## genome-wide studies using biocyc

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- 1 | functional profiling
- 2 | partial reconstruction
- 3 | functional associations
- 4 | evolutionary profiling
- 5 | network inference
- 6 | ancestral reconstruction

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# networks: function

## Metabolic pathways

- Profiling metabolic maps Genome Res 10:568, 11:1503
- Conservation of metabolism Genome Res 13:422
- Automatic pathway reconstruction, MjCyc Archaea 1:223
- Partial-genome reconstruction J Bioinfo Comput Biol 2:589
- Pathways for 160 genomes Nucl Acids Res 33:6083

## Functional modules, interaction networks

- Detection of functional modules by clustering Proteins 54:49
- Ancestral state reconstruction of interactions Mol Biol Evol 21:1171
- Exponential distribution of interactions Mol Biol Evol 22:421
- Functional associations in gene networks Alonso Cases & CAO, unpublished:2004



Dec 2005



# genomes: evolution

## Sequence clustering

All genomes

**Bioinformatics 19:1451** 

- All phylogenetic profiles
- All protein families
   Nucl Acids Res 31:4632
- All sequence similarities, ortholog clusters
- All genome based trees Nucl Acids Res 33:616
- All gene fusions

Genome Biol 2:r0034.1

### Genome evolution patterns

Comparison of different cellular processes

**Trends Microbiol 11:248** 

- Reconstruction of functional modules from genome constraints Proc Natl Acad Sci USA 100:15428
- Ancestral state reconstructions with loss and HGT Genome Res 13:1589
- Functional content of last universal common ancestor Res Microbiol in press:2005



Enright • Kunin • Kreil • Cases • Darzentas • Iliopoulos • Audit • Goldovsky

# algorithms

maine> cast SRm160.f
>SRm160

**CAST** for low-complexity masking

Bioinformatics 16:915

geneRAGE for domain clustering

Bioinformatics 16:451

**TRIBE-MCL** for rapid clustering

• Nucl Acids Res 30:1575

bioLayout for similarity visualization

Bioinformatics 17:853

geneTRACE for ancestral reconstruction

Bioinformatics 19:1412

TextQuest for document clustering Pac Symp BioComp 6:384







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# genome analysis in COGENT

## Closely following timed releases COGENT/++ as resource

### **Bioinformatics 21:3806**

- COGENT: complete genomes
- ProXSim: similarity information
- AllFuse: interactions by fusion
- Ofam: putative orthologs
- TRIBES: putative homologs
- ProfUse: phylogenetic profiles
- GPS: genome phylogenetic trees
- MeRSy: predictions from BioCyc
- GeneQuiz: automatic annotation

## Queriable by:

- sequence, via BLAST
- identifier, via MagicMatch

## MagicMatch

- >5 million links to major databases Bioinformatics 21:3429
- 12X UniProt in public-access data





# a sense of scale

Genomes	200 (243)
Genes	751,742 (915,554)
Annotations	359,482
Similarities	384,579,409
Phylogenetic profiles	181,986
Families	82,692
Interactions	2,192,019
Pathways	> 25,000

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# genome analysis with biocyc

**Examples of studies using the BioCyc environment** 

- Brief descriptions of what has been done
- Discussion of what else can be done...

Three cases of single-species analysis

Three cases of multiple-species analysis

All these computations external to COGENT
 ... except last (most complex) case, where the power of combining both systems is showcased



# 1 | functional profiling < >

## Analysis of known metabolic complement of E. coli

correlations between compounds, reactions, enzymes, pathways

Genome Res 10:568

Identified promiscuous reactions in terms of enzymes and pathways (example in figure below)

Model case study for comparative genomics, still unexplored territory

Today possible, for curated or automatically generated maps



**Figure 6** Diagram showing the number of reactions that are catalyzed by one or more enzymes. Most reactions are catalyzed by one enzyme, some by two, and very few by more than two enzymes.



**Figure 8** Diagram showing the number of reactions that participate in one or more pathways.

sequenced to date have identified virtually no multi-

## Global Properties of the Metabolic Map of *Escherichia coli*

#### Christos A. Ouzounis<sup>1,3</sup> and Peter D. Karp<sup>2</sup>

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# 2 | partial reconstruction < >

## **Benchmark on partial genomes**

 Simulated data from the S. pombe project

## Robustness and predictive power of reconstruction

J Bioinfo Comput Biol 2:589

- Build total map from complete genome as reference
- Take datasets from time points during genome project
- Count unique reactions, enzymes and pathways and compare with reference

## **Detection with partial genomes**

- Phase transition at 50% ?
- Pathways detected at various degrees, without closure
- Alternatively, build all maps
- Initial idea, requires more comparative work



- % of total pathways (at least 1 reaction)
- % of total reactions
- $\rightarrow$  % majority of reactions present (at least 50% of the reactions)



Cases

# 3 | functional associations • •

What are the functional relationships in real gene networks?

unpublished:2004

Analysis of *E. coli* networks using OperonDB and BioCyc

 detected all networks and overlayed functional information

**Example**: what is the relationship between pathways and transcription units in gene network?

1

- red: same pathway
- blue: same transcription unit
- green: both same pathway and transcription unit
- if same TU, then 95% in same pathway

Applications in functional genomics, e.g. microarray-based networks





Correlate genome size and environmental niche with enzymes and transcriptionassociated proteins (4D)

Cases

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- Larger genomes have more TAPs
- Free-living organisms have more TAPs, intracellular organisms have less
- No clear patterns for enzymes
- Small genomes somewhat richer in enzymes
  - Indirect evidence that metabolism is more conserved





#### Taps Score



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Tsoka • Pereira-Leal

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# 5 | network inference 🔺 🕶

Integrate all context-based predicted protein interactions Cluster protein network (various methods) Compare predicted cluster vs. known pathway assignments

74% of 583 metabolic enzymes cluster in 119 modules with 84% average pathway specificity and 49% average sensitivity

> Proc Natl Acad Sci USA 100:15428

Much of bacterial metabolism is encoded in genome structure! Novel functions predicted







# more complicated stuff...



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Kunin • Ahrén • Goldovsky • Janssen

## genome similarity maps

## Genome trees from both gene content and average sequence similarity

- 153 genomes (in publication), >200 now
- >25 M pairs
   Nucl Acids Res 33:616

•  $\sum (A,B) \neq \sum (B,A)$ •  $S(A,B) = \min(\sum (A,B), \sum (B,A))$ • D=D1 or D2•  $D1=1-S/\min((\sum (A,A), \sum (B,B)))$ •  $D2=-\ln(S/(\sqrt{2} * \sum (A,A) * \sum (B,B) / \sqrt{(\sum (A,A)^2 + \sum (B,B)^2)})$ 





# genome conservation trees





## Family distributions on a guide tree

- Can be rRNA tree, gene content, genome conservation
- Original implementation on rRNA trees
- Terminal nodes (extant genomes) contain gene/protein families

### Assumptions

Kunin

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- When most clade members contain representative, indication of vertical descent
- When few clade members do not contain representative, widely distributed in neighborhood, indication of gene loss
- Interspersed family distribution across remote clades indicative of horizontal gene transfer

## The GeneTrace algorithm

### **Bioinformatics 19:1412**

- Input: Phylogenetic profiles of families and guide tree
- Inner nodes represent ancestral organisms (states)
- Start at terminal nodes towards the root
- Scoring of gain/loss of parental nodes equal to sum of daughter nodes
- Scores transformed to assignments of presence/absence of EACH family
- Tags certain families as candidates for horizontal gene transfer





Kunin



## **Overlay family profiles on trees**

 Play parsimonious guesswork game of gain (genesis + HGT) and loss

## **Parameter** calibration

- HGT's 'expected relative frequency' ≈ loss/HGT
- On average: gain ≈ loss Genome Res 13:1589

## **Relative contributions**

loss:genesis:HGT = 3:2:1

## **Reconstruct ancestral states**

Families

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Interactions

Mol Biol Evol 21:1171





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HPYL-2

HPYL-J

1241

1243

CCRE

1940

RPRO

NMEN-Z

NMEN-M

1556

1543

594



# tree or network?

## Tree representations of species relationships

- Drawback: dealing only with vertical inheritance
- Data:
  - 184 genomes
  - Bidirectional best hits (putative orthologs...)
  - 3 different guide trees: gene content, average similarity, genome conservation

## HGT champions

Organism	Average ortholog similarity	Gene content	Genome conservation	STRING	
Pirellula sp.	2	1	1	Absent	
Bradyrhizobium japonicum	3	3	2	4	
Erwinia carotovora	5	2	4	Absent	
Clostridium acetobutylicum	4	4	10	5	
Chromobacterium violaceum	6	10	9	Absent	

ER

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Gene content reconstruction: ancestral states Both current and ancient genomes

- HGT events on tree
- Not directional!??
- **Power-law patterns:** 
  - Number of HGT's between any nodes
  - HGT network connectivity
    - Genome Res 15:954





# 6 | ancestral reconstruction - -

The gene content of the Last Universal Common Ancestor

 Reconstruction of ancestral states for 37,402 families from 184 genomes

Depending on method: 1,006 to 1,189 protein families

 Extremely robust estimates, based on ancestral state reconstruction, parsimony

Notion of similarity to extant genomes of obligate parasites not supported (also known as the 'minimal genome hypothesis')

> Res Microbiol in press:2005 (special issue on Exobiology)



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# **Computational Genomics Group**

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bioinformatics computer science biochemistry chemistry

statistics molecular biology bioinformatics computer science

fungal genomics theoretical physics microbiology parasitology bioinformatics fly genetics microbiology bioinformatics bioinformatics bioinformatics medical genetics cell biology biochemistry applied physics pattern discovery database design protein interactions metabolic pathways

database mining genome databases environmental genomics biological networks

metabolic profiling genome structure gene networks transcription factors sequence clustering text mining genome annotation compositional bias genome evolution error propagation human disease protein interactions phylogenetic profiling sequence identifiers

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# Q&A time...

Profile entire species with complete genomes, perform comparative analysis? Infer metabolic complement for species with incomplete genome information Overlay functional information with transcription profiles or inferred networks Compare the enzyme complement against entire functional categories Use the metabolic complement as a reference for functional module detection Reconstruct ancestral states of metabolism via enzymes, or even pathways?