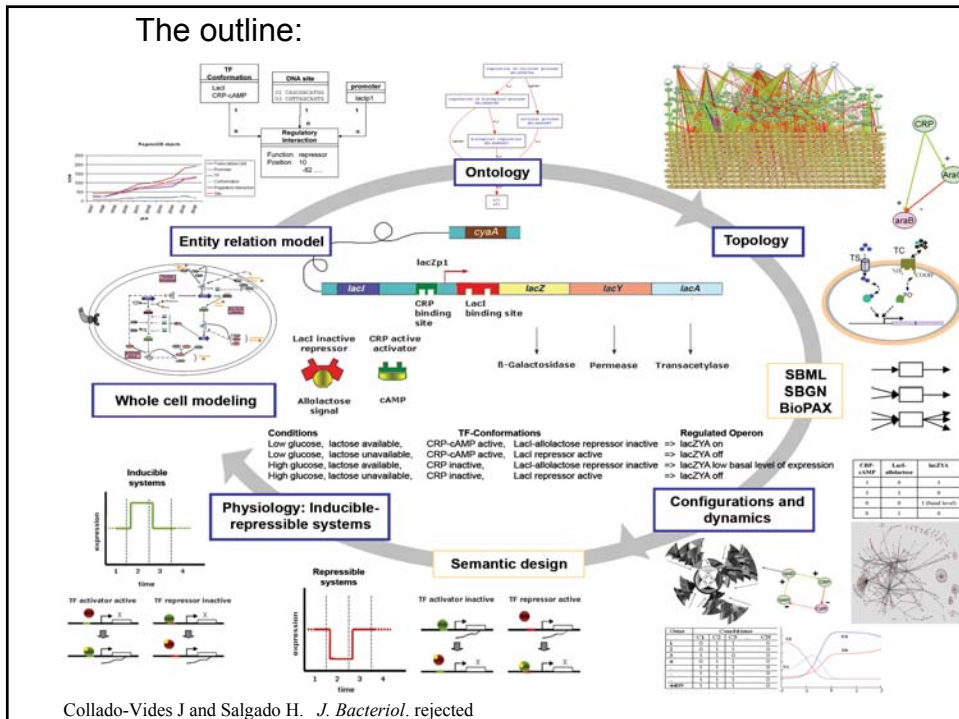
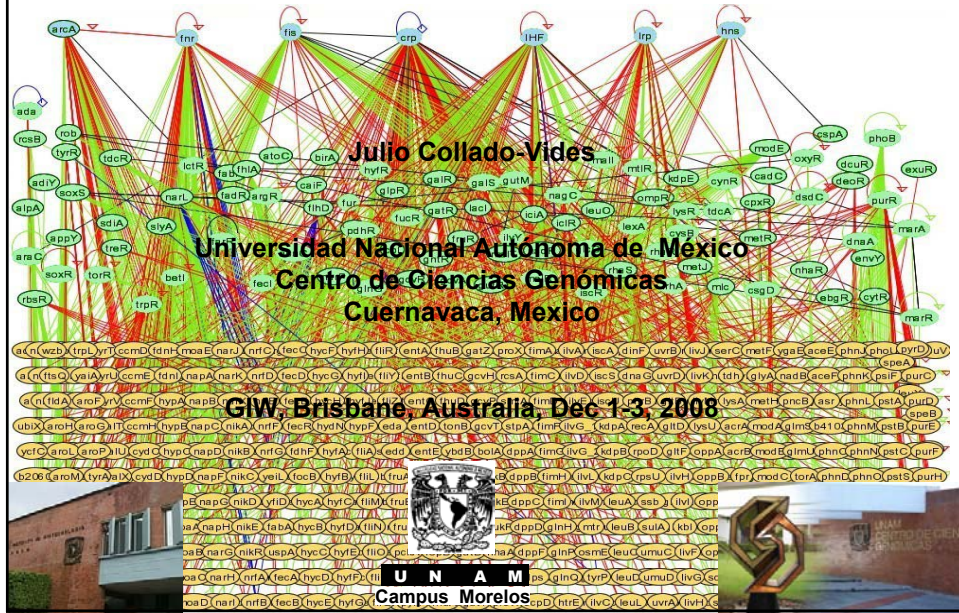


Computational Biology and Modeling of Gene Regulation in Bacteria



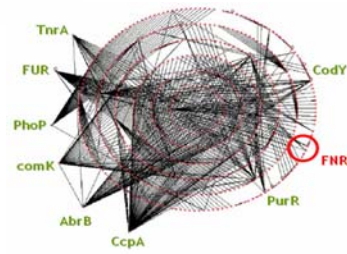
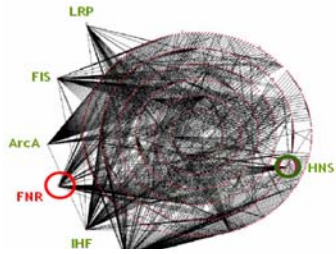
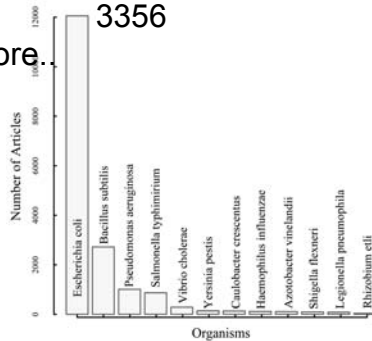
Collado-Vides J and Salgado H. *J. Bacteriol.* rejected

Escherichia coli: Model with knowledge

As of July 2008

Escherichia coli

Regulons 361
 TF BSs 2342
 Promoters 1754
 TFs 163
 TUs 3356
 and more...



Bacillus subtilis

Textpresso

Home : Sites : Downloads : Publications : About : Contact

Textpresso

Sites

- C. elegans
- D. melanogaster
- Neurospora
- Arabidopsis
- Drosophila
- Rat
- Zebrafish
- Neurospora
- Alzheimer's
-
- S. cerevisiae
- T. thermophila
- D. salina
- Bacillus
- Regulon DB
- OHSU
- Fungal
- Vaxpresso
- Phanexpress

Textpresso is a text-mining system for scientific literature. Textpresso's two major elements are (1) access to full text, so that entire articles can be searched, and (2) introduction of categories of biological concepts and classes that relate two objects (e.g., association, regulation, etc.) or describe one (e.g., methods, etc.). A search engine enables the user to search for one or a combination of these categories and/or keywords within an entire literature.

Textpresso is useful as a search engine for researchers as well as a curation tool. It was developed as a part of *Vermilab* and is used extensively by *C. elegans* curators. Textpresso has currently been implemented for 17 different literatures, and can readily be extended to other corpora of text.

Textpresso access to literature on gene regulation

News and updates

July 11th, 2008: A new server, and the software files missing or not working.

Software available

Site last updated: California 10/2008

2472 full papers
 3125 abstracts
 4200 curation notes

RegulonDB
 Escherichia coli K12 Transcriptional Network

Search:

Gene: **ompR**

Map position (nucleotides): 3533887 - 3534956 [Genome browser](#)

Sequence: [Get nucleotide sequence](#)

GC content: 55.14

View Matches Page

Show Matches

Note: The color scheme only highlights words that are in the selected categories. It does not mean that other specifications for that word are met. In the color scheme, keyword colors supersede category colors, but are superseded by the hyperlink color.

Query: Categories + regulation + effect - Keywords + ompR*

Display page 1 or previous or next page.

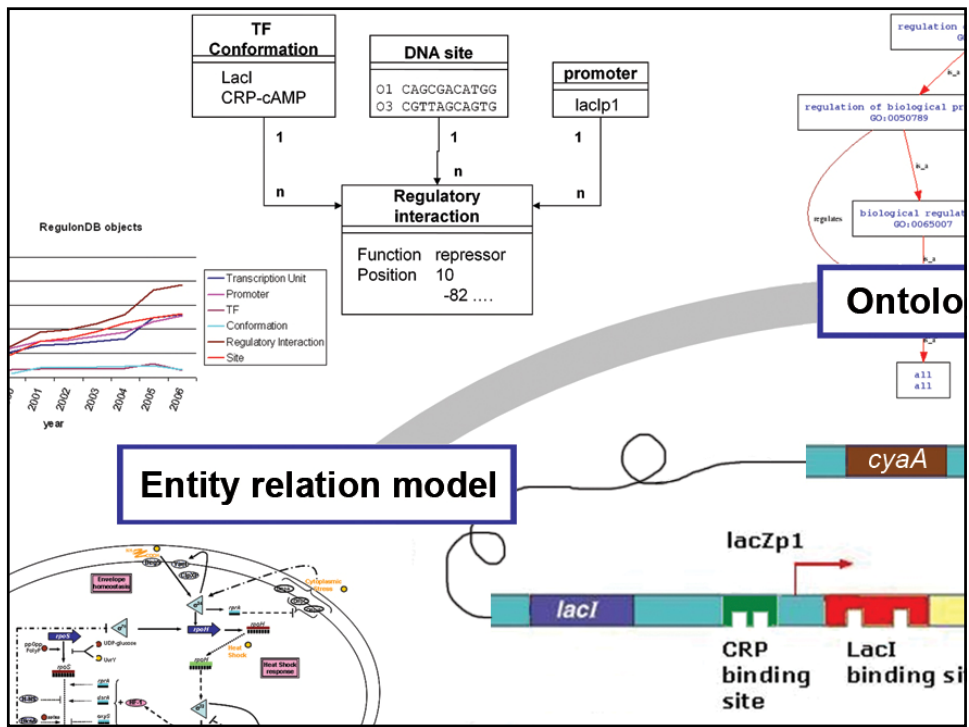
File ID pmid11158569 (Abstract), Sentences 1 to 11: *Escherichia coli* modulates its porin expression through a histidine kinase, EnvZ, and its cognate response regulator, OmpR. EnvZ is a bifunctional enzyme that possesses both OmpR kinase and phosphorylated OmpR (OmpR-P) phosphatase activities and thus controls the cellular level of OmpR-P. In an in vitro -assay system, the addition of OmpR to the reaction mixture consisting of the cytoplasmic domain of EnvZ (EnvZc) and ATP produces a barely detectable amount of OmpR-P because of the dual activities of EnvZ. Here we report that DNA fragments containing the upstream promoter regions of the porin genes (ompF and ompC) can shift the equilibrium between OmpR and OmpR-P dramatically toward OmpR-P. Among the four reactions occurring in the mixture, only the EnvZ phosphatase activity was inhibited severely by the specific DNA, in contrast to the previous report by Kenney and her associates that DNA stimulates OmpR phosphorylation by EnvZ [Ames, S.K., Frankema, N AND Kenney, L.J. (1999) Proc. Natl. Acad. Sci. USA 96, 11792-11797]. The autophosphorylation of EnvZc and the phosphotransfer from phosphorylated EnvZc to OmpR were not affected by DNA, whereas the autodephosphorylation of OmpR-P was inhibited slightly. We propose that the apparent inhibitory effect of DNA on the EnvZ phosphatase function is caused by sequestering OmpR-P from the reaction as a result of OmpR-P binding to DNA.

File ID pmid11158569 (Abstract), Sentences 1 to 12: *Escherichia coli* modulates its porin expression through a histidine kinase, EnvZ, and its cognate response regulator, OmpR. EnvZ is a bifunctional enzyme that possesses both OmpR kinase and phosphorylated OmpR (OmpR-P) phosphatase activities and thus controls the cellular level of OmpR-P. In an in vitro -assay system, the addition of OmpR to the reaction mixture consisting of the cytoplasmic domain of EnvZ (EnvZc) and ATP produces a barely detectable amount of OmpR-P because of the dual activities of EnvZ. Here we report that DNA fragments containing the upstream promoter regions of the porin genes (ompF and ompC) can shift the equilibrium between OmpR and OmpR-P dramatically toward OmpR-P. Among the four reactions occurring in the mixture, only the EnvZ phosphatase activity was inhibited severely by the specific DNA, in contrast to the previous report by Kenney and her associates that DNA stimulates OmpR phosphorylation by EnvZ [Ames, S.K., Frankema, N AND Kenney, L.J. (1999) Proc. Natl. Acad. Sci. USA 96, 11792-11797]. The autophosphorylation of EnvZc and the phosphotransfer from phosphorylated EnvZc to OmpR were not affected by DNA, whereas the autodephosphorylation of OmpR-P was inhibited slightly. We propose that the apparent inhibitory effect of DNA on the EnvZ phosphatase function is caused by sequestering OmpR-P from the reaction as a result of OmpR-P binding to DNA.

File ID pmid11158569 (Abstract), Sentences 1 to 15: *Escherichia coli* modulates its porin expression through a histidine kinase, EnvZ, and its cognate response regulator, OmpR. EnvZ is a bifunctional enzyme that possesses both OmpR kinase and phosphorylated OmpR (OmpR-P) phosphatase activities and thus controls the cellular level of OmpR-P. In an in vitro -assay system, the addition of OmpR to the reaction mixture consisting of the cytoplasmic domain of EnvZ (EnvZc) and ATP produces a barely detectable amount of OmpR-P because of the dual activities of EnvZ. Here we report that DNA fragments containing the upstream promoter regions of the porin genes (ompF and ompC) can shift the equilibrium between OmpR and OmpR-P dramatically toward OmpR-P. Among the four reactions occurring in the mixture, only the EnvZ phosphatase activity was inhibited severely by the specific DNA, in contrast to the previous report by Kenney and her associates that DNA stimulates OmpR phosphorylation by EnvZ [Ames, S.K., Frankema, N AND Kenney, L.J. (1999) Proc. Natl. Acad. Sci. USA 96, 11792-11797]. The autophosphorylation of EnvZc and the phosphotransfer from phosphorylated EnvZc to OmpR were not affected by DNA, whereas the autodephosphorylation of OmpR-P was inhibited slightly. We propose that the apparent inhibitory effect of DNA on the EnvZ phosphatase function is caused by sequestering OmpR-P from the reaction as a result of OmpR-P binding to DNA.

File ID pmid11158569 (Paper), Sentences 1 to 11: The critical role of DNA in the equilibrium between OmpR and phosphorylated OmpR mediated by EnvZ in *Escherichia coli* Ling Qiu, Takeshi Yoshida, and Masayori Inouye Department of Biochemistry, Robert Wood Johnson Medical School, 675 Hoes Lane, Piscataway, NJ 08854 Edited by Carol A Gross, University of California, San Francisco, CA and accepted December 11, 2000 Received for review August 10

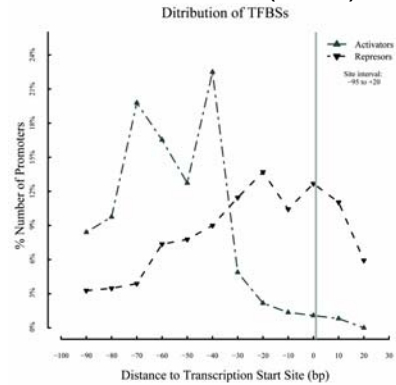
Rodríguez-Penagos, C., Salgado, H., Martínez-Flores, I., and Collado-Vides, J. (2007) "Automatic reconstruction of a bacterial regulatory network using Natural Language Processing" *BMC Bioinformatics* 8: 293.



**Proximal sites in $\sigma 70$ promoters:
enable direct contact with RNAP**

-65 to +20 (1991)
-95 to +20 (2008)

QuickTime™ and a decompressor are needed to see this picture.



1991: 119 $\sigma 70$ promoters
2008: 421 $\sigma 70$ promoters with 1382 TF-BSs
Only 26 promoters (6%) lack a proximal site

Collado-Vides J., Magasanik B. and Gralla J.D. (1991) "Control site location and transcriptional regulation in *Escherichia coli*" *Microbiol. Reviews.* 55:371-394
Collado-Vides et al., (2009) "Bioinformatics resources for the study of gene regulation in bacteria" *J. of Bacteriology* (inaugural issue in Computational Biology) -in press

Origins of RegulonDB: 1991 and 1998 -ten years ago

Collado-Vides J., Magasanik B. and Gralla J.D. (1991) "Control site location and transcriptional regulation in *Escherichia coli*" *Microbiol. Reviews.* 55:371-394

Huerta A.M.; Salgado H.; Thieffry D., and Collado-Vides J. (1998) "RegulonDB: A Database on Transcription Regulation in *Escherichia coli*" *Nucleic Acids Res.* 26: 55-60

Different sigma-types of promoter predictions

Sigma Factor	No. of Promoters	Promoters w- Strong Evidence*	Sigma	# Predictios	# Regions	Cutoff ($\mu - X \sigma$)	Accuracy (sensitivity, precision)
Sigma24	61	60	Sigma24*	39	39	-1.9 (0.14 - 2.* 0.98)	0.85 (0.75,0.95)
Sigma28	20	5	Sigma28	123	121	7.32 (10.8 - 1.* 3.6)	0.9 (0.9,0.9)
Sigma32	31	22	Sigma32	455	412	6.23 (9.56 - 1.* 3.2)	0.6 (0.5,0.6)
Sigma38	81	38	Sigma38	1767	1243	4.48 (5.6 - 1.* 2.2)	0.6 (0.6,0.5)
Sigma19	1	1	Sigma19	4	4	NA	-
Sigma54	32	16	Sigma54	152	147	7.10 (9.7 - 0.5.* 3.2)	0.8 (0.7,1.0)
Total	226	142	Total	3,316	2,609	-	-

Araceli Huerta

*Rhodius et al. 2006 PLoS

Position Weight matrices

Counts position	1	2	3	4	5	6	7	8	9	10	Sum
A	2	3	19	0	1	1	1	17	5	3	52
C	0	2	0	16	1	1	17	1	11	3	52
G	3	12	0	1	17	0	0	0	1	9	43
T	14	2	0	2	0	17	1	1	2	4	43

$$w_{i,j} = \ln \left(\frac{f_{i,j}}{p_i} \right)$$

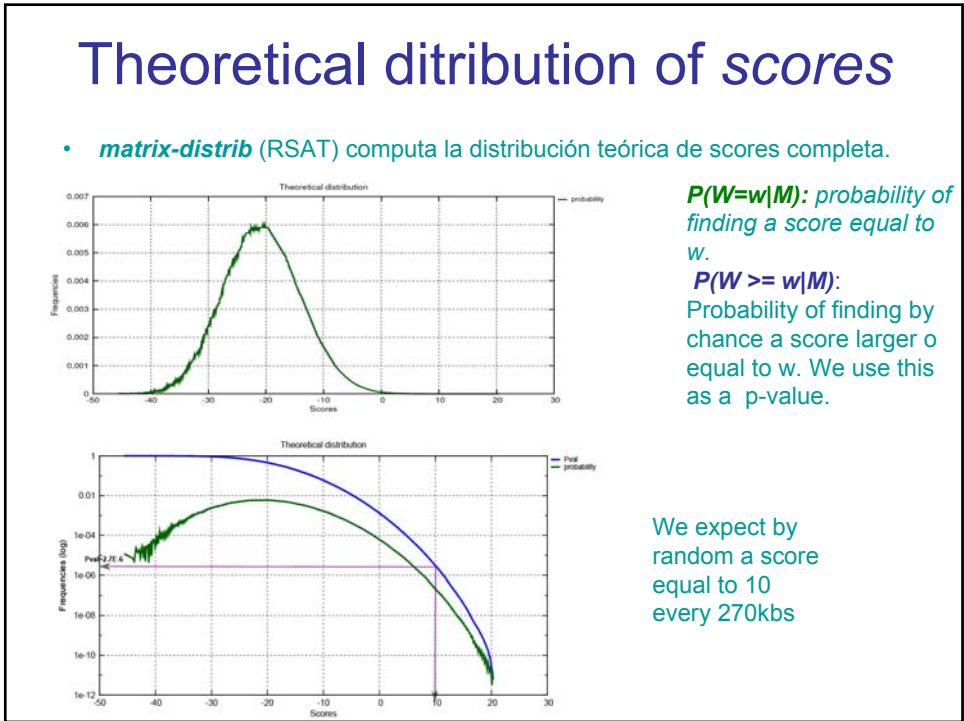
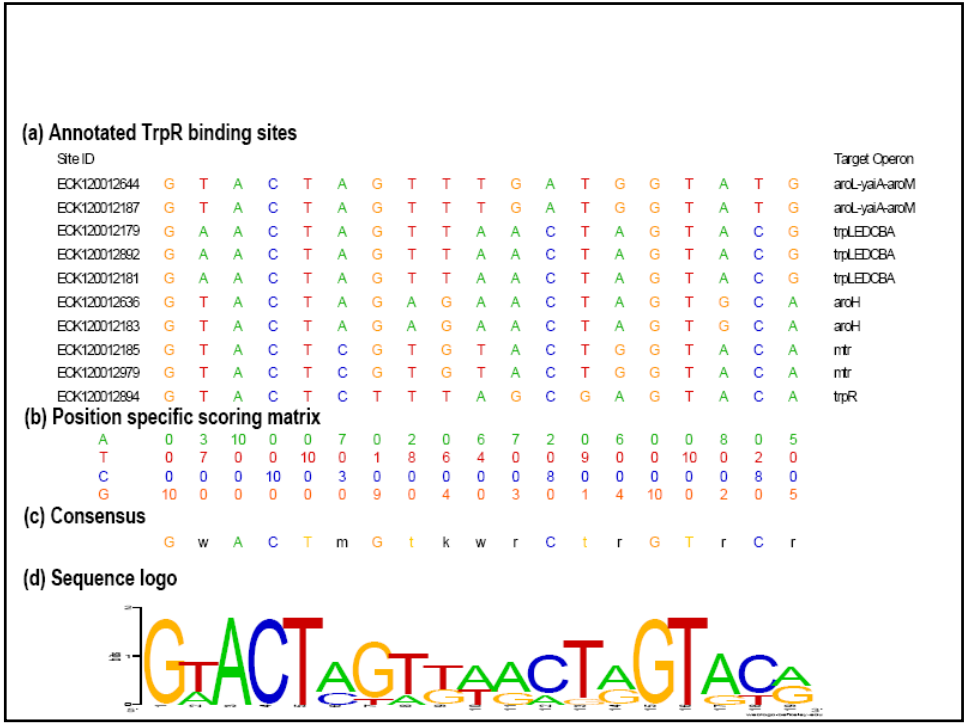
Weight matrix position	1	2	3	4	5	6	7	8	9	10	Sum
A	-0.80	-0.43	1.35	-2.99	-1.38	-1.38	-1.38	1.24	0.05	-0.43	-6.15
C	-3.00	-0.80	-3.00	1.18	-1.39	-1.39	1.24	-1.39	0.81	-0.43	-8.16
G	-0.43	0.89	-3.00	-1.39	1.24	-3.00	-3.00	-3.00	-1.39	0.61	-12.47
T	1.05	-0.80	-2.99	-0.80	-2.99	1.24	-1.39	-1.39	-0.80	-0.16	-9.03
sum	-3.18	-1.13	-7.64	-4.00	-4.53	-4.53	-4.53	-4.53	-1.33	-0.41	-35.80



Weight of a sequence segment, computed from the weight matrix

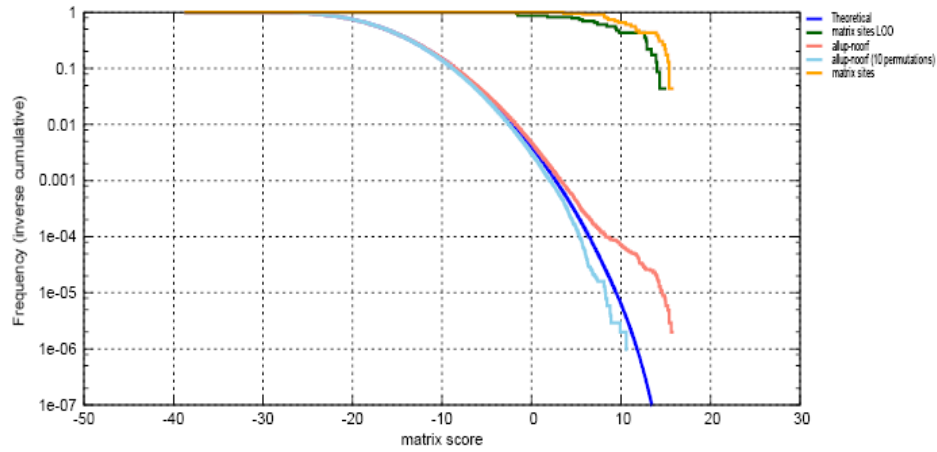
residue	r	T	G	T	A	A	T	A	A	T	A
W(r)	1.05	0.89	-2.99	-2.99	-1.38	1.24	-1.38	1.24	-0.80	-0.43	
Weight	-5.558	=SUM[W(r)]									

Olivier Sand, Jean Valéry Turatsinze and Jacques van Helden



Empirical score distribution

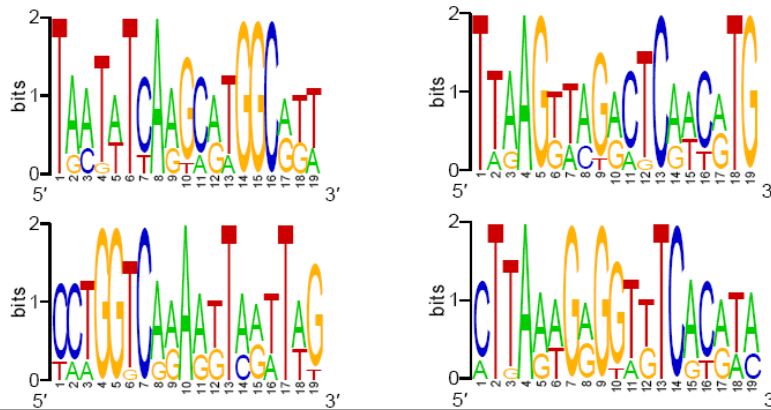
- 10 permutations

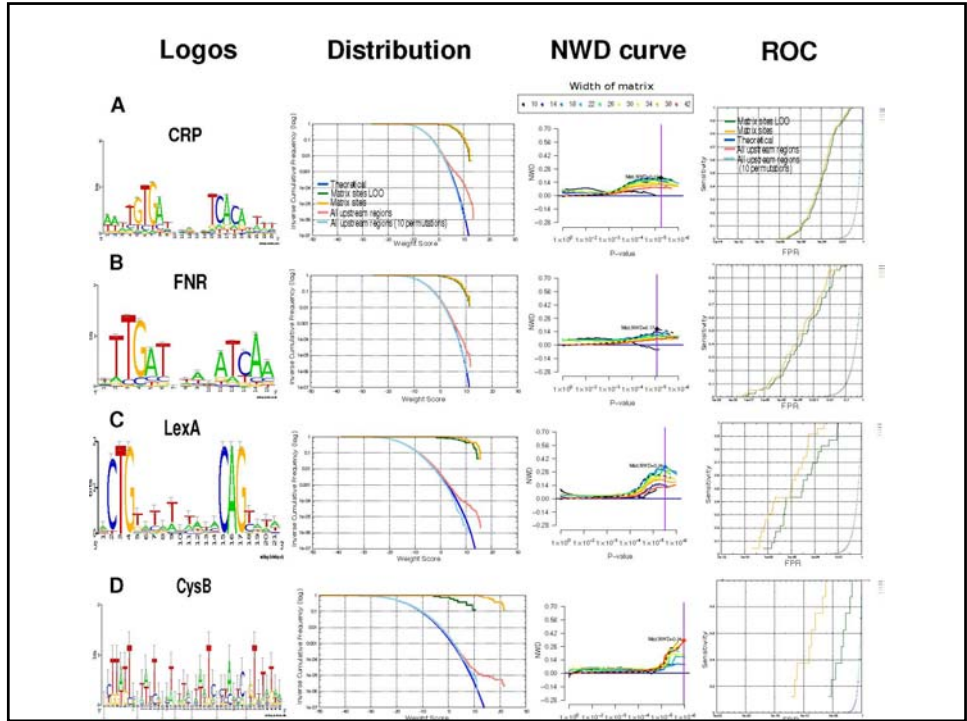


Shuffled Matrices



(e) Logos of column-permuted matrices



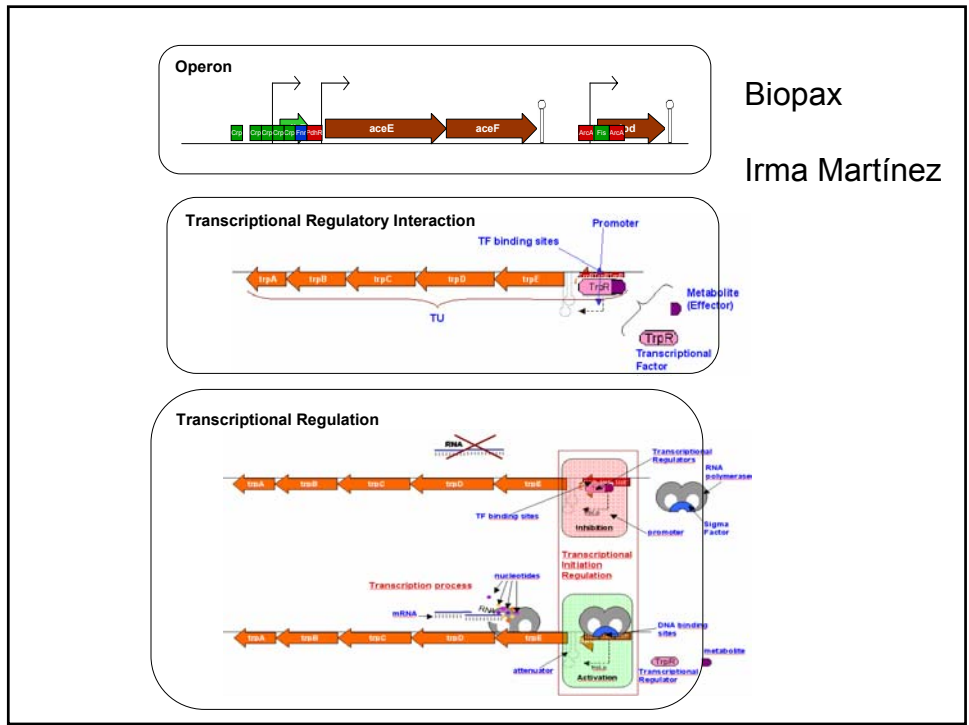
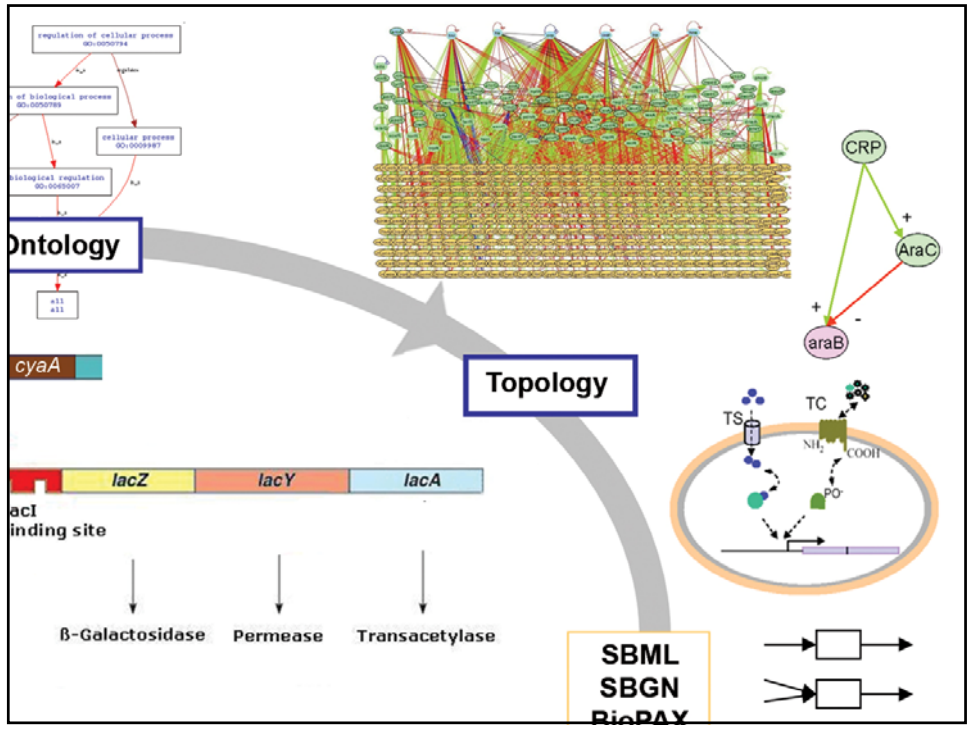


A comprehensive repertoire of PWMs for 54 TFs

- P-values 1.60E-05 or smaller

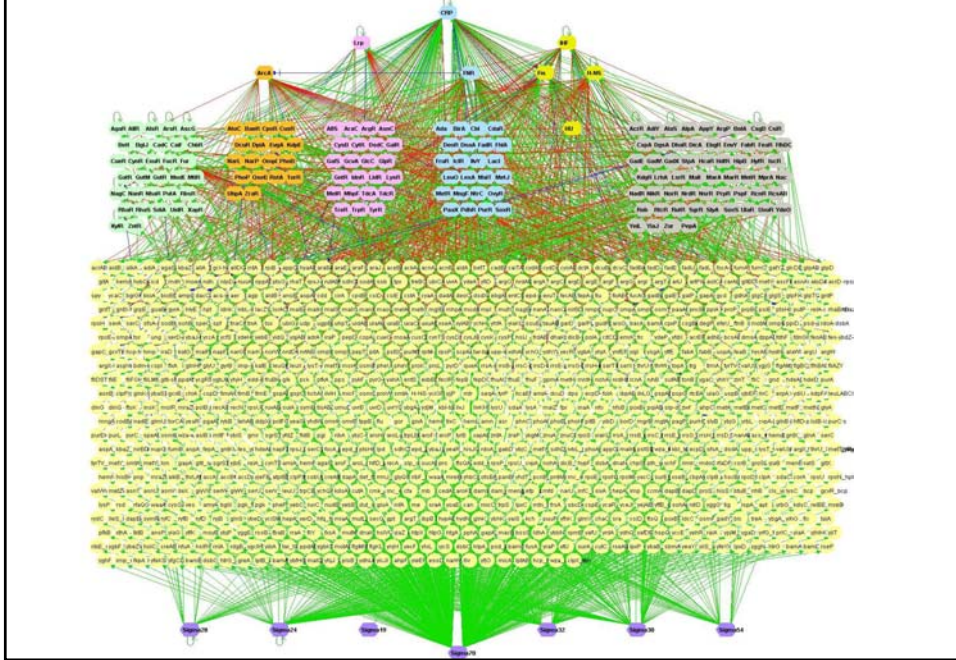
Factor	AnnotatedSites	PredictedSites	TP	PPV	Sensitivity	Recov. Percent
DgsA	7	10	7	0.7	1.0	100.0
FliA	4	10	4	0.4	1.0	100.0
GcvA	5	7	5	0.7	1.0	100.0
HipB	4	41	4	0.1	1.0	100.0
MeiR	10	42	10	0.2	1.0	100.0
MetR	6	17	6	0.4	1.0	100.0
UjaR	4	12	4	0.3	1.0	100.0
XylR	4	23	4	0.2	1.0	100.0
ArgR	30	70	28	0.4	0.9	93.3
GntR	13	37	12	0.3	0.9	92.3
Rob	6	23	5	0.2	0.8	83.3
NtrC	17	92	14	0.2	0.8	82.4
OxyR	11	42	9	0.2	0.8	81.8
GadE	5	5	4	0.8	0.8	80.0
NarR	5	10	4	0.4	0.8	80.0
MetJ	25	73	18	0.2	0.7	72.0
ModE	7	12	5	0.4	0.7	71.4
MaiT	20	20	14	0.7	0.7	70.0
TyrR	19	50	13	0.3	0.7	68.4
CysB	9	13	6	0.5	0.7	66.7
Fur	76	48	50	1.0	0.7	65.8
LexA	29	77	19	0.2	0.7	65.5
NagC	14	20	9	0.5	0.6	64.3
GlpR	23	63	14	0.2	0.6	60.9
CytR	10	15	6	0.4	0.6	60.0
TorR	8	9	4	0.4	0.5	50.0
TtpR	10	11	5	0.5	0.5	50.0
Agar	11	12	5	0.4	0.5	45.5
IscR	10	25	4	0.2	0.4	40.0
PhoB	20	69	8	0.1	0.4	40.0
PurR	19	92	7	0.1	0.4	36.8
FNR	79	109	26	0.2	0.3	32.9
IclR	10	18	3	0.2	0.3	30.0
OmpR	22	40	6	0.2	0.3	27.3
ArgP	4	3	1	0.3	0.3	25.0
Nac	12	12	3	0.3	0.3	25.0
NanR	9	5	2	0.4	0.2	22.2
SoxS	18	34	4	0.1	0.2	22.2
RcsAB	10	12	2	0.2	0.2	20.0
MarA	16	21	3	0.1	0.2	18.8
AraC	18	19	3	0.2	0.2	16.7
CRP	241	176	40	0.2	0.2	16.6
DnaA	8	13	1	0.1	0.1	12.5
GalR	8	19	1	0.1	0.1	12.5
ArcA	89	37	9	0.2	0.1	10.1
IHF	100	69	10	0.1	0.1	10.0
FadR	12	15	1	0.1	0.1	8.3
Fis	219	134	5	0.0	0.0	2.3
NarL	80	24	1	0.0	0.0	1.3
CsgD	4	7	0	0.0	0.0	0.0
DeoR	7	0	0	0.0	0.0	0.0
FlhDC	20	25	0	0.0	0.0	0.0
FruR	12	28	0	0.0	0.0	0.0
GalS	7	14	0	0.0	0.0	0.0
Irp	56	18	0	0.0	0.0	0.0
NarP	18	36	0	0.0	0.0	0.0
PhoP	20	65	0	0.0	0.0	0.0
RhaS	4	35	0	0.0	0.0	0.0

Medina-Rivera A., Abreu-Goodger C., Salgado H., Collado-Vides J., and van Helden J. "The good, the bad and the ugly: evaluating transcription factor binding motifs in a genome repertoire (submitted)



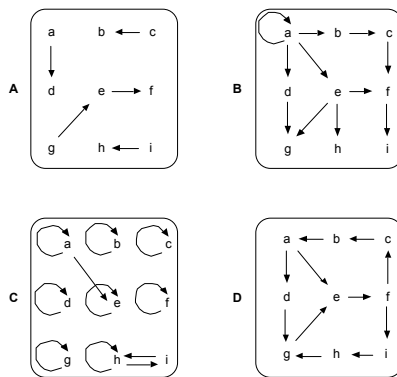
Biopax
Irma Martínez

TFs and sigma specific transcription initiation



Topology of this updated network ?

Figure 1



Freyre-González J., Alonso-Pavón J.A., Treviño-Quintana L., and Collado-Vides J. "Functional architecture of *Escherichia coli*: new insights provided by a natural decomposition approach" (2008) *Genome Biology* 9: R154

Identifying global TFs by network decomposition (Julio Freyre)

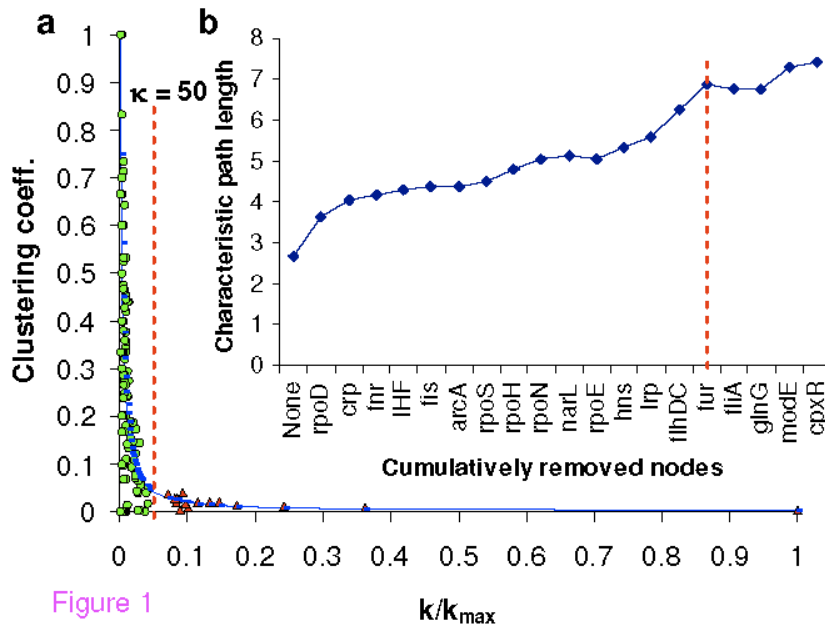
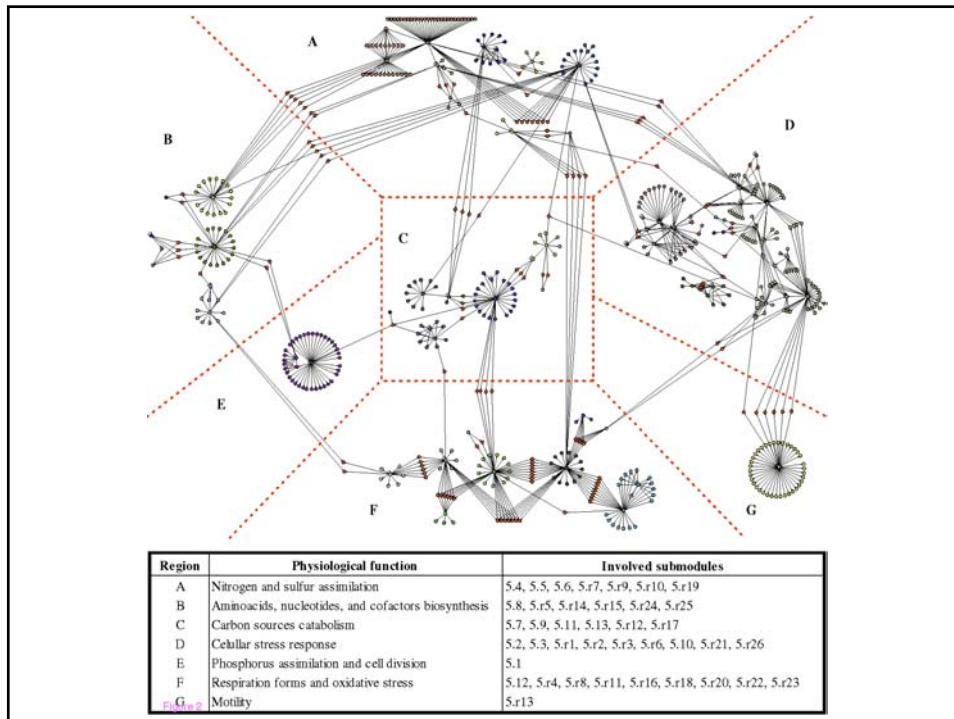
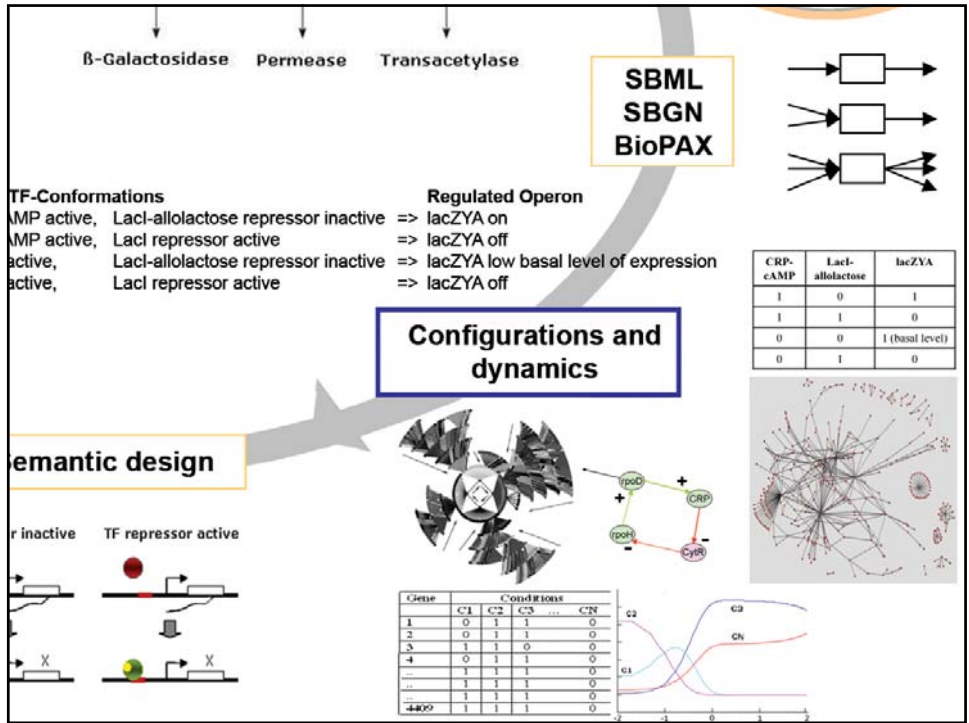


Figure 1





Feedback loops in the regulatory network: several positive ones

Para ver esta película, debe disponer de QuickTime™ y de un descompresor.

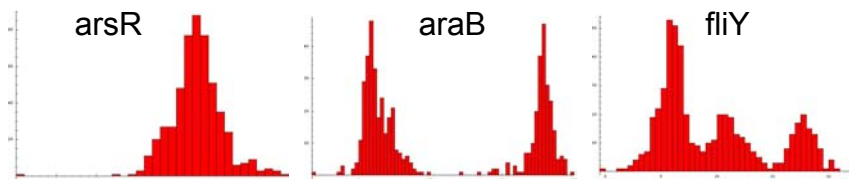
Freyre-González J., Alonso-Pavón J.A., Treviño-Quintana L., and Collado-Vides J. "Functional architecture of *Escherichia coli*: new insights provided by a natural decomposition approach" (2008) *Genome Biology* 9: R154

Phenotypic space: Discretization of expression -dialogue with RegulonDB.

Discretization-1: *An Absolute Reference*

To discretize the expression of a gene across many experiments, we used the complete dynamic range of expression as the reference.

This will allow us to explore the plasticity of a gene's expression.

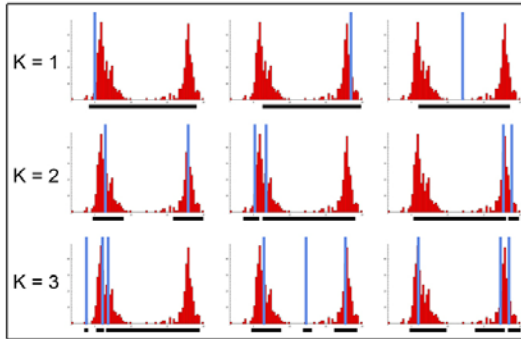


Histograms of the expression level of different genes in different conditions. The dynamic range and the typical values are easily observed.

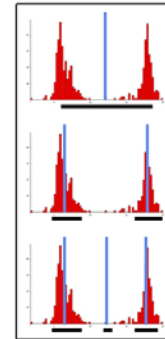
1) 380 experiments, 730 chips. Faith JJ et al., *Many Microbe Microarrays Database: uniformly normalized Affymetrix compendia with structured experimental metadata*. Nucleic Acids Research.

Discretization-2: *The Algorithm*

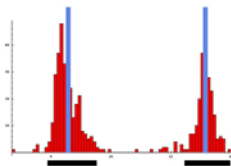
1) Run K-means algorithm with K up to 3. For each K restart several times to avoid local minima.



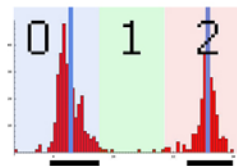
2) For each K select the set of means that concentrates the maximum of data in the minimum distance.



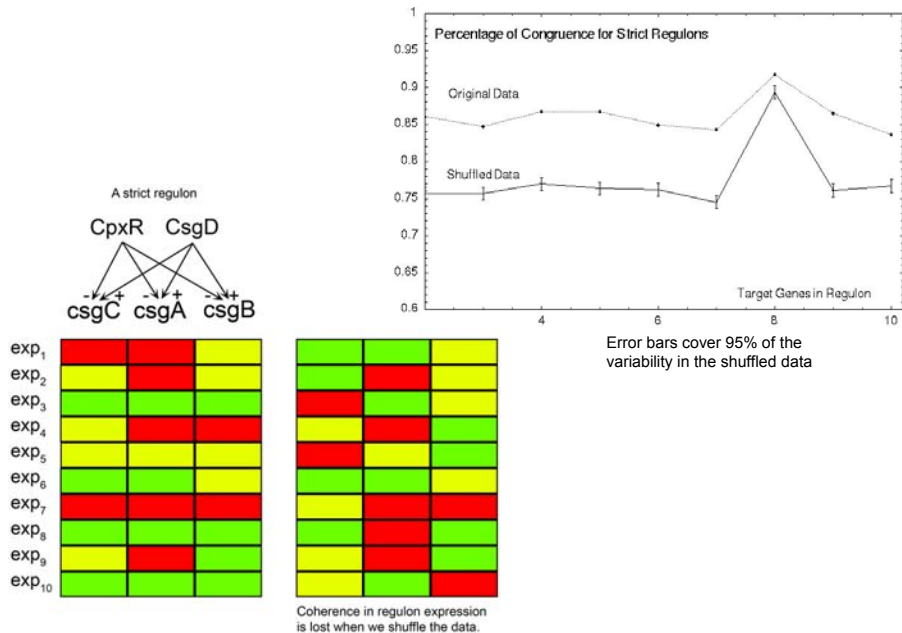
3) Select the optimum set of means.



4) Create categories. The width of the central category must be bigger than the biological-experimental variance.



Discretization-3: *Measuring Congruence to Validate*



Discretization-4: *Discretized data will enable..*

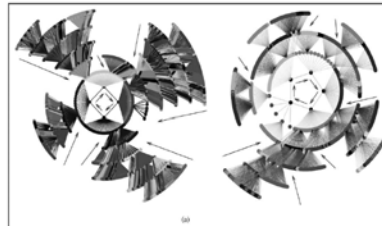
- Inferred discrete regulatory functions.

Towards a discrete dynamical model of *E. coli*'s transcriptional regulatory network

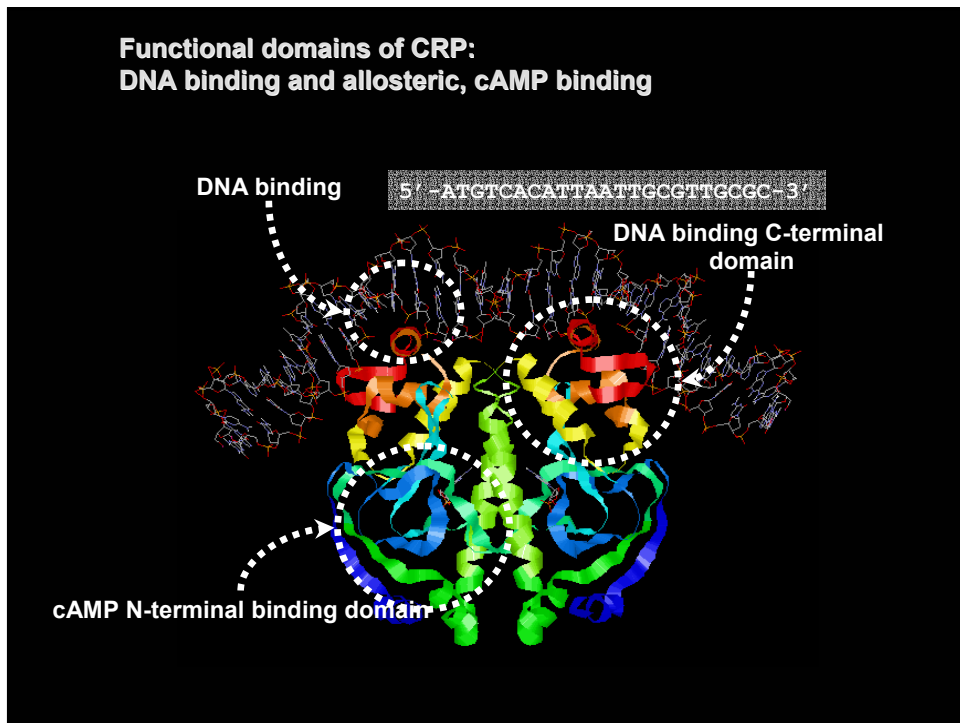
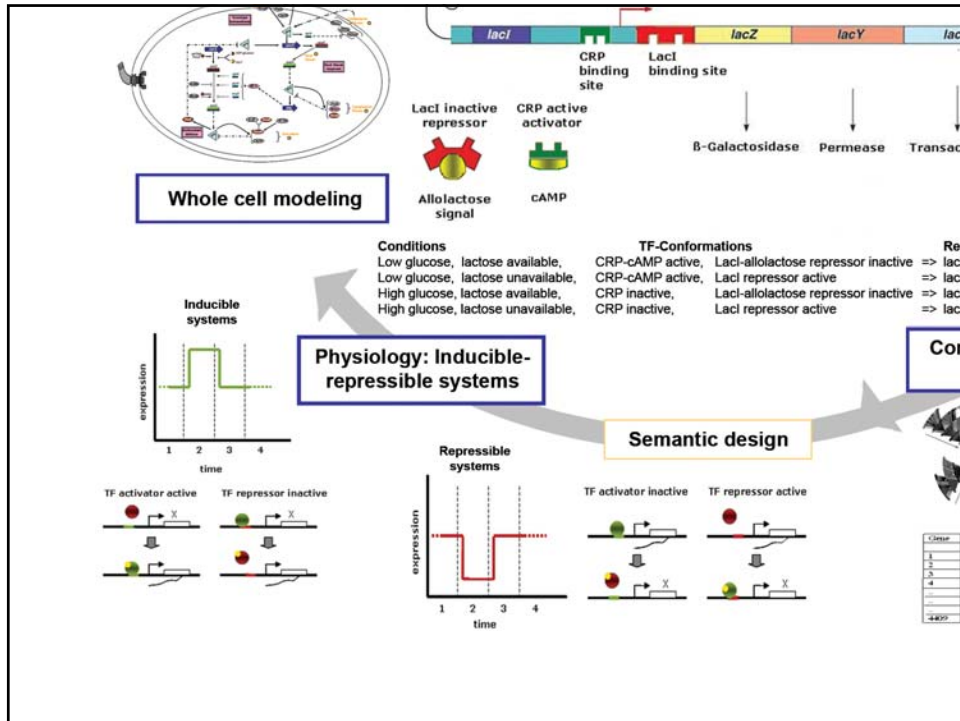
- Estimate the number of functional states (cell types).

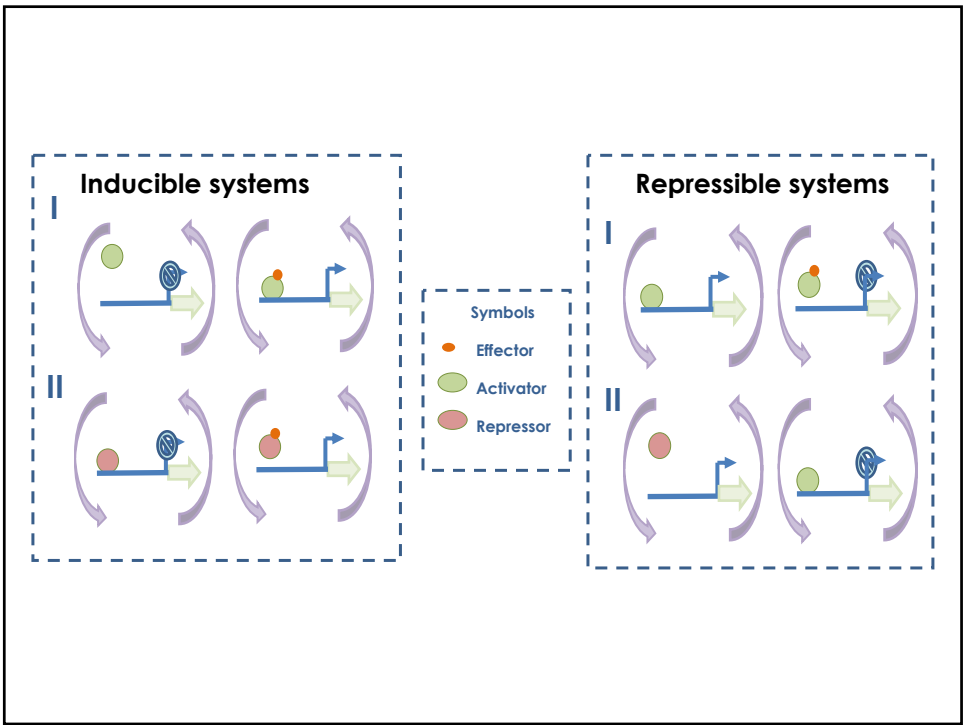
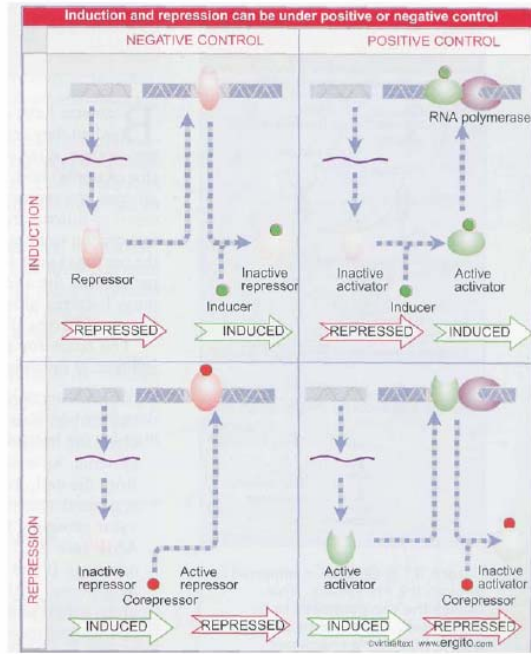
How many essentially different gene expression patterns does *E. coli*'s network encode?

In the simulated *E. coli* network, We expect 3-5 steps to reach an attractor vs a larger number in a chaotic network .

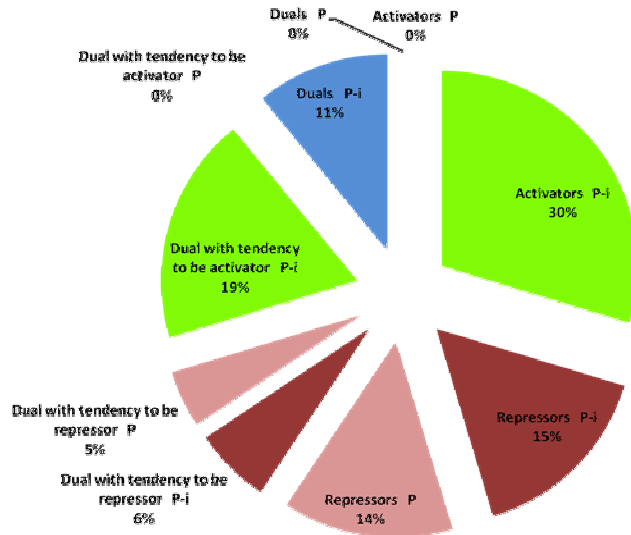


Santiago Sandoval and Maximino Aldana



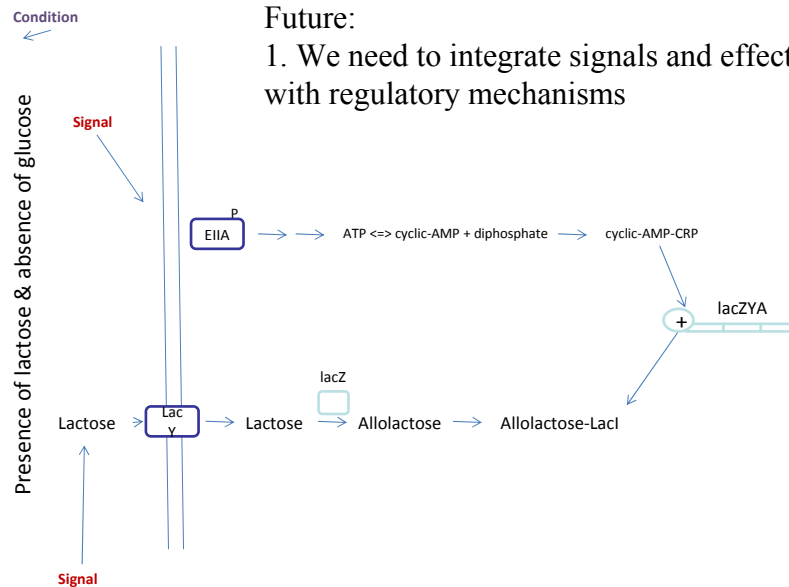


Preferences in active-DNA-conformation of transcriptional factors in *E. coli* K-12

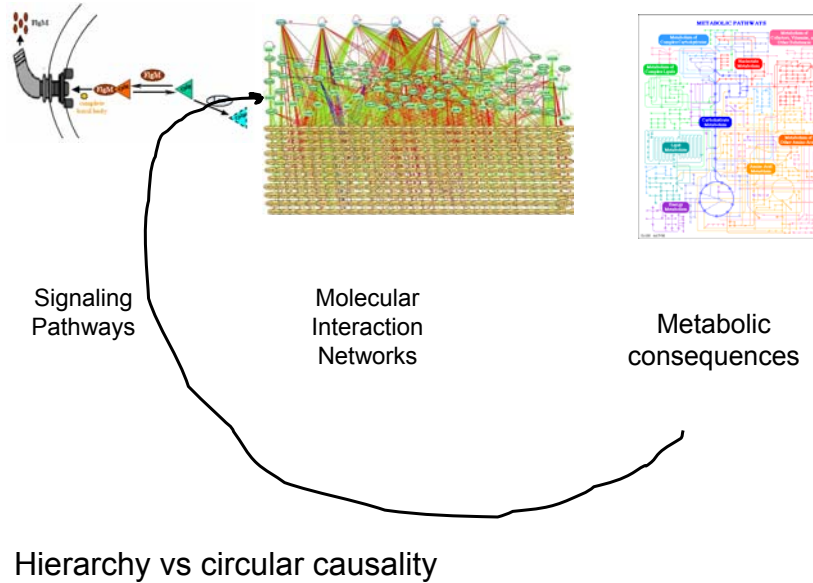


Future:

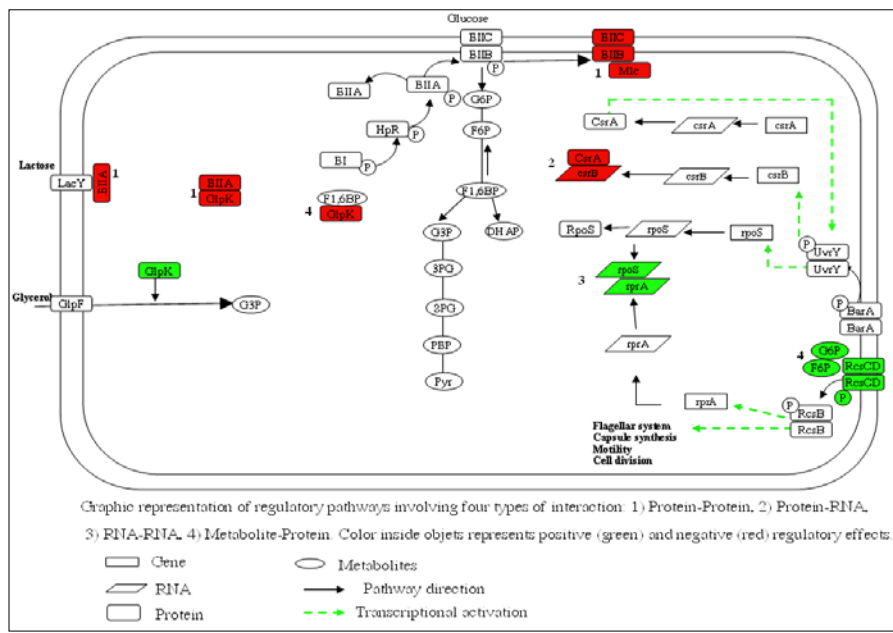
1. We need to integrate signals and effectors with regulatory mechanisms

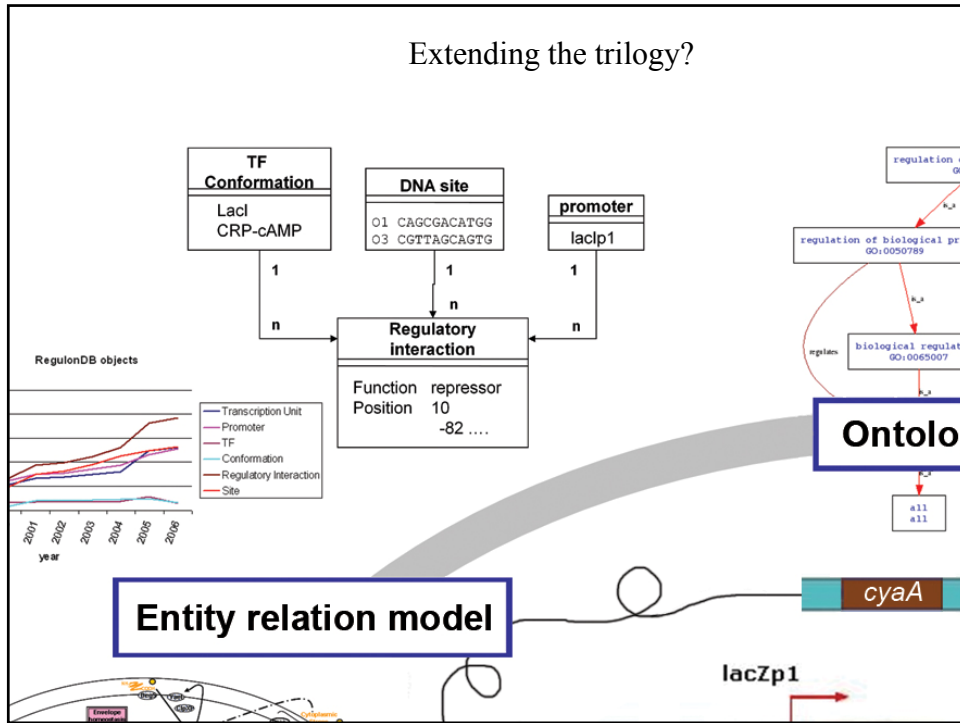


2. integrating signal sensing, regulation, affected metabolic capabilities and feedback.



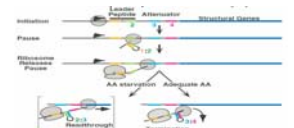
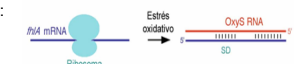
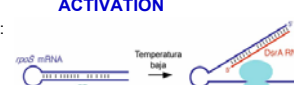
3. Regulation: a variety of mechanisms, interactions and effects

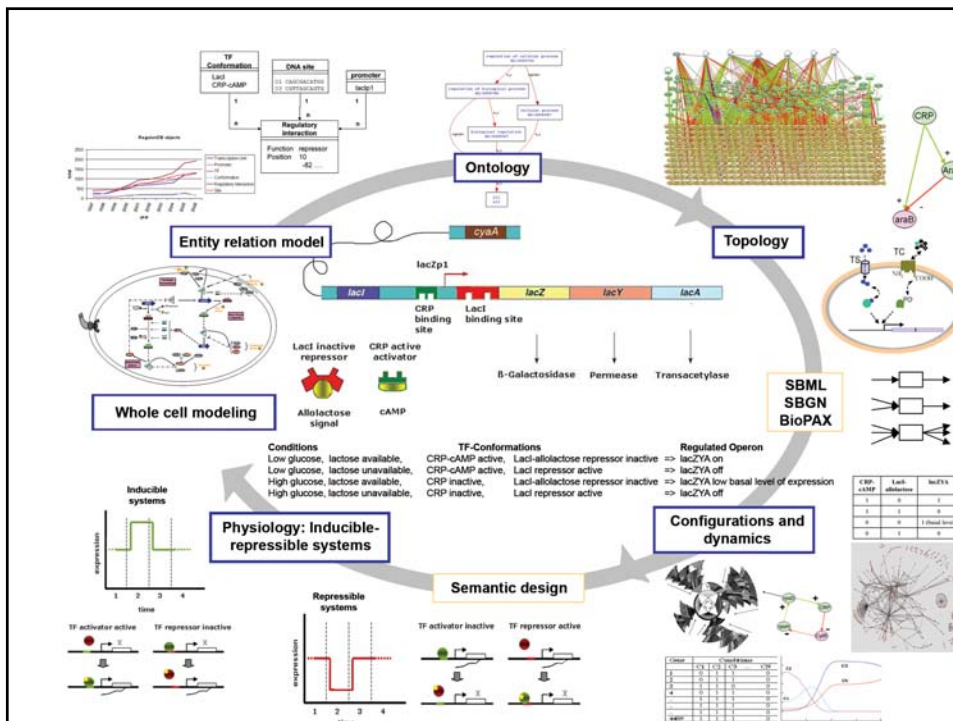




GENETIC REGULATION			
Mechanism	Executors	Target site	Effect
Transcription			
Transcription Factors: Protein-DNA REPRESSION Example: ACTIVATION Example:	Protein TF: transcription factor	DNA TFBS: transcription factor binding sites. Operator sites	Regulates promoter activity.
Structural: DNA-Protein Supercoiling, Methylation of DNA 	Protein NAS: nucleoid associated proteins (IHF, HNS, HU)	DNA DNA can be altered as a result of interaction of skilled proteins	Modifies accessibility of promoter to DNA regions.
Sigma Factors: Protein-Protein Example: Sigma Factors 	Protein Provides specificity of RNAP-promoter	Protein RNAP core have to associates with different sigma factors to be activated under different environmental conditions.	Confers functionality and specificity to RNAP for promoter recognition

GENETIC REGULATION

Mechanism	Executors	Target Interaction	Effect
Transcription			
<p>Transcriptional Attenuation: Protein-DNA</p> <p>Example: Attenuation in amino acid biosynthetic trp operon</p> 	<p>RNA</p> <p>RNA structure formed as stable transcript hairpin followed by a series of U's.</p>	<p>Protein</p> <p>The RNA polymerase gets stuck in the termination fork and finally it is released, stopping the transcription.</p>	<p>It affects termination of transcription readthrough at a single discrete site that precedes a regulated gene.</p>
Postranscriptional Regulation			
<p>Antisense RNA: RNA-RNA REPRESSION</p> <p>Example: <i>trpA</i> mRNA + Estres oxidativo → <i>OxyS</i> RNA</p>  <p>ACTIVATION</p> <p>Example: <i>α35</i> mRNA + Temperatura baja → <i>DsrA</i> RNA</p> 	<p>RNA</p> <p>It acts by pairing with their target mRNAs.</p>	<p>RNA</p> <p>The target is the mRNA with complementary sequence</p>	<p>Regulation of synthesis of proteins, by affecting mRNA translation</p>



High throughput transcription start sites (TSSs) mapping. Active annotation

Aim: To enrich our understanding of transcriptional regulation by mapping as many TSS as possible to identify:

Promoters

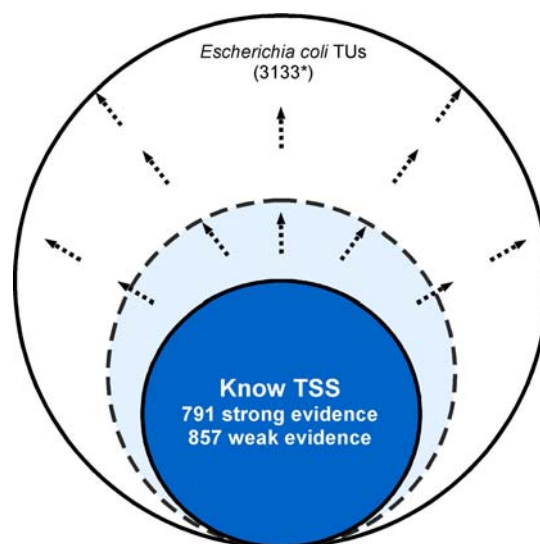
Transcription Factors Binding sites

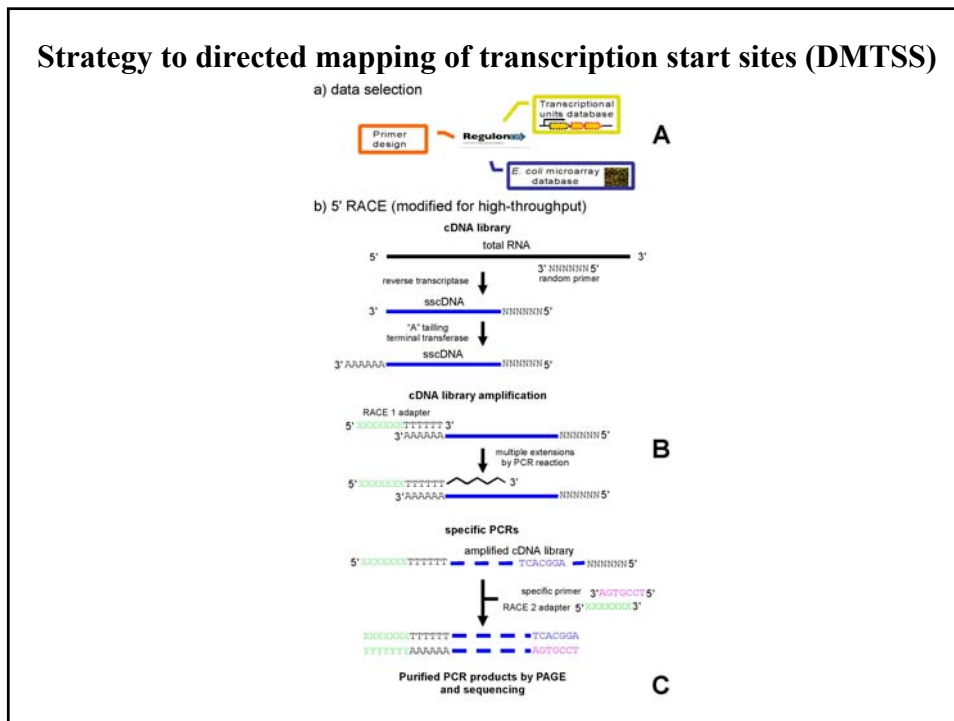
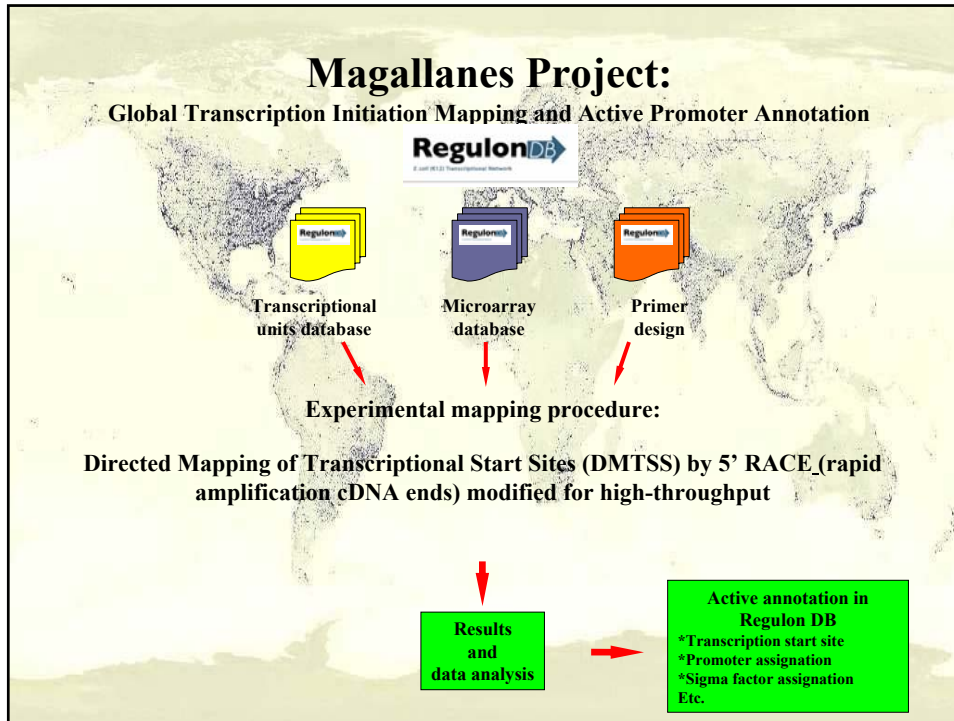
Operon structure

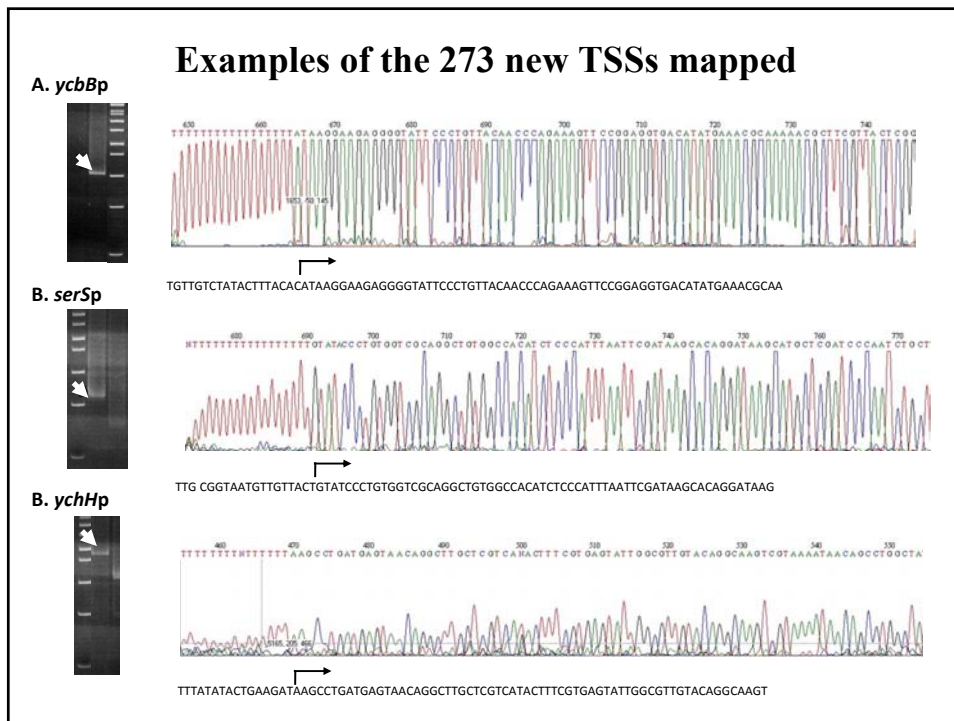
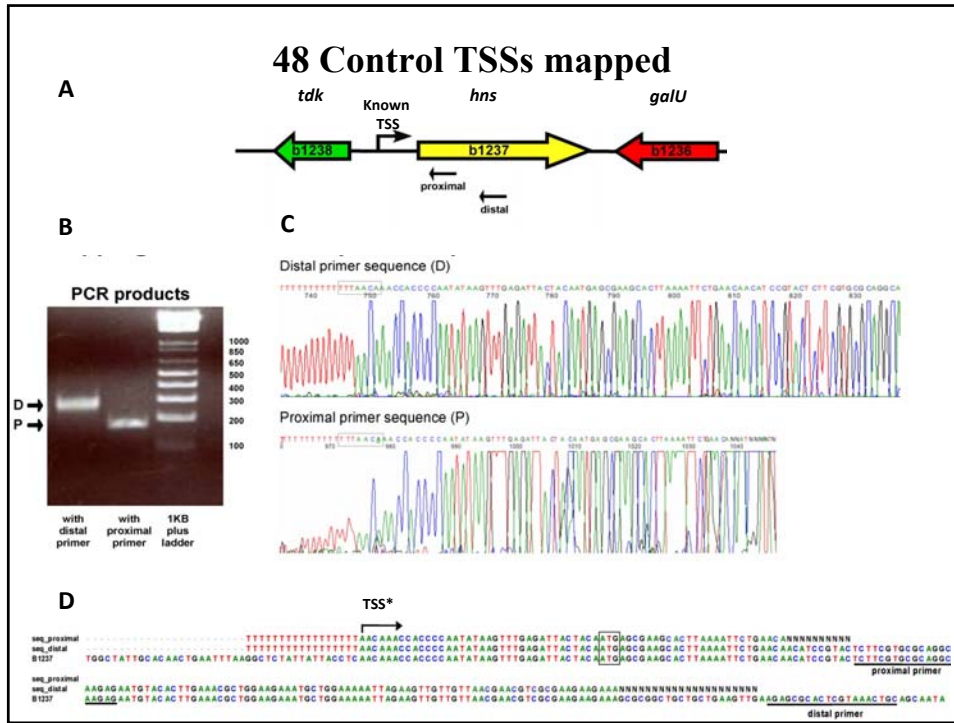
In collaboration with Enrique Morett, IBT, UNAM



Our limited knowledge of promoters in *E. coli*

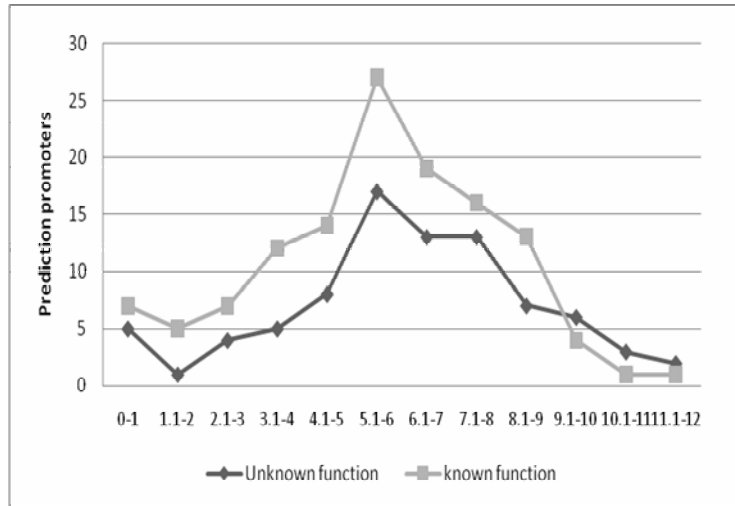






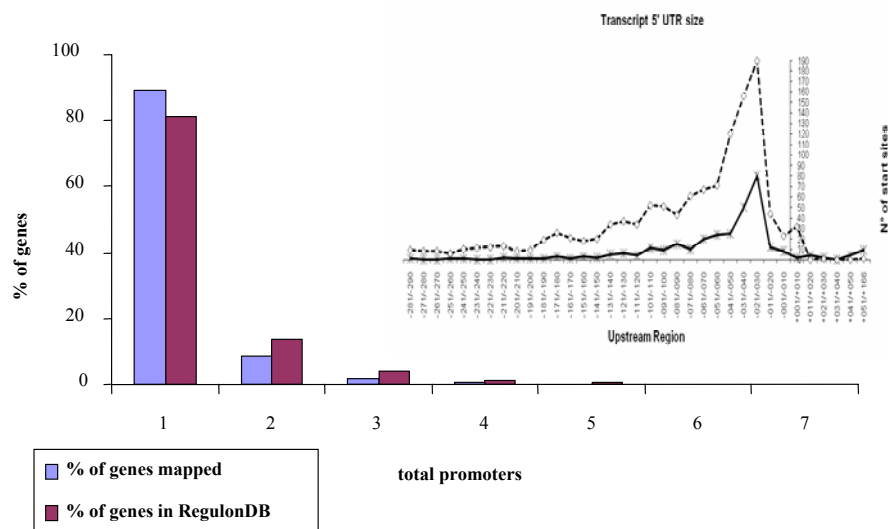
Promoter scores of genes with known or unknown function

About half of the new TSS are for genes with unknown function
Are they the product of real promoters?

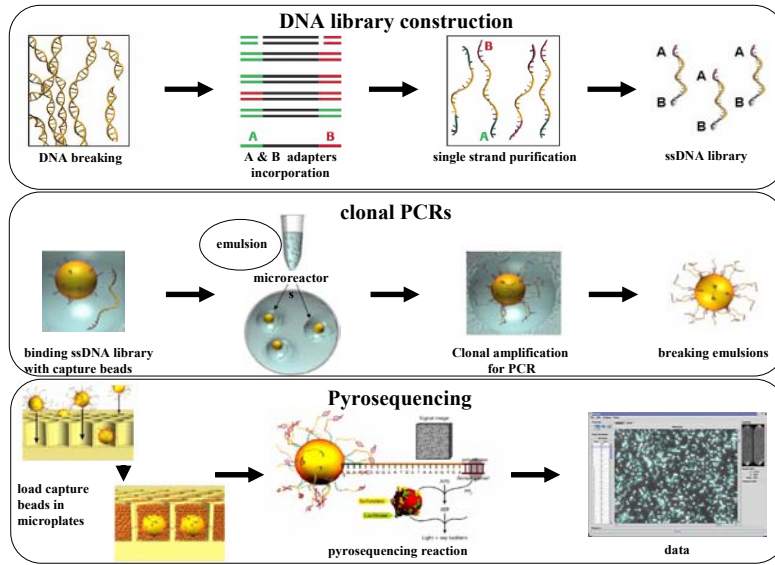


Total promoters by gene

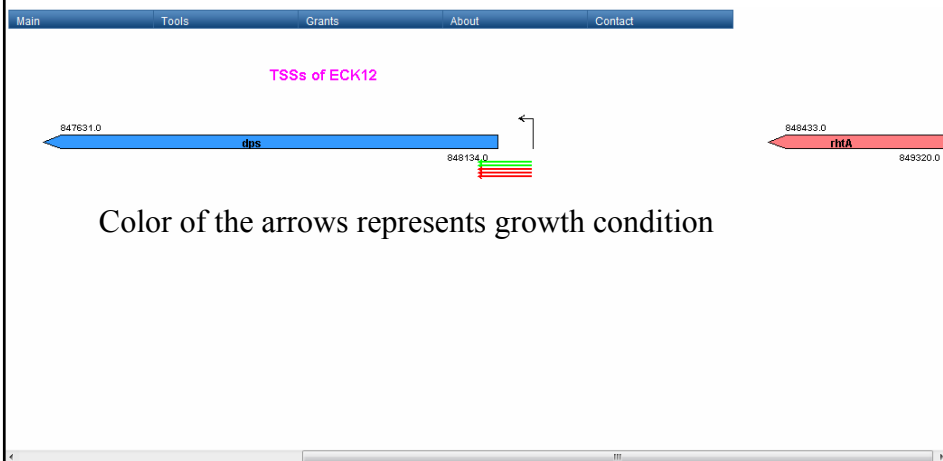
Distance of the TSS from the initial ATG codon



Sequencing with 454 GS20 instrument Clonal and fast!



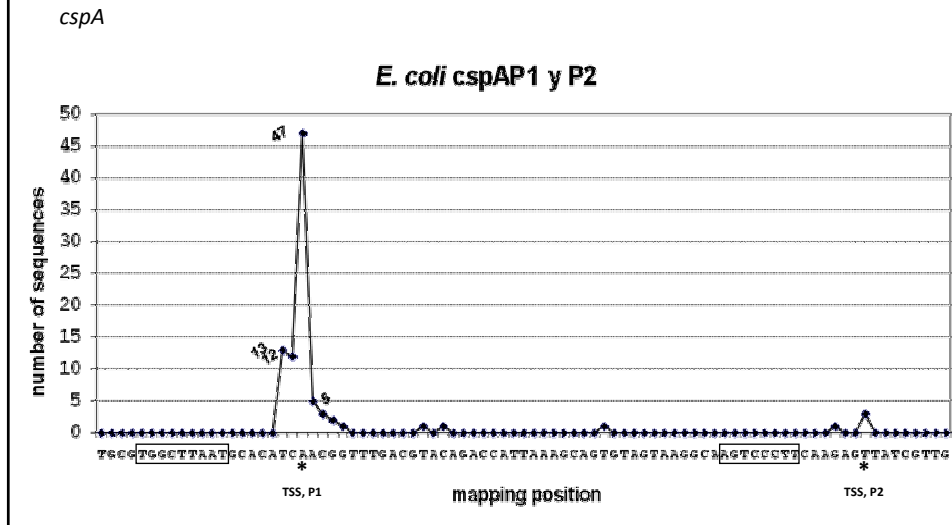
Examples of the pyrosequencing results: Multiple sequences of known TSS.



Color of the arrows represents growth condition

Confirmation of the recently mapped *dps* TSS by S. Busby's lab

Transcription not always starts at an exact nucleotide!



Results of Pyrosequencing

More than 1000 TSSs mapped. Only the high quality ones will be incorporated in RegulonDB

New information about relative expression in different conditions.

New information about mechanism of transcription initiation.

Computational Genomics Program



Funding: NIGMS, UNAM and Conacyt

Undergraduate program in genomics

4.4.5 Mapa curricular del plan de estudios propuesto

MAPA CURRICULAR DE LA LICENCIATURA EN CIENCIAS GENÓMICAS (2008)						
SEMESTRE	ASIGNATURAS					TOTAL SEMESTRE
NIVEL BÁSICO						
1	Matemáticas 1 (8C,4HT)	Principios de Programación (8C,3HT,2HP)	Biología Celular (8C,4HT)	Bioquímica (8C,4HT)	Biología Molecular (8C,4HT)	Seminario 1 (8C,3HT,2HP)
2	Matemáticas 2 (8C,4HT)	Computación (8C,3HT,2HP)	Principios de Estadística (8C,3HT,2HP)	Genética (8C,4HT)	Principios de Evolución (8C,4HT)	Seminario 2 (8C,3HT,2HP)
NIVEL AVANZADO						
3	Matemáticas 3 (8C,4HT)	Bioinformática y Estadística 1 (8C,3HT,2HP)	Genómica Funcional 1 (8C,4HT)	Genómica Evolutiva 1 (8C,4HT)	Modelos Genómicos (8C,3HT,2HP)	Seminario 3 (8C,3HT,2HP)
4	Matemáticas 4 (8C,4HT)	Bioinformática y Estadística 2 (8C,3HT,2HP)	Genómica Funcional 2 (8C,4HT)	Genómica Evolutiva 2 (8C,4HT)	Genómica Humana (8C,3HT,2HP)	Seminario 4 (8C,3HT,2HP)
NIVEL INTEGRATIVO						
5	Genómica Integrativa 1 (8C,3HT,2HP)	Genómica Integrativa 2 (8C,3HT,2HP)	Fronteras de la Genómica 1 (8C,4HT)	Fronteras de la Genómica 2 (8C,4HT)	Aplicaciones de la Genómica 1 (8C,4HT)	Aplicaciones de la Genómica 2 (8C,4HT)
6	Genómica Integrativa 3 (8C,3HT,2HP)	Genómica Integrativa 4 (8C,3HT,2HP)	Fronteras de la Genómica 3 (8C,4HT)	Fronteras de la Genómica 4 (8C,4HT)	Aplicaciones de la Genómica 3 (8C,4HT)	Aplicaciones de la Genómica 4 (8C,4HT)
NIVEL DE INVESTIGACIÓN						
7	Trabajo de Investigación 1 (6C,3HP)	Trabajo de Investigación 2 (6C,3HP)	Trabajo de Investigación 3 (6C,3HP)	Tópico Seleccionado 1 (10C,5HT)	Tópico Seleccionado 2 (10C,5HT)	Seminario de Investigación 1 (10C,4HT,2HP)
8	Trabajo de Investigación 4 (6C,3HP)	Trabajo de Investigación 5 (6C,3HP)	Trabajo de Investigación 6 (6C,3HP)	Tópico Seleccionado 3 (10C,5HT)	Tópico Seleccionado 4 (10C,5HT)	Seminario de Investigación 2 (10C,4HT,2HP)

COLD SPRING HARBOR LABORATORY Ph.D. fellowship



ensum académico 3622

QuickTime™ and a decompressor are needed to see this picture.

