

## V9 – 7. Protein-DNA contacts

- Transcription factors (TFs)
- Transcription factor binding sites (TFBS)
- Experimental detection of TFBS
- Position-specific scoring matrices (PSSMs)
- Binding free energy models
- Cis-regulatory motifs

Tue, May 15, 2018

# 7. DNA-binding proteins

DNA binding proteins include:

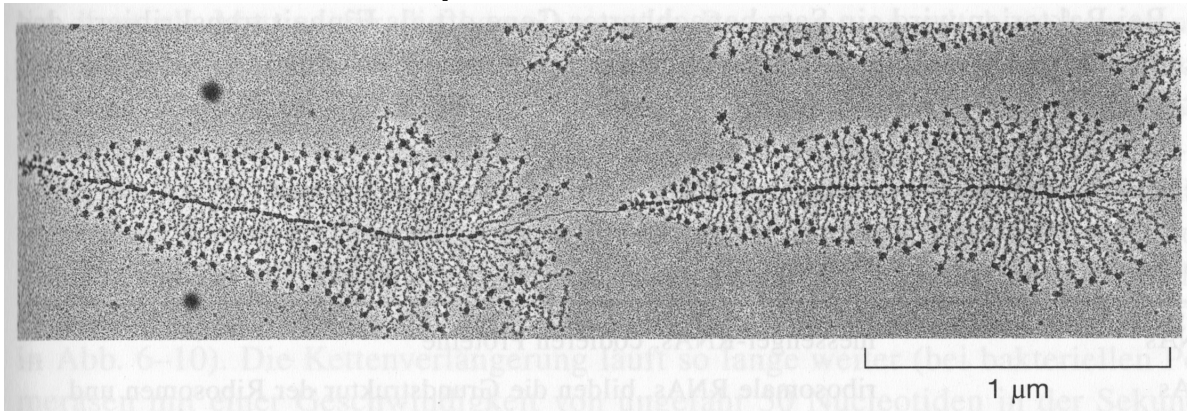
- TFs that activate or repress gene expression
- Enzymes involved in DNA repair
- Enzymes that place chemical (epigenetic) modifications on DNA
- Proteins that pack or unpack the chromatin structure
- Proteins that help to unzip double-stranded DNA
- DNA topoisomerases that are involved in DNA supercoiling etc.

From this long list, we will discuss here only TFs.

# Transcription Initiation

In eukaryotes:

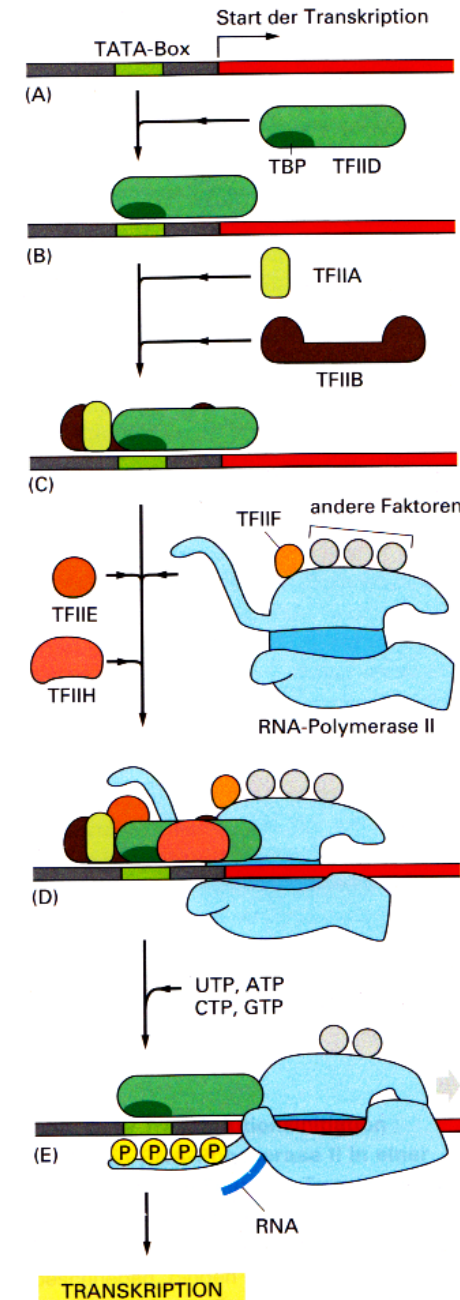
- several **general** transcription factors **have** to bind to gene promoter
- **specific** enhancers or repressors **may** bind
- then the RNA polymerase binds
- and starts transcription



Shown here: many RNA polymerases read central DNA at different positions and produce ribosomal rRNAs (perpendicular arms). The large particles at their ends are likely ribosomes being assembled.

Alberts et al.

"Molekularbiologie der Zelle", 4. Aufl.



## 7. Binding forces

There is generally **electrostatic attraction** between the negatively charged phosphate groups of the DNA backbone and positively charged amino acids on the protein surface.

This interaction involves only the DNA-backbone and is thus mostly independent from the DNA sequence.

**Attractive** contribution:

**specific** polar and non-polar **interactions** between the nucleotide bases of particular DNA sequence motifs and their protein binding partner.

# p53: example of a Protein-DNA-complex

PDB-structure 1TUP: tumor suppressor **p53**

Determined by X-ray crystallography

Purple (left): p53-protein (multiple copies)

Blue/red DNA double strand (right)

The protective action of the wild-type *p53* gene helps to suppress tumors in humans. The *p53* gene is the most commonly mutated gene in human cancer, and these mutations may actively promote tumor growth.

[www.sciencemag.org](http://www.sciencemag.org) (1993)



[www.rcsb.org](http://www.rcsb.org)

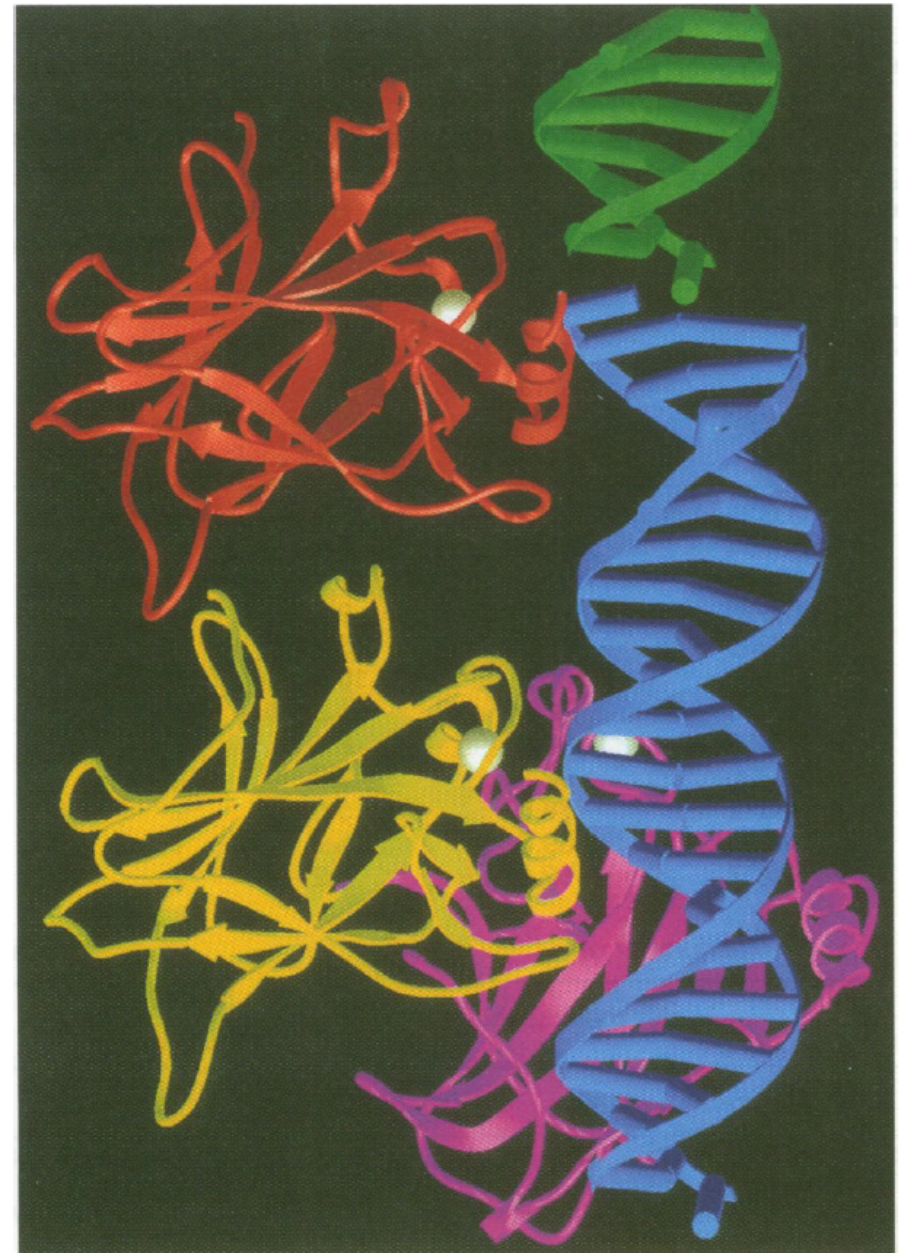


# Contacts establish specific binding mode



Nikola Pavletich,  
Sloan Kettering  
Cancer Center

**Fig. 3.** Schematic ribbon drawing of the asymmetric unit, which contains three p53 core domain molecules and one DNA duplex. Two of the core domains bind DNA (blue); one (yellow) interacts extensively with a consensus binding site, and the other (red) binds at a nonconsensus site at the interface of DNA fragments related by crystallographic symmetry (a portion of the symmetry-related DNA fragment is shown in green). The third core domain molecule (purple) does not bind DNA, but makes protein-protein contacts stabilizing crystal packing. The zinc atoms are shown as white spheres.

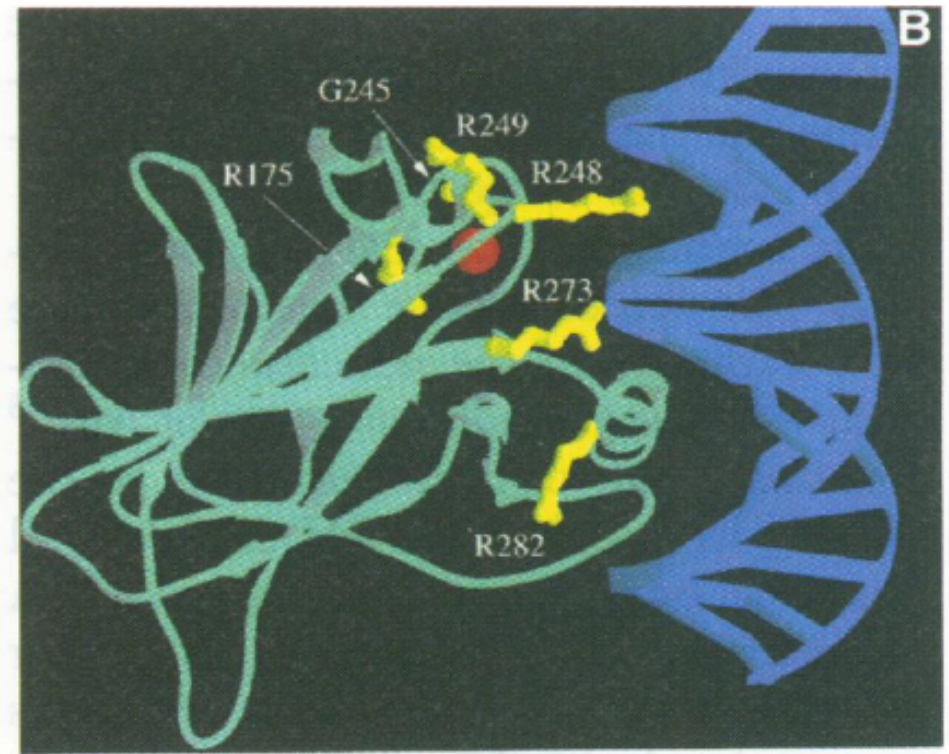
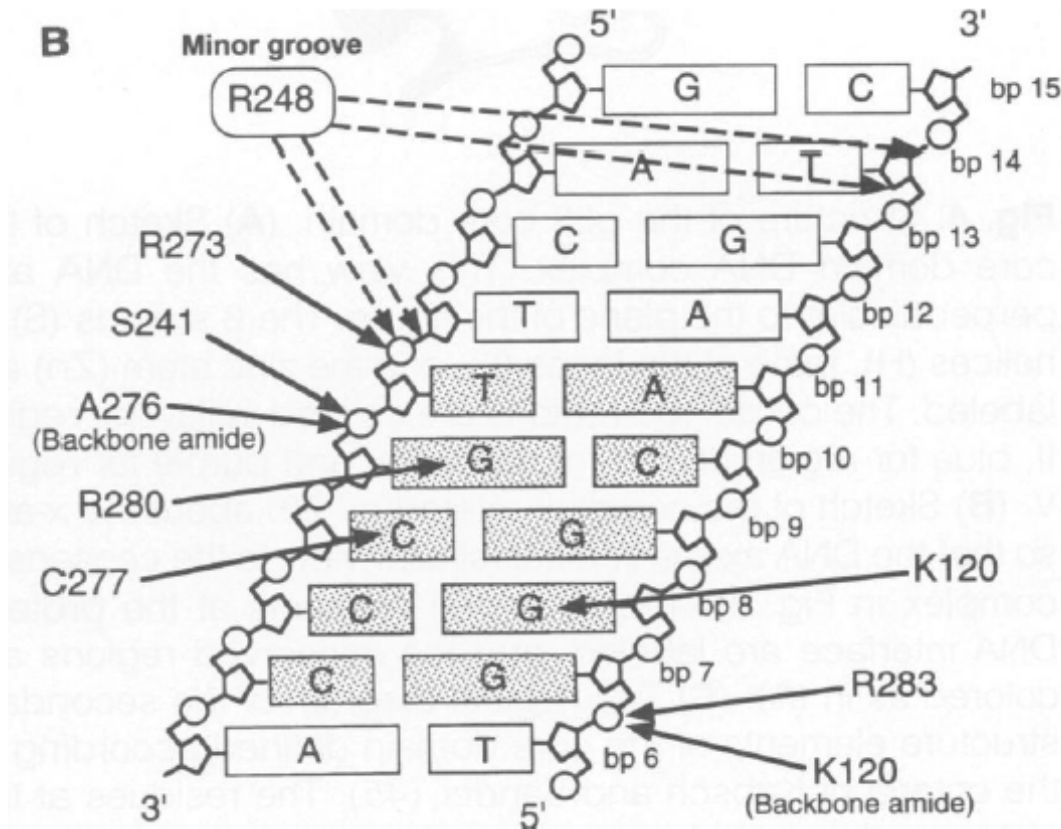


## RESEARCH ARTICLE

### Crystal Structure of a p53 Tumor Suppressor–DNA Complex: Understanding Tumorigenic Mutations

Yunje Cho, Svetlana Gorina, Philip D. Jeffrey, Nikola P. Pavletich

# Contact residues



Left: Protein – DNA contacts involve many arginine (R) and lysine (K) residues

Right: the 6 most frequently mutated amino acids (yellow) in cancer.  
5 of them are Arginines.

In p53 all 6 residues are located at the binding interface for DNA!

Science 265, 346-355 (1994)



# Structural view at *E. coli* TFs

Determine homology between the domains and protein families of TFs and regulated genes and proteins of known 3D structure.

→ Determine uncharacterized *E.coli* proteins with DNA-binding domains (DBD)

→ identify large majority of *E.coli* TFs.



Sarah Teichmann  
EBI



Madan Babu,  
MRC

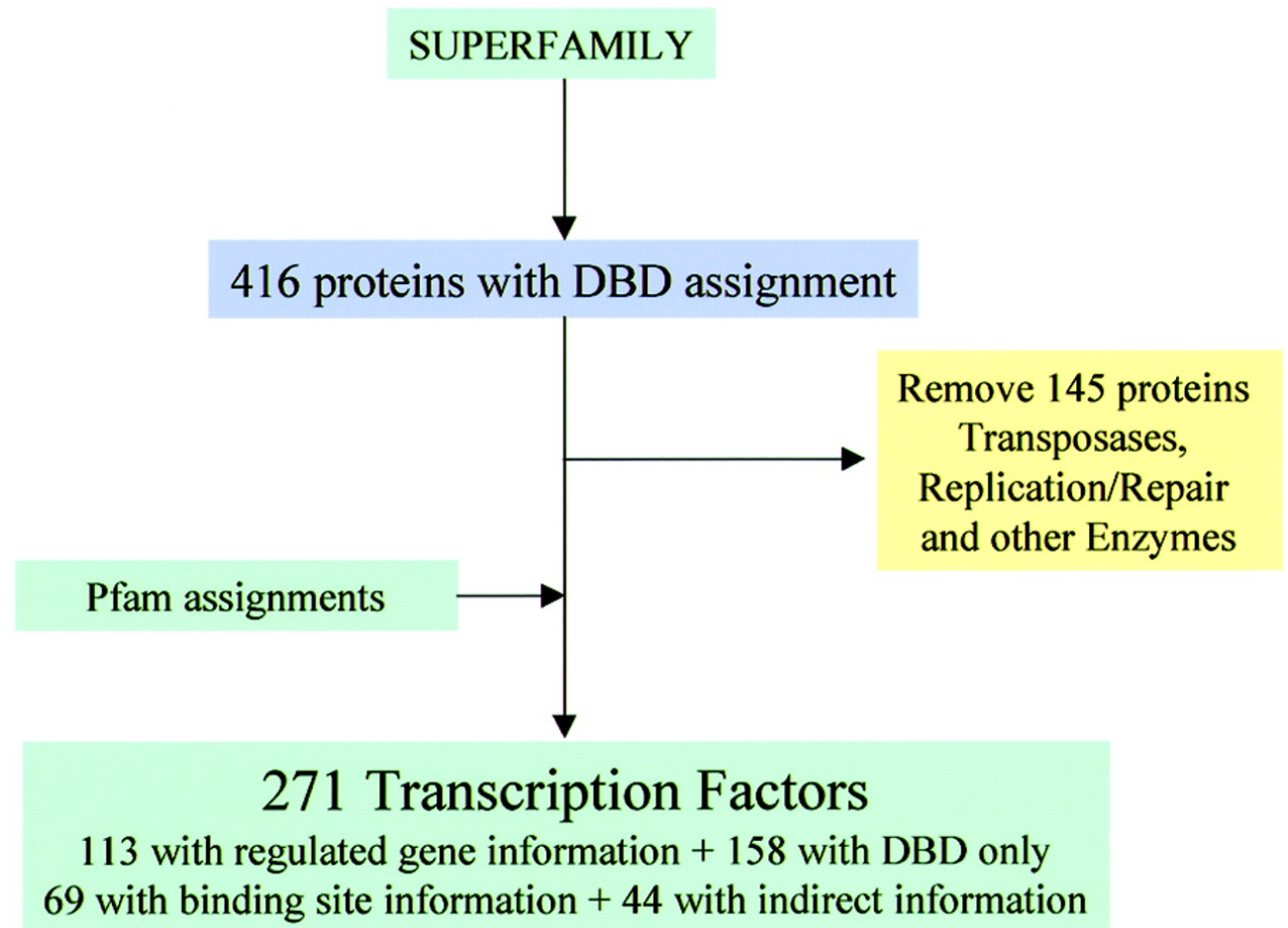
Babu, Teichmann, Nucl. Acid Res. 31, 1234 (2003)



# Flow chart of method to identify TFs in E.coli

SUPERFAMILY database (C. Chothia) contains a library of HMM models based on the sequences of proteins in SCOP for predicted proteins of completely sequenced genomes.

Remove all DNA-binding proteins involved in replication/repair etc.



Babu, Teichmann, Nucl. Acid Res. 31, 1234 (2003)

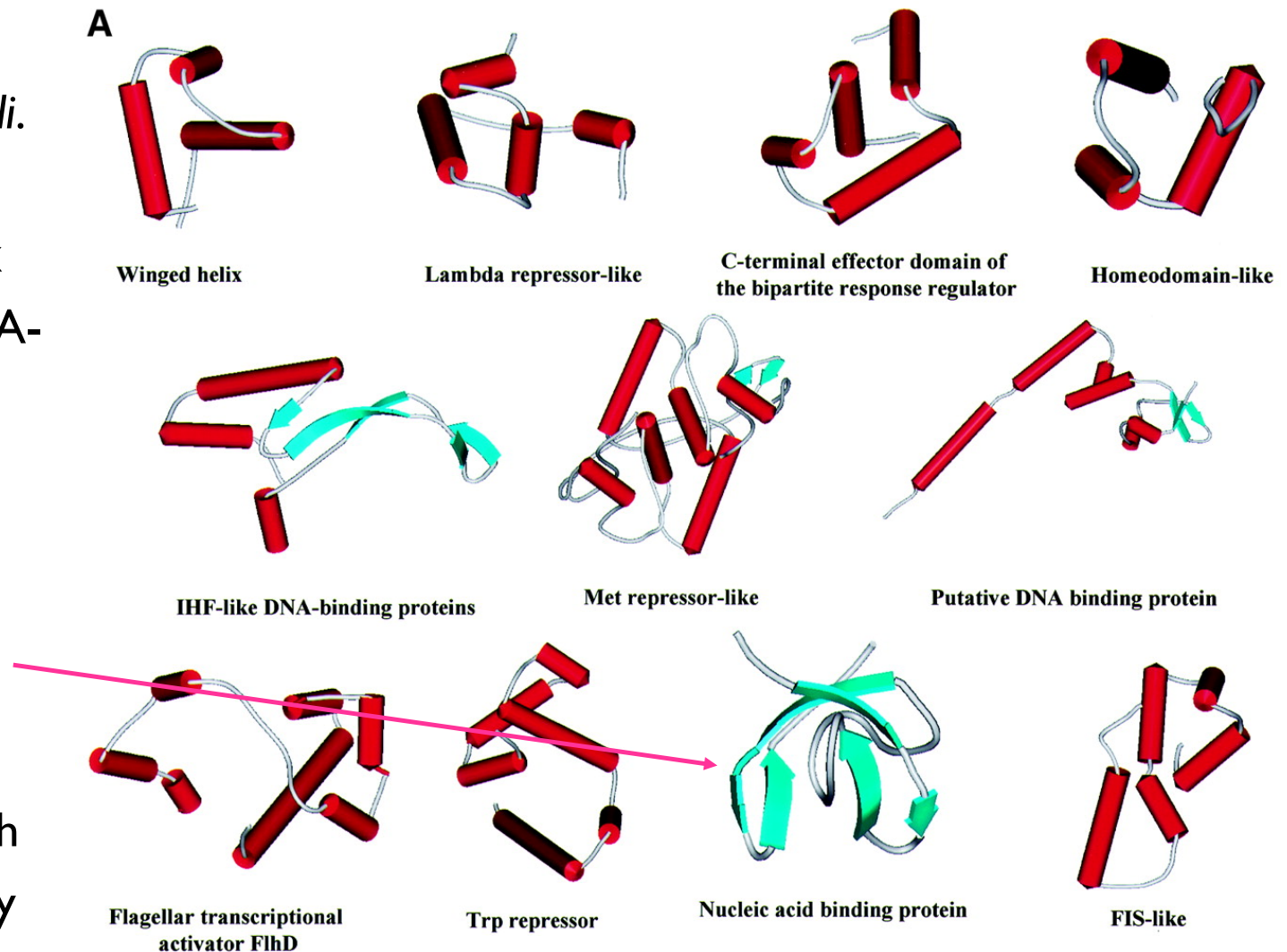
# 3D structures of putative (and real) TFs in *E.coli*

3D structures of the 11 DBD families seen in the 271 identified TFs in *E.coli*.

The **helix–turn–helix motif** is typical for DNA-binding proteins.

It occurs in all families except the nucleic acid binding family.

Still the scaffolds in which the motif occurs are very different.



Babu, Teichmann, Nucl. Acid Res. 31, 1234 (2003)

# Domain architectures of TFs

The 74 unique domain architectures of the 271 TFs.

The **DBDs** are represented as rectangles.



The partner domains are represented as

hexagons (**small molecule-binding domain**),



triangles (**enzyme** domains),



circles (protein interaction domain),



diamonds (domains of unknown function).



The receiver domain has a pentagonal shape.

A, R, D and U stand for activators, repressors, dual regulators and TFs of unknown function.

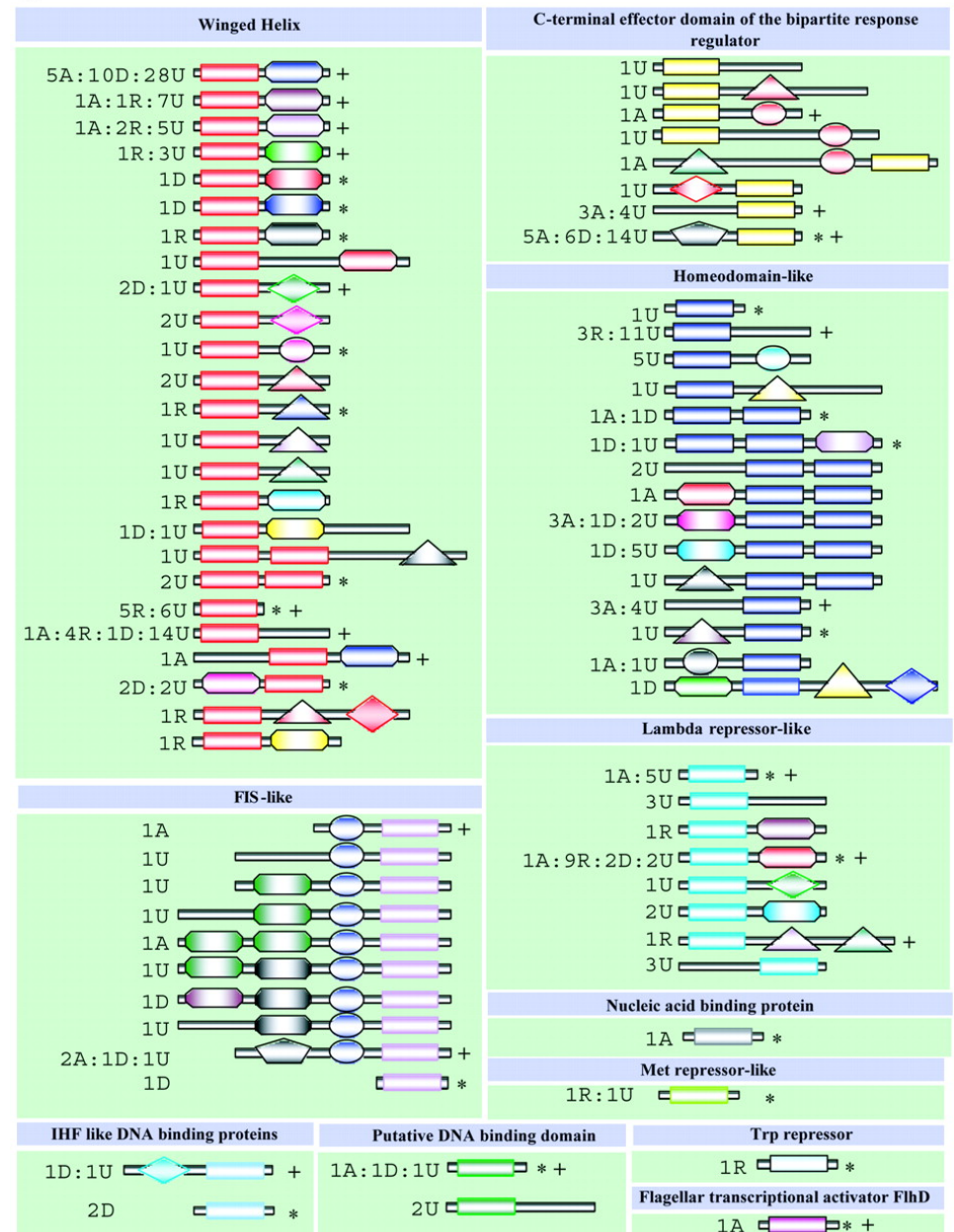
The number of TFs of each type is given next to each domain architecture.

Architectures of known 3D structure are denoted by asterisks.

‘+’ are cases where the regulatory function of a TF has been inferred by indirect methods, so that the DNA-binding site is not known.

Babu, Teichmann, Nucl. Acid Res. 31, 1234 (2003)

B



# Evolution of TFs

- 10% 1-domain proteins
- 75% 2-domain proteins
- 12% 3-domain proteins
- 3% 4-domain proteins

TFs have evolved by apparently extensive recombination of domains.

Proteins with the same sequential arrangement of domains are likely to be direct duplicates of each other.

74 distinct domain architectures have duplicated to give rise to 271 TFs.

Babu, Teichmann, Nucl. Acid Res. 31, 1234 (2003)



# Evolution of the gene regulatory network

Table 1

Numbers of DNA-binding transcription factors in five organisms<sup>a</sup>.

Organism	Number of transcripts	Number of proteins with DNA-binding domains	Percentage of transcripts containing DNA-binding domains
<i>E. coli</i>	4280	267	6.2
<i>S. cerevisiae</i>	6357	245	3.9
<i>C. elegans</i>	31 677	1463	4.6
<i>H. sapiens</i>	32 036 <sup>b</sup>	2604	8.1
<i>A. thaliana</i>	28 787	1667	5.7

<sup>a</sup>DNA-binding domain assignments from Pfam and SUPERFAMILY are used to establish the repertoire of DNA-binding transcription factors in five model organisms. An expectation value threshold of 0.002 was used in making the assignments. Co-regulators that do not bind DNA directly are excluded. <sup>b</sup>Predicted by Ensembl v19.34a [42].

Most genomes contain hundreds up to a few thousands of TFs.

Larger genomes tend to have more TFs per gene.

Babu et al. Curr Opin Struct Biol. 14, 283 (2004)

# Transcription factors in yeast *S. cerevisiae*

*Q: How can one define transcription factors?*

Hughes & de Boer consider as TFs proteins that

- (a) bind DNA directly and in a sequence-specific manner and
- (b) function to regulate transcription nearby sequences they bind

*Q: Is this a good definition?*

Yes. Only 8 of 545 human proteins that bind specific DNA sequences and regulate transcription lack a known DNA-binding domain (DBD).

# Transcription factors in yeast

Hughes and de Boer list 209 known and putative yeast TFs.

The vast majority of them contains a canonical DNA-binding domain.

## Most abundant:

- GAL4/zinc cluster domain (57 proteins),  
largely specific to fungi (e.g. yeast)
- zinc finger C2H2 domain (41 proteins),  
most common among all eukaryotes.



## Other classes :

- bZIP (15),
- Homeodomain (12),
- GATA (10), and
- basic helix-loop-helix (bHLH) (8).

# TFs of *S. cerevisiae*

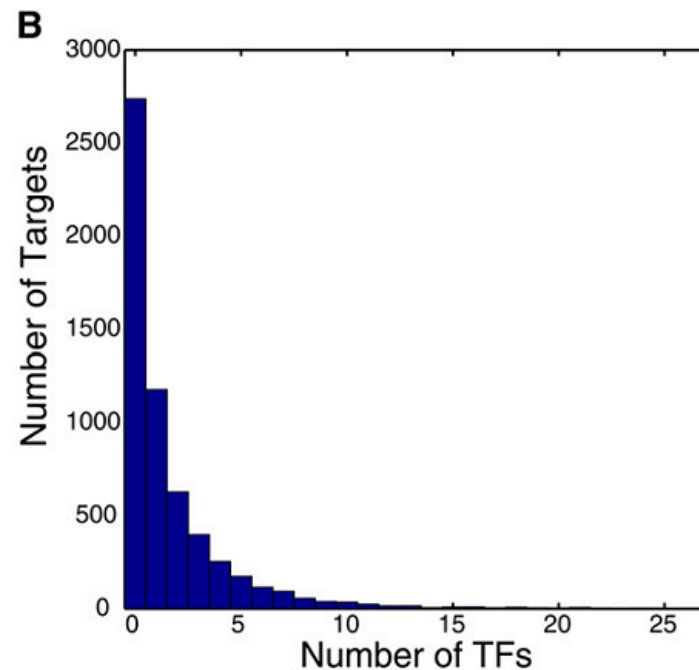
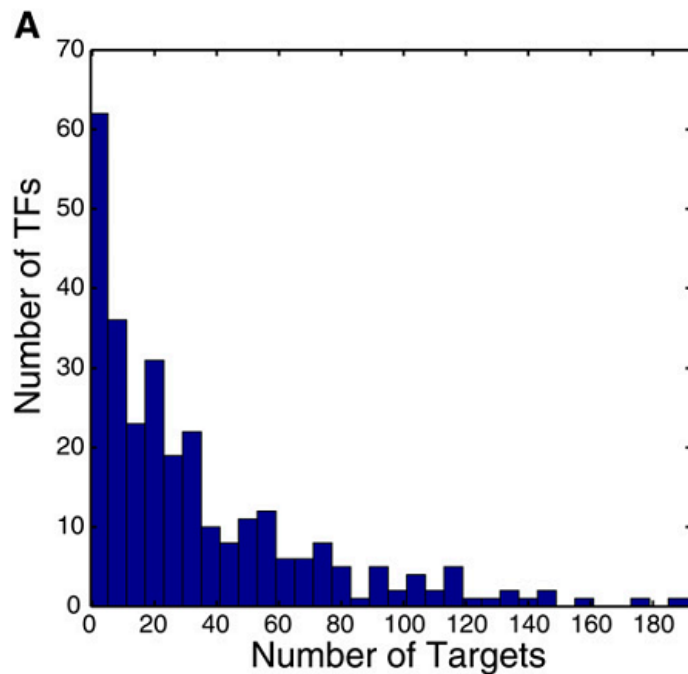
(A) Most TFs tend to bind relatively few targets.

57 out of 155 unique proteins bind to  $\leq 5$  promoters in at least one condition.

17 did not significantly bind to any promoters under any condition tested.

In contrast, several TFs have hundreds of promoter targets.

These TFs include the general regulatory factors (GRFs), which play a global role in transcription under diverse conditions.



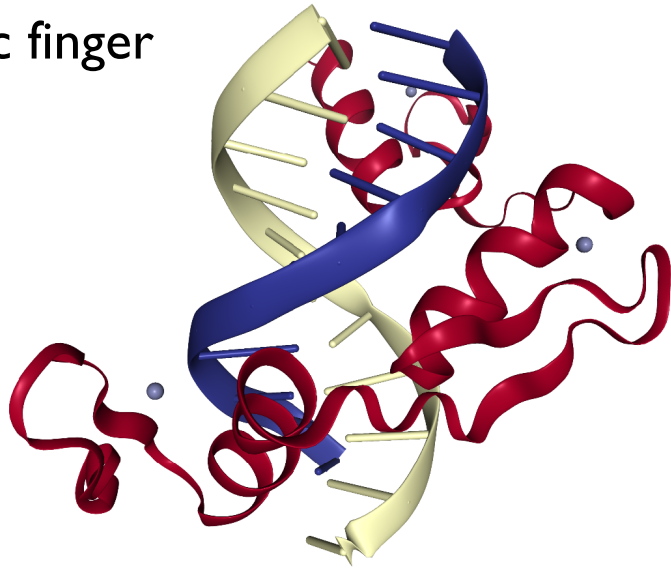
(B) # of TFs that  
bind to  
one promoter.

Hughes, de Boer  
(2013) Genetics 195,  
9-36

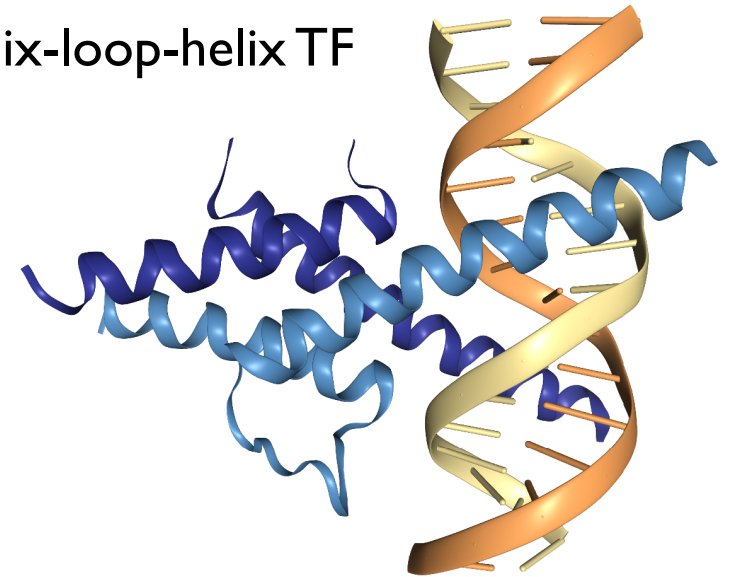


## 7.1 Structural types of TFs

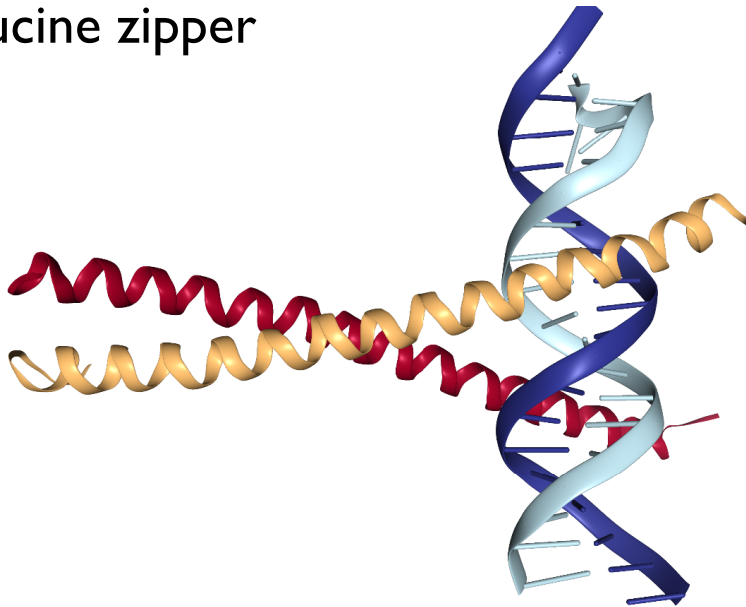
Zinc finger



Helix-loop-helix TF



Leucine zipper



High mobility group TF



## 7.2 Transcription factor binding sites (TFBSs)

TFBS: DNA region that forms a **specific physical contact** with a particular TF.

TFBS are usually between 8 and 20 bp long  
and contain a 5-8 bp long **core region** of well-conserved nucleotide bases.

Most TFs bind in the **major groove** of double-stranded DNA,  
the others bind in the minor groove.

The **periodicity** of double-stranded DNA is around 10 bp.  
Thus, the core regions of TFBS are a bit longer than half a turn of dsDNA.

TFs may recognize DNA sequences that are similar,  
but not identical, differing by a few nucleotides.

# Sequence logos represent binding motifs

A **logo** represents each column of the alignment by a stack of letters.

The height of each letter is proportional to the **observed frequency** of the corresponding amino acid or nucleotide.

The overall height of each stack is proportional to the **sequence conservation** at that position.

**Sequence conservation** is defined as difference between the maximum possible entropy and the entropy of the observed symbol distribution:

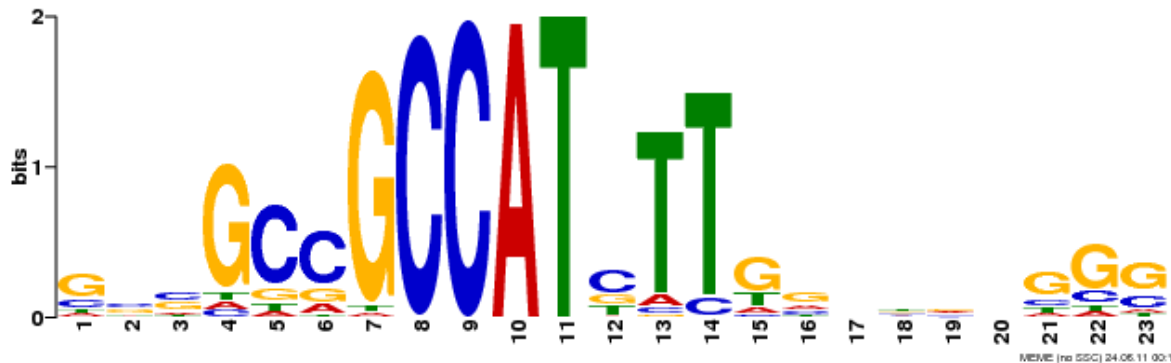
$$R_{seq} = S_{max} - S_{obs} = \log_2 N - \left( - \sum_{n=1}^N p_n \log_2 p_n \right)$$

$p_n$  : observed frequency of symbol  $n$  at a particular sequence position

$N$  : number of distinct symbols for the given sequence type, either 4 for DNA/RNA or 20 for protein.

# YY1 sequence logo

Sequence-logos are a convenient way to visualize the degree of degeneracy in the TFBS.



Sequence logo for the DNA binding motif that the TFYY1 (Yin Yang I) binds to.

The motif was derived from the top 500 TF ChIP-seq peaks by the ENCODE consortium. For YY1, 468 out of 500 sequences contained this motif.

Figure from Factorbook repository (Wang et al. 2013).

$$H_i = - \sum_{b=A}^T f_{b,i} \times \log_2 f_{b,i}$$
$$R_i = \log_2(4) - (H_i + e_n)$$
$$e_n = \frac{1}{\ln 2} \times \frac{s - 1}{2n}$$

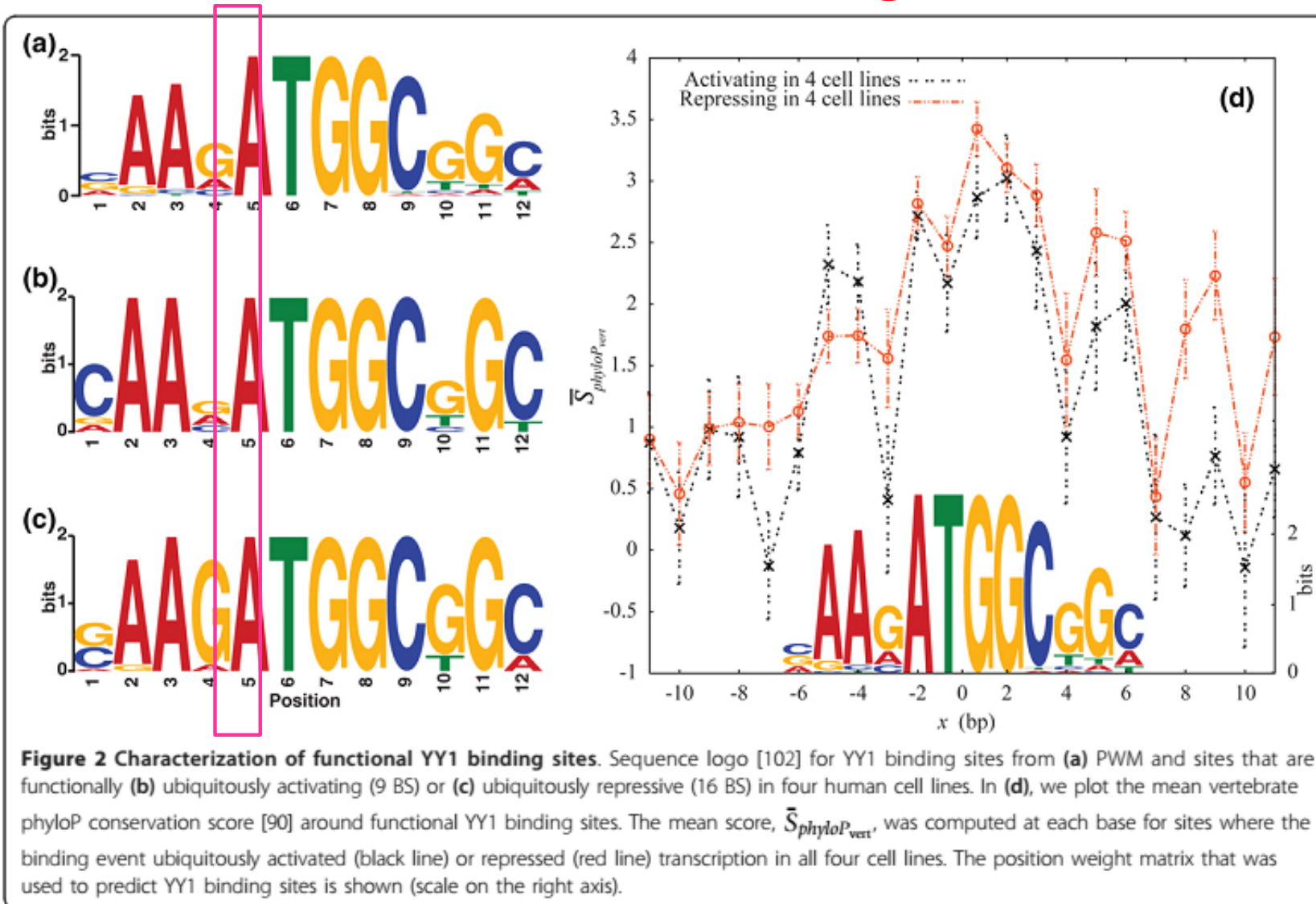
$H_i$  : uncertainty (Shannon entropy) of position  $i$

$R_i$  : information content (y-axis) of position  $i$

$e_n$  : small-sample correction,  
 $s = 4$  for nucleotides,  $n$  :  
number of sequences

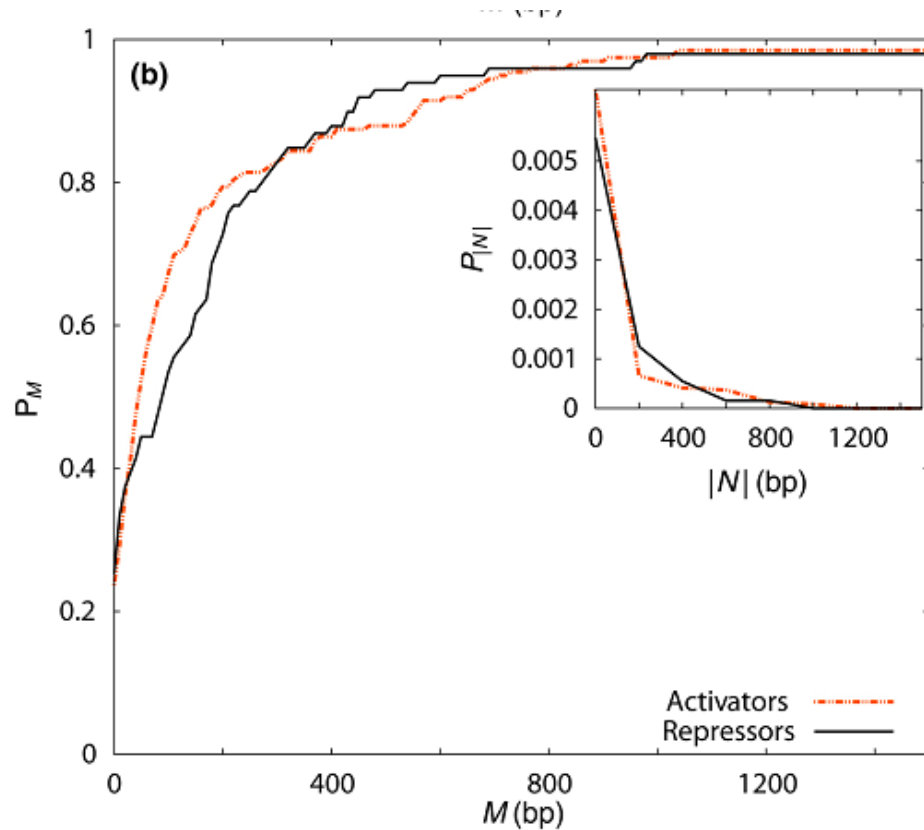


# YY1 binding motifs



No noticeable difference in motifs of activated or repressed target genes.

# Where are TFBS relative to the TSS?



Inset: probability to find binding site at position  $N$  from transcriptional start site (TSS)

Main plot: cumulative distribution.

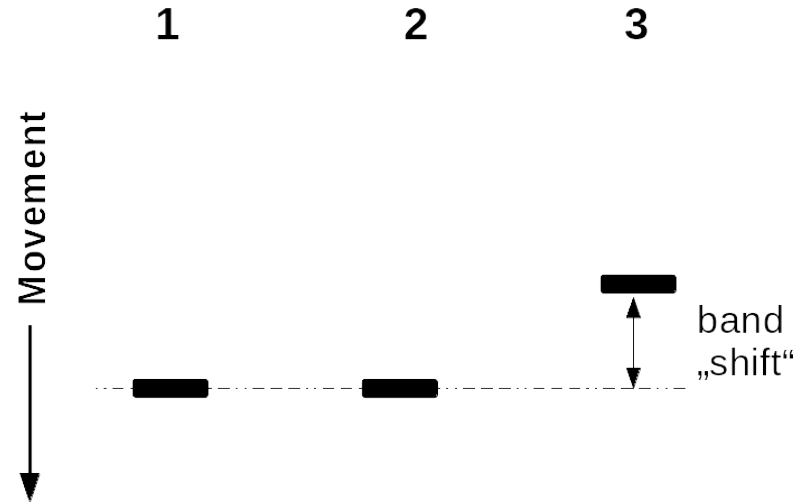
Activating TF binding sites are closer to the TSS than repressing TF binding sites ( $p = 4.7 \times 10^{-2}$ ).

## 7.3 Experimental TFBS detection: EMSA shift assay

An **electrophoretic mobility shift assay** (EMSA) or **gel shift assay** is an affinity electrophoresis technique for identifying **specific binding** of a protein–DNA or protein–RNA pair **in vitro**.

The samples are electro-phoretically separated on a polyacrylamide or agarose gel.

The results are visualized by radioactive labelling of the DNA with  $^{32}\text{P}$  or by tagging a fluorescent dye.



Control lane (1) contains DNA probe without protein. Obtained at the end of the experiment is a single band that corresponds to the unbound DNA.

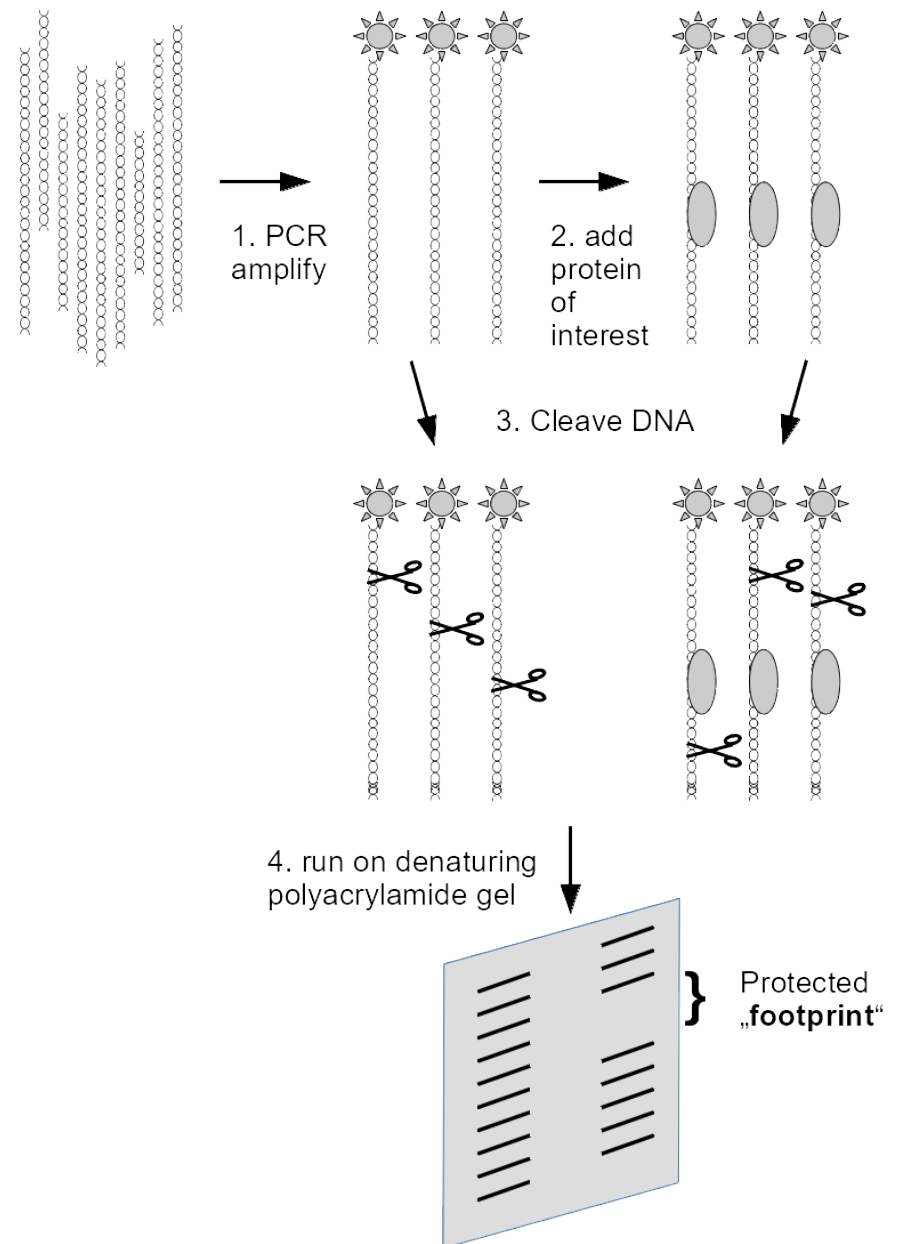
Lanes (2) and (3) each contain a mixture of the DNA with a protein. If the protein actually binds to the DNA (3), this lane will show an up-shifted band relative to (1) which is due to the larger and less mobile protein:DNA complex.

## 7.3.2. DNase footprinting

In DNase footprinting, a **DNase enzyme** is added to the sample that **cleaves** DNA non-specifically at many positions.

On a polyacrylamide gel, the cleaved DNA fragments of differing lengths will show up as different lanes (left figure).

In a second experiment, the protein of interest is added (right lane). If this protein binds specifically at a particular position of the DNA, it will prevent cleavage by DNase at this position. Then, this DNA fragment cannot be found on the gel (bottom, right lane) and represents thus the specific binding motif in the investigated DNA sequence for the protein.





### 7.3.3. High-throughput methods

There exist also several high-throughput *in vitro* methods to measure the TF-DNA binding affinity of large numbers of DNA variants.

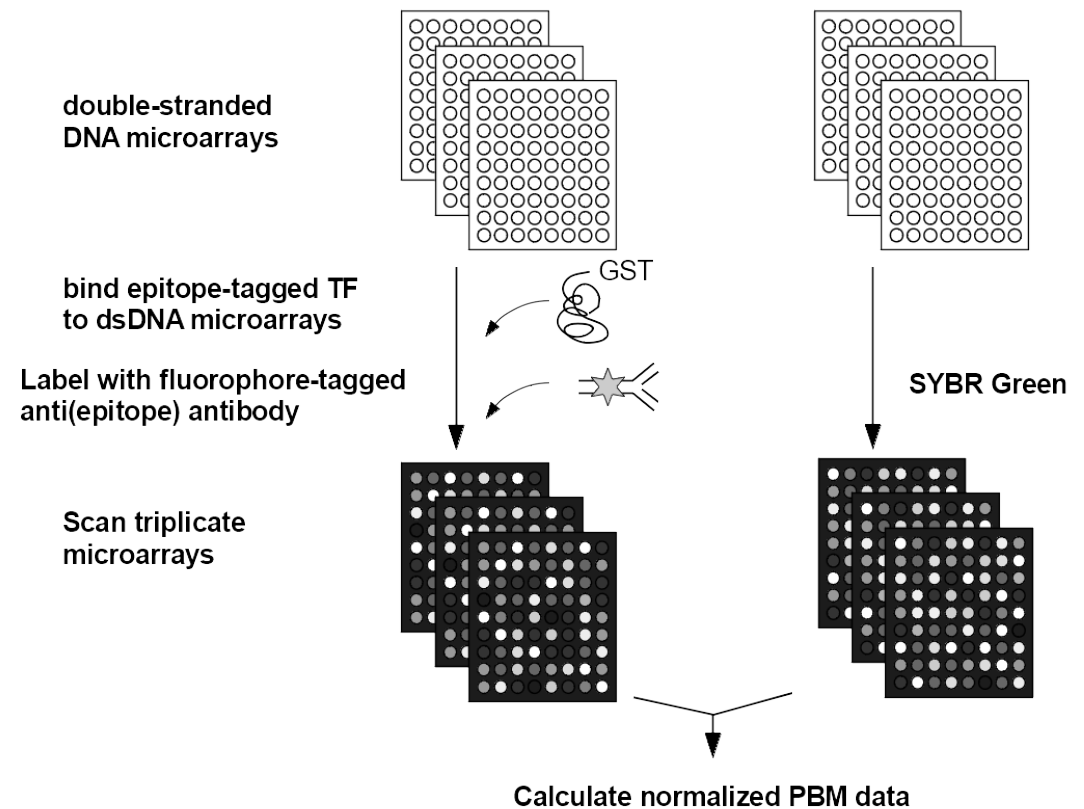
One of them is a DNA microarray-based method called protein binding microarray (PBM) (Berger and Bulyk, 2006).

With this technology, one can characterize the binding specificity of a single DNA binding protein *in vitro* by adding it to the wells of a microarray spotted with a large number of putative binding sites in double-stranded DNA.

### 7.3.3. Protein binding microarray

The protein of interest carrying an epitope tag is expressed and purified and then applied to the microarray.

After removing nonspecifically bound protein by a washing step, the protein is detected in a labeling step where a fluorophore-conjugated antibody binds specifically to the epitope tag.



One identifies all spots carrying a significant amount of protein.

In the DNA sequences belonging to these spots, one identifies enriched DNA binding site motifs for the DNA binding protein of interest.

### 7.3.3. Problems of in vitro methods

Due to the short length of TFBS motifs and the relatively small number of invariant nucleotide positions in it, some motifs are found millions of times in the genome.

Thus, although any motif instance could potentially be bound *in vivo*, only about 1 in 500 are actually bound in organisms with large genomes.

As a specific example, the mouse genome contains ~8 million instances of a match to the binding site motif of **GATA-binding factor 1**, but only ~15,000 DNA segments are bound by this transcription factor in erythroid cells (Hardison and Taylor, 2012).

### 7.3.3. *in vivo* methods

To overcome the limitations of *in vitro* assays, new massively parallel methods such as ChIP-chip and ChIP-seq can identify TF binding sites *in vivo*.

These methods are based on DNA microarrays and new sequencing techniques, respectively.

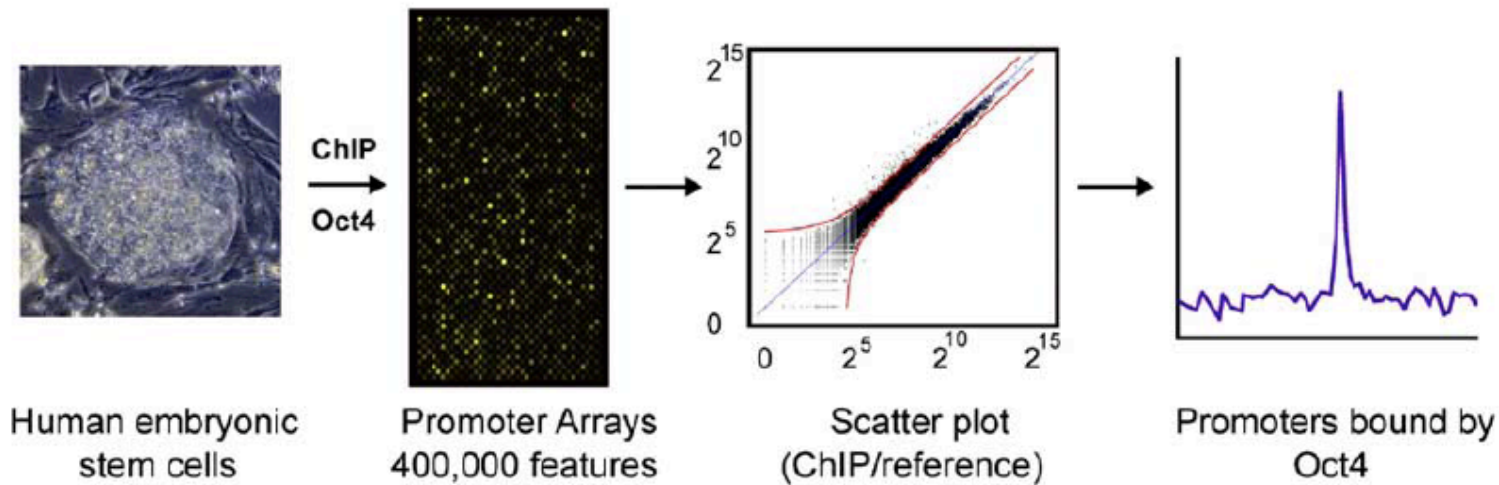
In **Chip-seq** experiments, a cellular extract is purified using an antibody against a particular TF.

Then, the DNA sequences bound to the TF are digested using a restriction enzyme. The remaining DNA can be considered as tightly bound to the TF.

This DNA is washed and sequenced.

All DNA reads correspond to DNA fragments that were bound to the TF before.

# Which TF binds where?



Chromatin immuno precipitation: use e.g. antibody against Oct4

- ➔ "fish" all DNA fragments that bind Oct4
- ➔ sequence DNA fragments bound to Oct4
- ➔ align them + extract characteristic sequence features
- ➔ Oct4 binding motif

Boyer et al. Cell 122, 947 (2005)

## 7.4. Position-specific scoring matrix

PSSMs are used to represent motifs (patterns) in biological sequences.

	Position 1	Position 2	Position 3	Position 4
Sequence 1	A	C	A	T
Sequence 2	A	C	C	T
Sequence 3	A	G	G	G
Sequence 4	C	C	T	G
Sequence 5	A	T	A	G
Sequence 6	C	A	G	T

Toy example of six DNA sequences that are 4 bp long.

	Position 1	Position 2	Position 3	Position 4
Frequency A	4	1	2	0
Frequency C	2	3	1	0
Frequency G	0	1	2	3
Frequency T	0	1	1	3

Frequency  $n_i^j$  of nucleotide bases ( $i$ ) at the 4 positions ( $j$ ).

Out of  $6 \times 4 = 24$  nucleotides in the four sequences, 7 are adenine, 6 are cytosine, 6 are guanine, and 5 are thymine. Thus, the frequencies  $p_i$  of the four nucleotides are 0.29 (A), 0.25 (C and G), and 0.21 (T).



## 7.4. Position-specific scoring matrix

From the frequency matrix, one computes the score matrix using

$$s_i^j = \ln \frac{(n_i^j + p_i)/(N+1)}{p_i},$$

where,  $N$  is the number of considered sequences (here,  $N = 6$ ).

	Position 1	Position 2	Position 3	Position 4
score A	0.75	-0.45	0.12	-1.94
score C	0.25	0.62	-0.34	-1.94
score G	-1.94	-0.34	0.25	0.62
score T	-1.94	-0.19	-0.19	0.78

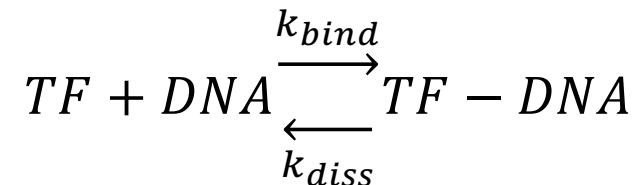
Adding the frequencies  $p_i$  in the denominator and dividing by  $N + 1$  avoids problematic cases with  $n_i^j = 0$  where the logarithm would not be defined otherwise.

Positions with score  $s_i^j = 0$  occur at the frequency that is expected randomly, positive entries denote enriched nucleotides at this position, negative entries denote the opposite case.

## 7.5 Binding free energy models

The binding of a TF to single- or double-stranded DNA is an elementary biomolecular association reaction.

The binding free energy model of Djordjevic (2003) describes the reversible binding of a TF to a short piece of DNA with sequence  $S$ ,



with the sequence-dependent rate constants  $k_{bind}$  and  $k_{diss}$  for TF binding and dissociation, respectively.

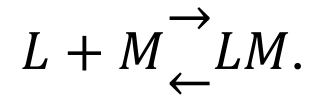
In equilibrium,  $[TF] \cdot [S] \cdot k_{bind}(S) = [TF:S] \cdot k_{diss}(S)$

The ratio of the bound and free forms thus equals the ratio of the two rate

constants and is equal to  $\frac{[TF:S]}{[TF][S]} = \frac{k_{bind}(S)}{k_{diss}(S)} = \frac{1}{K_D} = c \cdot e^{-\frac{\Delta G(S)}{kT}}$ , where  $c$  is a constant and  $\Delta G(S)$  is the (usually negative) binding free energy of the TF to its recognition sequence  $S$  on the DNA.

## 7.5 Binding free energy models

Let us consider the binding reaction of two molecules  $L$  and  $M$ :



The dissociation equilibrium constant  $K_D$  is defined as:

$$K_D = \frac{[L][M]}{[LM]} = \frac{k_{diss}}{k_{bind}}$$

, where  $[L]$ ,  $[M]$ , and  $[LM]$  are the molecular concentrations of  $L$  and  $M$  and of the complex  $LM$ .

In equilibrium, we may take  $T$  as the total concentration of molecule  $L$

$$T = [L] + [LM].$$

$y$  is the fraction of molecules  $L$  that have reacted (bound),

$$y = \frac{[LM]}{[LM] + [L]}$$

.

## 7.5 Binding free energy models

$$y = \frac{[LM]}{[LM] + [L]}$$

Substituting  $[LM]$  by  $[L][M] / K_D$  gives

$$y = \frac{([L][M])/K_D}{([L][M])/K_D + [L]} = \frac{([M])/K_D}{([M])/K_D + 1}$$

When a solution contains both the DNA sequence and the TF with total concentration  $n_{tf}$ , the equilibrium probability that the DNA is bound to a TF molecule is (replace in upper eq.  $[M]$  by  $n_{tf}$ ):

$$p(TF \text{ is bound to } S) = \frac{\frac{1}{K_D} \cdot n_{tf}}{\frac{1}{K_D} \cdot n_{tf} + 1} = \frac{c \cdot e^{-\Delta G(S)/kT} \cdot n_{tf}}{c \cdot e^{-\Delta G(S)/kT} \cdot n_{tf} + 1}$$

We multiply this with  $e^{+\Delta G(S)/kT}$  and divide by  $c \cdot n_{tf}$ .

## 7.5 Binding free energy models

This gives:  $P(TF \text{ is bound to } S) = \frac{1}{1 + \frac{e^{\Delta G(S_i)/kT}}{c \cdot n_{tf}}}$ ,

where  $\Delta G(S_i)$  : free energy of the TF binding to  $S_i$  .

We set  $c \cdot n_{tf} = e^{\frac{\mu}{kT}}$  or  $\mu = kT \cdot \ln(c \cdot n_{tf})$

$\mu$  : chemical potential set by the TF concentration. This gives

$$P(TF \text{ is bound to } S) = \frac{1}{1 + e^{(\Delta G(S_i) - \mu)/kT}}$$

,

This is the so-called Fermi-Dirac form of binding probability.

A sequence having a binding free energy well below the chemical potential ( $\Delta G(S_i) - \mu \ll 0$ ) is almost always bound to the TF.

( $P(TF \text{ is bound to } S) \rightarrow 1$  because the exponential term is very small.)

In cases when the binding free energy is well above the chemical potential, the sequence is rarely bound.

## 7.5 Binding free energy models

The binding energy model (BEM) uses a vector of (free) energy contributions,  $\vec{E}$ .

For any sequence  $S_i$ , the binding energy predicted by the BEM model is

$$E(S_i) = \vec{E} \cdot \vec{S}_i$$

where  $\vec{S}_i$  is the vector encoding of sequence  $S_i$  that can include whatever features of the sequence are relevant to its binding energy.

If the only relevant features are which bases occur at each position within the binding site, then  $\vec{E}$  will be a PSSM with the characteristic that each element is a (free) energy contribution.



## 7.5 Binding free energy models

When the (free) energy contributions of each position are independent,  $\vec{E} \cdot \vec{S}_i$  can be written as:

$$E(S_i) = \sum_{b=A}^T \sum_{m=1}^L \epsilon(b, m) S_i(b, m)$$

where  $L$  : length of the binding site,  $\epsilon(b, m)$  : (free) energy contributions of base  $b$  at position  $m$ , and  $S_i(b, m)$ : indicator variable with  $S_i(b, m) = 1$  if base  $b$  occurs at position  $m$  of sequence  $S_i$  and  $S_i(b, m) = 0$  otherwise.

If the positions are not independent, one can include pairwise interactions between adjacent positions  $m$  and  $n$  by adding **interaction terms** to the energy function such that  $\vec{E} \cdot \vec{S}_i$  is

$$E(S_i) = \sum_{b=A}^T \sum_{m=1}^L \epsilon(b, m) S_i(b, m) + \sum_{m=1}^{L-1} \sum_{n=m+1}^L \sum_{b=A}^T \sum_{c=A}^T \epsilon(b, m, c, n) S_i(b, m, c, n)$$

where  $\epsilon(b, m, c, n)$  : energy contribution of having base  $b$  at position  $m$  and base  $c$  at position  $n$ .

## 7.6 Cis-regulatory motifs

Although hundreds of TFs are present in a typical eukaryotic cell, the complex expression patterns of thousands of genes can only be implemented by a regulatory machinery involving combinations of TFs.

Thus, prokaryotic and eukaryotic gene promoters often bind multiple TFs simultaneously.

These TFs may also make structural contacts to each other and thus affect their mutual binding affinities in a cooperative manner.

In that case, for steric reasons, the distance between TFBSs of contacting TFs is constrained to a certain range.

All such combinatorial and cooperative effects are difficult to capture in a quantitative manner by a PSSM-based approach.

## 7.6 Cis-regulatory motifs

A **cluster** of TFBSs is termed a **cis-regulatory module** (CRM).

The existence of such a CRM is a footprint of a **TF complex**.

For metazoans, a typical CRM may be more than 500 bp long and is made up of 10 to 50 TFBSs to which between three and 15 different sequence-specific TFs bind.

If there exist multiple similar binding sites, this

- enhances the sensitivity for a TF,
- results in a more robust transcriptional response and
- affects how morphogen TFs are activated when the local TF concentration is low,

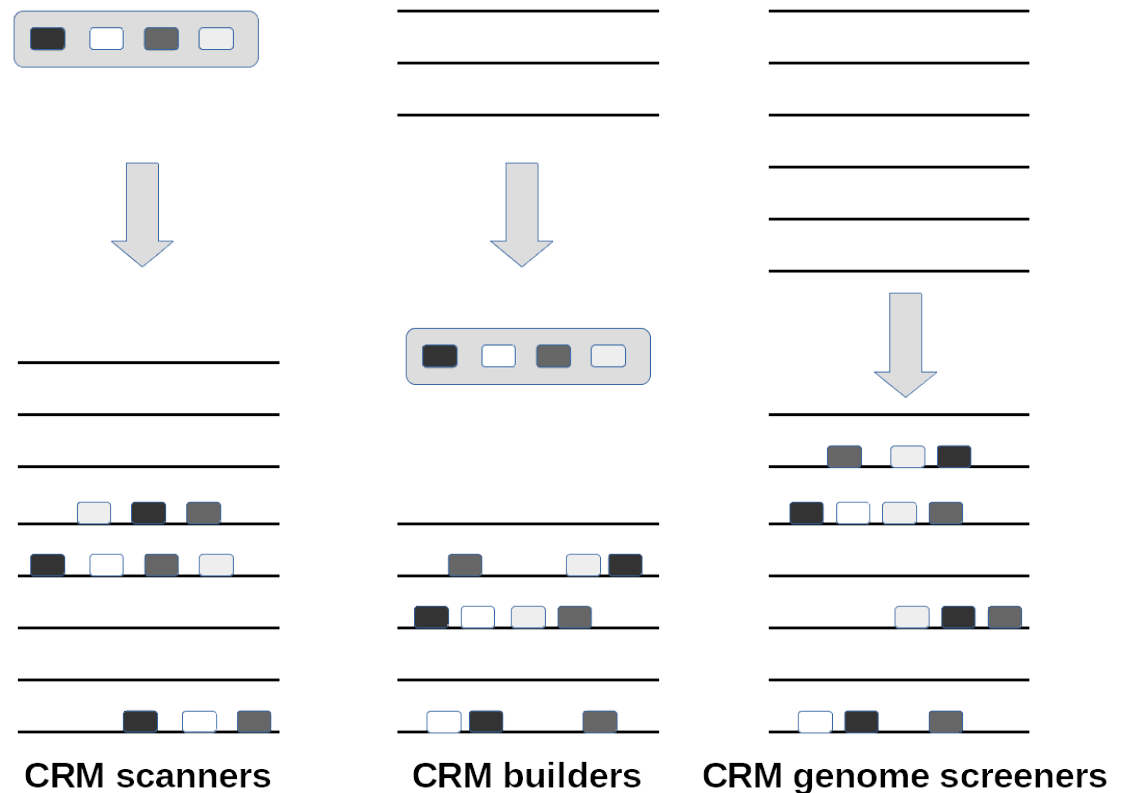
or they may simply favor the binding of a homo-oligomeric TF (e.g. p53, or NF- $\kappa$ B).

Some transcription factors such as the TF pair Oct4 and Sox2 have well known interaction partners.

## 7.6 identify Cis-regulatory motifs

(left) CRM scanners require user-defined motif combinations as input to search for putative regulatory regions.

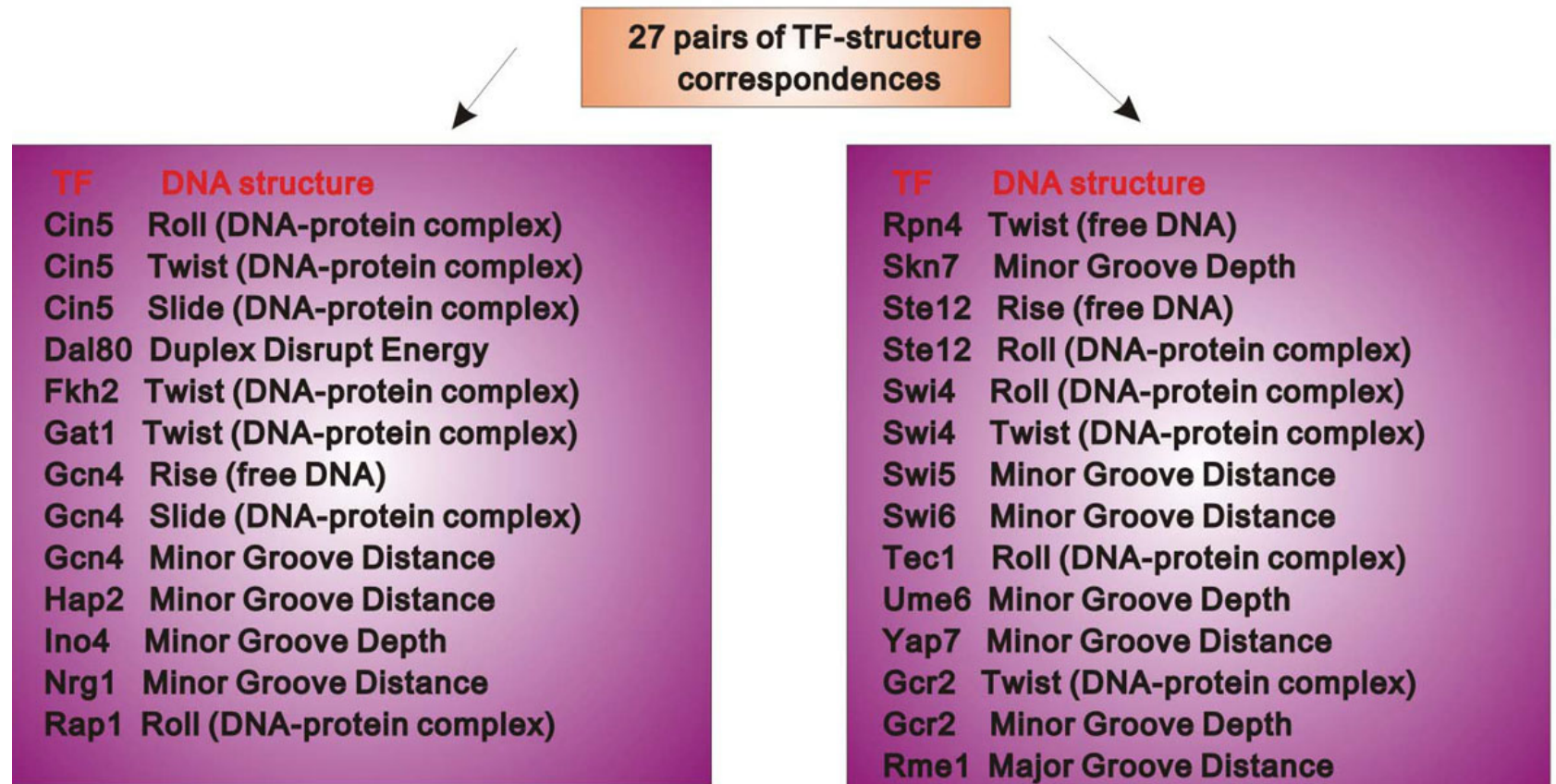
(middle) CRM builders analyze a set of co-regulated genes as input and produce candidate motif combinations, as well as similar target regions.



(right) CRM genome screeners search for homotypic or heterotypic motif clusters without making assumptions about the involved TFs.

# What do TFs recognize?

- (1) Amino acids of TFs make specific contacts (e.g. hydrogen bonds) with DNA base pairs
- (2) DNA conformation depends on its sequence
- Some TFs „measure“ different aspects of the DNA conformation



# Co-expression of TFs and target genes?

Overexpression of a TF often leads to induction or repression of target genes.

This suggests that many target genes can be regulated simply by the abundance (expression levels) of the TF.

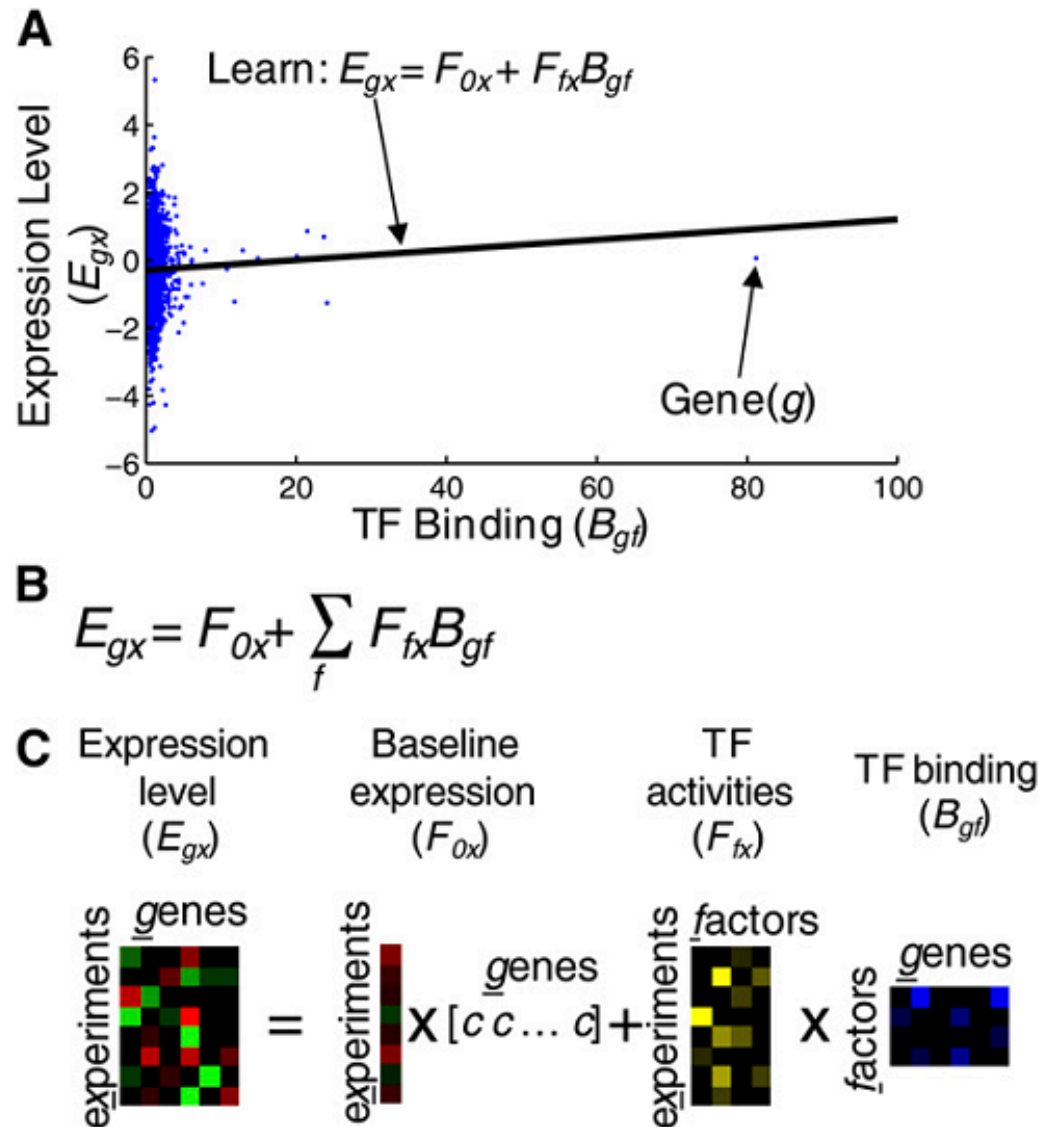
However, across 1000 microarray expression experiments for yeast, the **correlation** between a TF's expression and that of its ChIP-based targets was typically **very low** (only between 0 and 0.25)!

At least some of this (small) correlation can be accounted for by the fact that a subset of TFs autoregulate themselves.

→ In yeast, TF expression accounts for only a minority of the regulation of TF activity.



# Using regression to predict gene expression



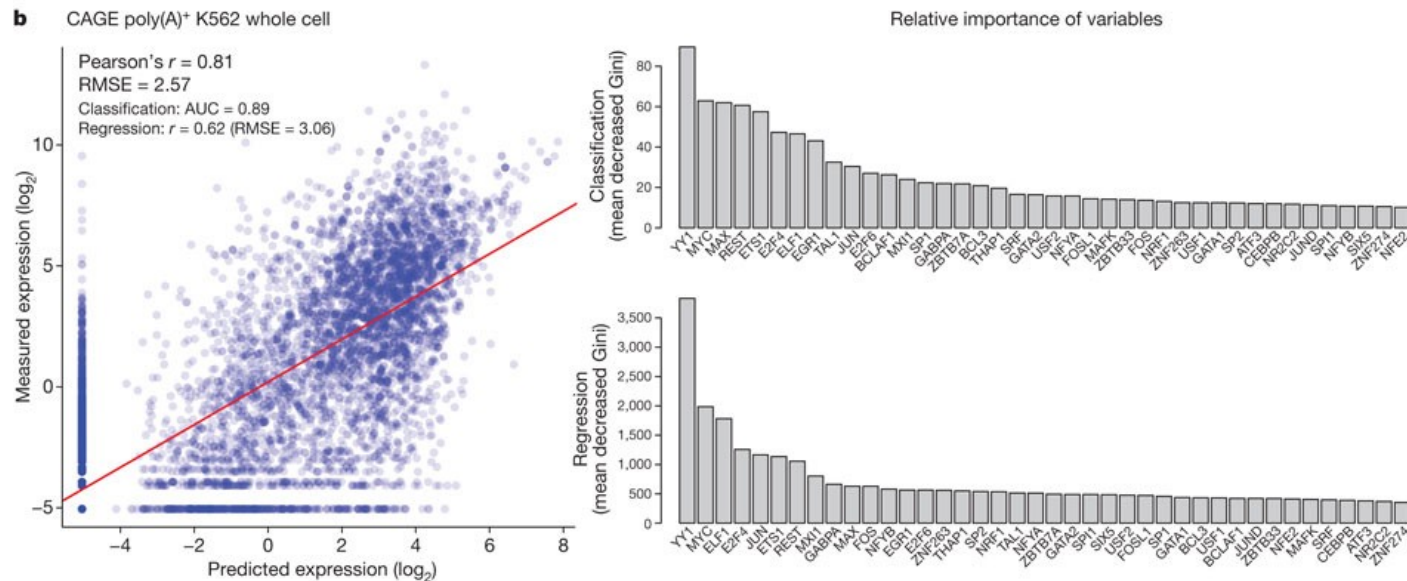
(A) Example where the relationship between expression level ( $E_{gx}$ ) and TF binding to promoters ( $B_{gf}$ ) is found for a single experiment ( $x$ ) and a single TF ( $f$ ). Here, the model learns 2 parameters: the background expression level for all genes in the experiment ( $F_{0x}$ ) and the activity of the transcription factor in the given experiment ( $F_{fx}$ ).

(B) The generalized equation for multiple factors and multiple experiments.

(C) Matrix representation of the generalized equation.

Baseline expression is the same for all genes and so is represented as a single vector multiplied by a row vector of constants where  $c = 1/(\text{no. genes})$ .

# ENCODE



AUC: area under curve;  
 Gini: Gini coefficient;  
 RMSE: root mean square error.

The ENCODE project studied how well the occupancy of TFBS is correlated with RNA production in human K562 cells.

(left) Scatter plot comparing a linear regression curve (red line) with observed values for RNA production (blue circles).

(right) Bar graphs showing the most important TFs both in the initial classification phase (top) or the quantitative regression phase (bottom). Larger values indicate increasing importance of the variable in the model.

ENCODE Project Consortium, Nature 489, 57 (2012)

# Transcription factors in human: ENCODE

Some TFs can either activate or repress target genes.

The TF YY1 shows the largest mixed group of target genes.

TF	Ubiquitously activated	Ubiquitously repressed
YY1	COQ5 <sup>cd</sup> CPNE1 CPSF2 <sup>cd</sup> CR613718 IP6K2 <sup>d</sup> NARS <sup>ac</sup> PAK4 <sup>d</sup> PSMB4 <sup>ac</sup> UBR5	AC091153.1 ATP5O BIRC6 <sup>d</sup> CAPZA2 CXorf26 DKFZp434H247 EFHA1 MRPS10 <sup>c</sup> MRPS18B <sup>acd</sup> NUP160 OXCT1 PSMD8 <sup>ac</sup> SNX27 SNX3 <sup>ad</sup> SRP68 <sup>ad</sup> TNKS



IUBD.pdb  
human YY1

Whitfield et al. Genome Biology 2012, 13:R50

# Summary transcription

- Gene transcription (mRNA levels) is controlled by transcription factors (activating / repressing) and by microRNAs (degrading) (see later lecture)
- Binding regions of TFs are ca. 5 – 10 bp stretches of DNA
- Global TFs regulate hundreds of target genes
- Global TFs often act together with more specific TFs
- TF expression only weakly correlated with expression of target genes (yeast)
- Some TFs can activate or repress target genes. Use similar binding motifs for this.