha: GR-alpha
2.1.1.1.1.2	L GR-beta: GR-beta
2.1.1.1.2	└─MR (NR3C2): <u>MR MR</u>
2.1.1.2	
2.1.1.3	∄ Androgen receptor (NR3C)
2.1.1.4	Estrogen receptor (NR3A)
2.1.1.5	Estrogen related receptor (NR3B)
2.1.2	🛨 Thyroid hormone receptor-like factors (NR0, NR1, NR2, NR4, NR5)
2.2	<u> </u>
2.3	Cys2His2 zinc finger domain
2.4	■ Cys6 cysteine-zinc cluster
2.5	
	⊞ Helix-turn-helix
4	beta-Scaffold Factors with Minor Groove Contacts
3 4 0	⊕ Other Transcription Factors
	1 H 7 CONTACT



#### **Homeo domains**









In a next step, it will be investigated whether whole DBDs or just a few residues are required to reveal mutual dependencies between DBDs and their target DNA sequences.

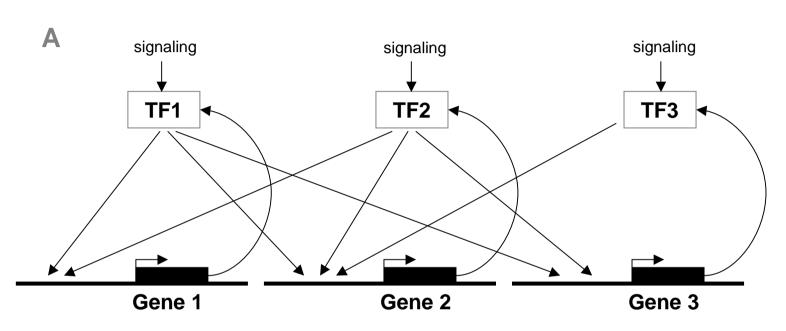


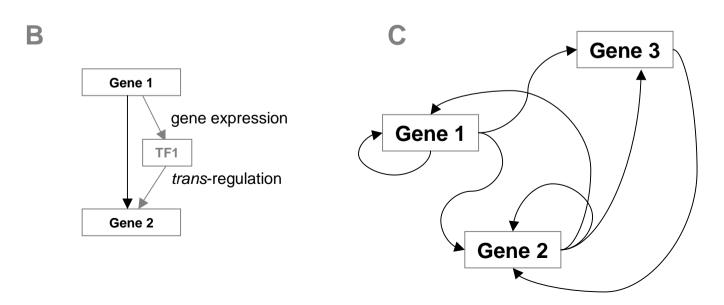
## Goal #6:

Re-engineering and analyzing the complete transcription network of a cell / an organism.



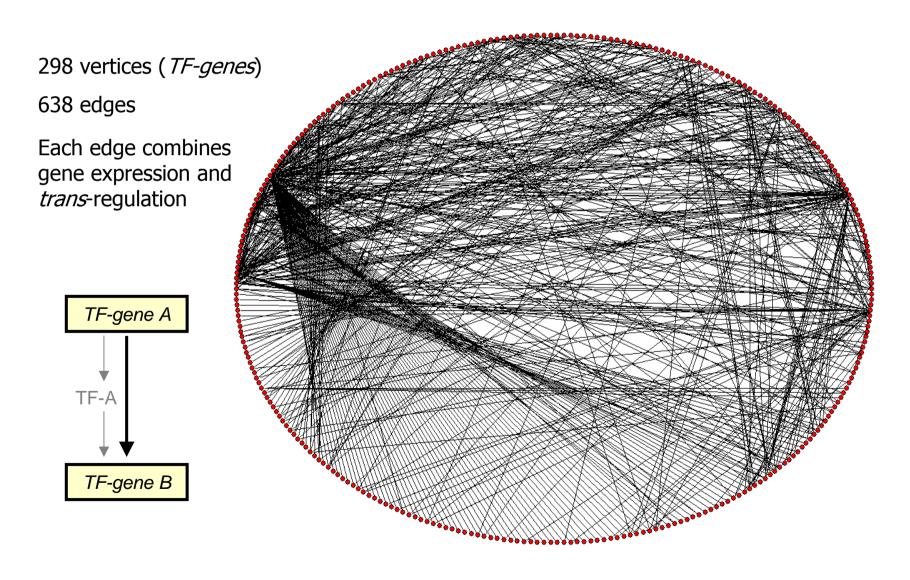
#### **Definition**







#### Mammalian network of transcription factor genes (TFG\_RN)





#### **Disconnectivity network analysis**

Disconnectivity: Effect of removing a vertex

$$Dis(v) = \frac{N_0 - N_{-v}}{N_0} = 1 - \frac{N_{-v}}{N_0}$$

 $N_{\mathcal{O}}$  - number of all-pairs shortest paths

 $N_{-\nu}$  - number of all-pairs shortest paths after removing vertex  $\nu$  from the graph.

$$Dis(v) = \frac{\sum_{s \neq v \in V} \delta_{sv} + \sum_{s \neq v \neq t \in V} \delta_{st}(v) + \sum_{t \neq v \in V} \delta_{vt}}{\sum_{s \neq t \in V} \delta_{st}} \qquad s \rightarrow v \rightarrow t$$

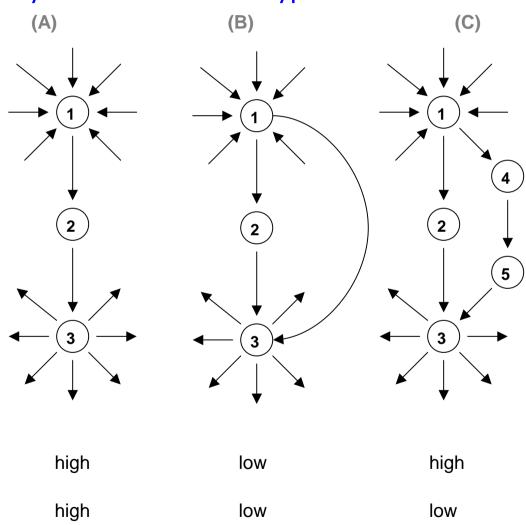
Fraction of all-pairs connections for which vertex  $\mathbf{v}$  is crucial: there is no other parallel path(s) between vertices of such a pair and these vertices will not be connected anymore if vertex  $\mathbf{v}$  is deleted.

The betweenness centrality and the disconnectivity index  $D_i$  display different aspects of individual vertex topology in a whole network.



### **Disconnectivity network analysis**

Disconnectivity sensitive towards bypasses

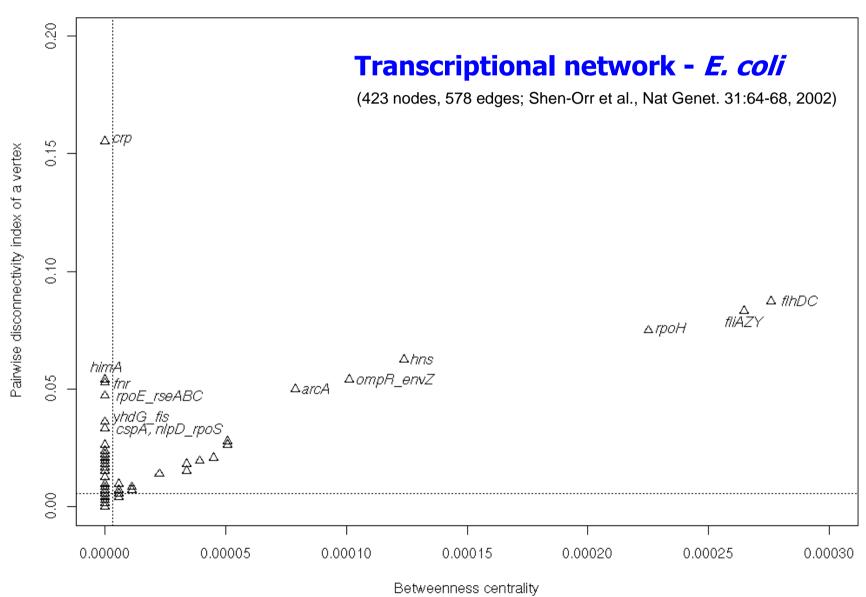


Potapov et al., BMC Bioinformatics, 9, 227 (2008)

*B*(*v*2):

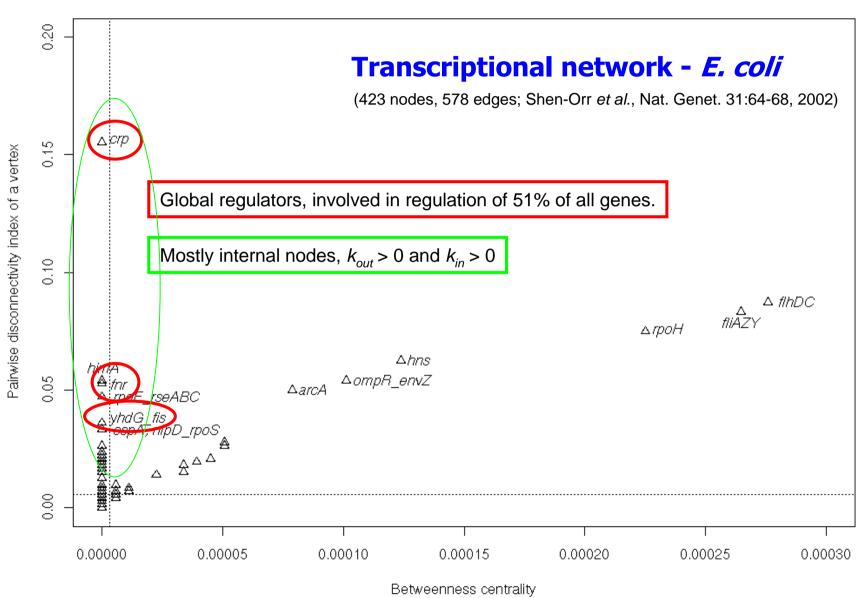
*Dis*(*v*2):





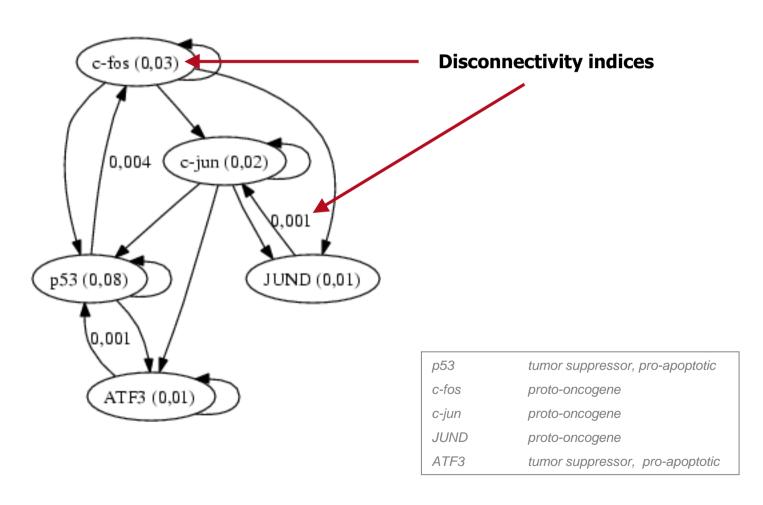
Potapov et al., BMC Bioinformatics, 9, 227 (2008)







#### 'p53' strongly connected sub-graph in TFG\_RN





Pattern / motif characterization by Disconnectivity

		E. coli			S. cerevisiae			Mammals		
Pattern	ID	Freq	Z- $S$ core	$\overline{Dis}$	Freq	Z- $S$ core	$\overline{Dis}$	Freq	$Z ext{-}Score$	$\overline{Dis}$
·	6	4777	11.23	0.0039	11892	14.54	0.0018	1916	-0.39	0.0023
·	12	160	-11.21	0.0189	295	-14.25	0.0135	1068	-1.67	0.011
·	14	-	-	-	18	-1.30	0.0063	73	-10.47	0.0079
·\										

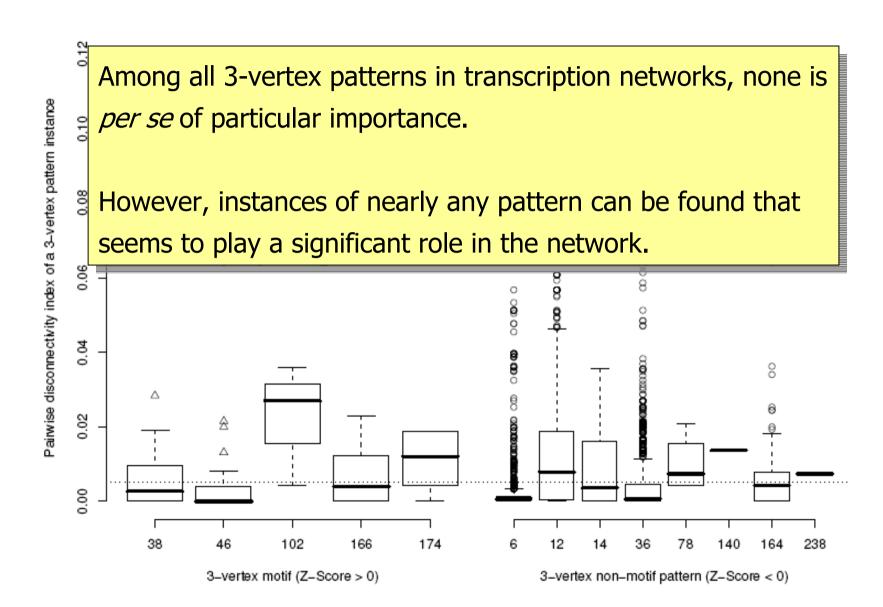
In transcription networks, the only 3-vertex motif appearing in all networks analyzed is the feed-forward loop (confirmed).

However, motifs *per se* do not play a more important role for the coherence of the network than non-motif patterns.

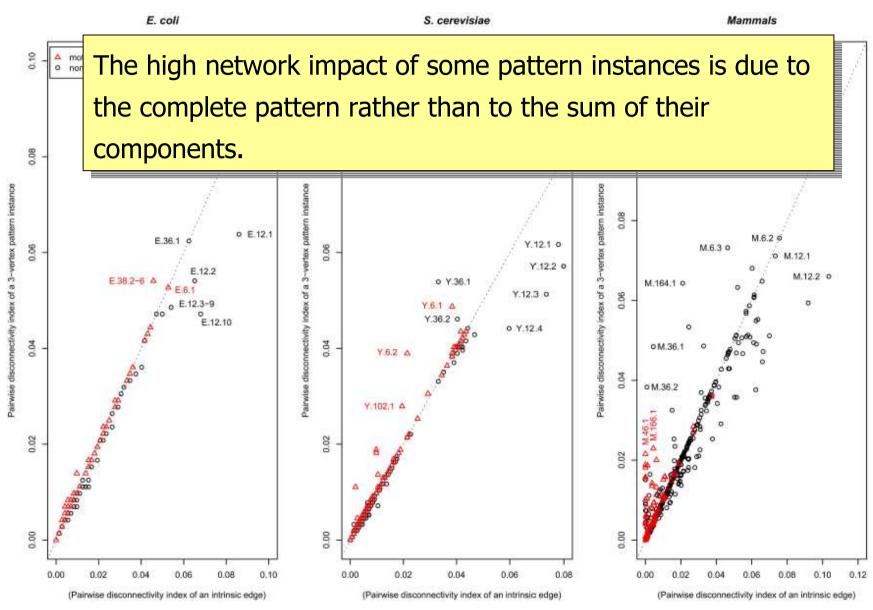
164	-	-	-	-	-	-	197	-6.52	0.0051
	-	-	-	1	4.65	0.0052	20	7.12	0.0058
174	-	-	-	-	-	-	6	7.31	0.0109
238	-	-	-	-	-	-	1	-	0.0073

Goemann et al., BMC Syst. Biol. 3, 53 (2009)





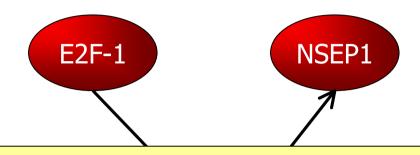




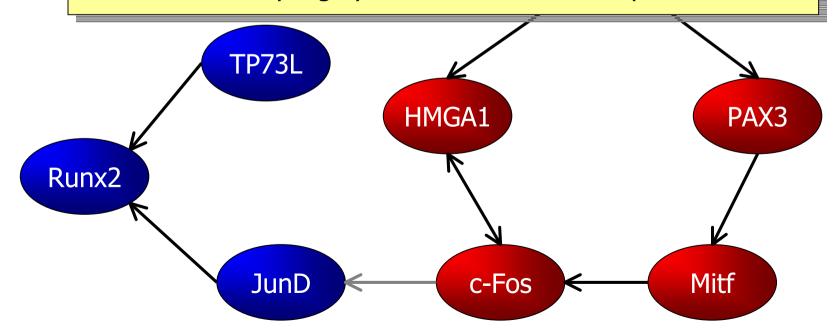
Goemann et al., BMC Syst. Biol. 3, 53 (2009)



# The largest mammalian subnetwork according to motif analysis is functionally linked with cell-cycle control



Topological (e.g., disconnectivity) analysis can reveal functionally highly relevant network components.



## **Conclusions (1)**



- Through the past 20 years, our understanding of the mode how transcription factors (TFs) operate has made remarkable progress.
- New HTP technologies have increased the chances to come up with a genome-wide map of TF binding sites (TFBS), which, however, will always be a kind of snapshot of a selection of cellular states.
- Context-dependent approaches to TFBS prediction that also make use of comparative genomics may provide a major breakthrough.

## **Conclusions (2)**



- While TRANSFAC provides the knowledge base for this, state-ofthe-art algorithms implemented in ExPlain™ reveal the syntax of promoters.
- The knowledge compiled in the TRANSFAC database may enable us to even predict the DNA-binding specificity of newly discovered TFs.
- Analyzing the architecture of the transcriptional network of cells will reveal key regulators that may be ideal candidates for diagnostic, therapeutic or biotechnological purposes.