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Systems Biology



MOL.921 Molecular Biotechnology II







³ What is Systems Biology?

• Systems Biology - The study of the mechanisms underlying complex biological processes as integrated systems of many interacting components.

Systems biology involves

(1) collection of large sets of experimental data

(2) proposal of mathematical models that might account for at least some significant aspects of this data set,

(3) accurate computer solution of the mathematical equations to obtain numerical predictions,

(4) assessment of the quality of the model by comparing numerical simulations with the experimental data.



Systems Biology

Analysis Understanding

Translation into Applications

Systems Biotechnology



Cells Networks Proteins COMPUTER SCIENCE

http://redpoll.pharmacy.ualberta.ca/bioinfo301/Systemsbiology 6.pdf



Systems Biology







What's it good for?

- Basic Science/"Understanding Life"
- Predicting Phenotype from Genotype
- Understanding/Predicting Metabolism
- Understanding Cellular Networks
- Understanding Cell-Cell Communication
- Understanding Pathogenicity/Toxicity
- "Raising the Bar" for Biologists

Making Biology a Predictive Science



Present Situation

- 100's of completed genomes
- 1000's of known reactions
- 10,000's of known 3D structures
- 100,000's of protein-ligand interactions
- 1,000,000's of known proteins & enzymes
- Decades of biological/chemical know-how
- Computational & Mathematical resources

The Push to Systems Biology





The Technologies of Systems Biology

- Genomics (HT-DNA sequencing)
- Mutation detection (SNP methods)
- Transcriptomics (Gene/Transcript measurement, SAGE, gene chips, microarrays)
- Proteomics (MS, 2D-PAGE, protein chips, Yeast-2-hybrid, X-ray, NMR)
- Metabolomics (NMR, X-ray, capillary electrophoresis)



Going From Technology to Systems Biology

- Genomics → Genometrics
- Proteomics → Proteometrics
- Metabolomics → Metabometrics
- Phenomics \rightarrow Phenometrics
- Bioinformatics \rightarrow Biosimulation

Quantify, quantify, quantify



Data aquisition

Single Unit Data $\leftarrow \rightarrow$ Comprising Data Sets

Single Pathways

← → Integrated Networks

Single Regulatory Units

Static Systems ← → Dynamic Systems





Single Systems – examples

Enzyme Catalyzed Reaction Step – Kinetics

Enzyme – catalytic mechanism

Regulatory Protein – Action on Transcription

Receptor – Interaction with Signal Molecule

Genome Analysis

Genome Sequencing

New Technologies – Ultrafast Sequencing

Future Vision: personal Genome \rightarrow < 1000 €

Sequence Analysis

Annotation

Comparative Genome Analysis

Functional Assays

Genome-wide Gene Knock out Mutagenesis

Genome-wide gene silencing





Next Generation Sequencing

Applied Biosystems Sanger sequencing

Technology : Sanger dideoxy sequencing method

Read Length : up to 1,000bp

Throughput : About 1Mb/day and machine

Roche/454 GS FLX

Technology : Pyrosequencing Read Length : Average 350-400 bp Throughput : 400 Mbp/run Single-read accuracy : > 99.5%

Applied Biosystems SOLiD

As a Certified Sequencing provider, Beckman Coulter Genomics' experience with the SOLiD sequencing platform surpasses that of any other sequencing provider.

Read lengths: 50 bp fragment, 25 bp and 35 bp paired Throughput (reads): > 160 million reads per slide, fragment

NCBI Data Formats: SOLiD Native or SRF

Robust di-base chemistry for accurate base-calling Color space analysis with AB proprietary software for high confidence SNP calls

www.beckmangenomics.com

Illumina Genome Analyser

Technology : Reversible terminators and Clonal Single Molecule Array technology Read Length : 36 bp, up to 75 bp (and growing) Throughput : 1Gbp per single read, > 3Gbp for paired-end libraries Per base read accuracy : > 98.5%

http://www.dnavision.com

Pyrosequencing







Luciferase

Figure 12.1 Pyrosequencing





The Roche 454/GS FLX Sequencing Technology





The Roche 454/GS FLX Sequencing Technology







http://www.genengnews.com/gen-articles/third-generation-sequencing-debuts/3257/?page=2



The Roche 454/GS FLX Sequencing Technology



The GS FLX sequencer supports sequencing of various different nucleic acid starting materials such as genomic DNA, PCR products, BACs and cDNA. Samples consisting of longer sequences are first sheared into a random library of 300-800 base-pair long fragments.

http://www.454.com/



The Roche 454/GS FLX Sequencing Technology



Aided by the adaptors individual fragments are captured on their own unique beads. A bead and the bound fragment together with a water-in-oil emulsion form a microreactor so that each fragment can be amplified without contamination via the so called emulsion PCR (emPCR). The entire fragment collection is amplified in parallel.



The emPCR amplifies each fragment several million times. After amplification the emulsion shell is broken and the clonally amplified beads are ready for loading onto the fibreoptic PicoTiterDevice for sequencing

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Pyro-Sequencing – 454 Technology



Figure 1. Scanning electron micrograph of etched well in 454 PicoTiter sequencing plate. 454's technology is based on performing hundreds of thousands of simultaneous sequencing reactions in 75 picoliter (44 μ m) wells. All molecular biology reactions—DNA amplification, sequencing by synthesis, and signal light generation—occur in a single well.

http://www.454.com/





Sequencing is accomplished by synthesizing the complementary strands of the bead attached templates. In a number of cycles the four bases (ATGC) are sequentially washed over the PicoTiterPlate. The incorporation of a new base is associated with the release of inorganic pyrophosphate starting a chemical cascade. This results in the generation of a light signal which is captured by a CCD camera.

http://www.454.com/



DNA Sequencing with Illumina (Solexa[®]) Technology

FIGURE 2: SEQUENCING TECHNOLOGY OVERVIEW



Randomly fragment genomic DNA and ligate adapters to both ends of the fragments.

Bind single-stranded fragments randomly to the inside surface of the flow cell channels.

Add unlabeled nucleotides and enzyme to initiate solid-phase bridge amplification.



DNA Sequencing with Illumina (Solexa®) Technology



- 5. DENATURE THE DOUBLE-STRANDED MOLECULES
- 6. COMPLETE AMPLIFICATION



The enzyme incorporates nucleotides to build double-stranded bridges on the solid-phase substrate.



Denaturation leaves single-stranded templates anchored to the substrate.



Several million dense clusters of doublestranded DNA are generated in each channel of the flow cell.

www.illumina.com/documents/products/.../techspotlight_sequencing.pdf



DNA Sequencing with Illumina (Solexa®) Technology



First chemistry cycle: to initiate the first sequencing cycle, add all four labeled reversible terminators, primers and DNA polymerase enzyme to the flow cell.



9. DETERMINE SECOND BASE



After laser excitation, capture the image of emitted fluorescence from each cluster on the flow cell. Record the identity of the first base for each cluster. Second chemistry cycle: to initiate the next sequencing cycle, add all four labeled reversible terminators and enzyme to the flow cell.

www.illumina.com/documents/products/.../techspotlight_sequencing.pdf



DNA Sequencing with Illumina (Solexa®) Technology



After laser excitation, collect the image data as before. Record the identity of the second base for each cluster.

Repeat cycles of sequencing to determine the sequence of bases in a given fragment a single base at time. Align data, compare to a reference, and identify sequence differences.

www.illumina.com/documents/products/.../techspotlight_sequencing.pdf





http://www.454.com/

PacBio Sequencing







Genomics Proteomics Bioinformatics 13 (2015) 278–289



2.6.16

Genome Analysis

Genome Sequencing

New Technologies – Ultrafast Sequencing

Future Vision: personal Genome → < 1000 €

Sequence Analysis

Annotation

Comparative Genome Analysis

Functional Assays

Genome-wide Gene Knock out Mutagenesis

Genome-wide gene silencing





Sequence Analysis - Annotation



Figure 12.6 Open reading frames: display of coding sequences; edited display from analysis of a DNA sequence and databank annotations using Artemis

Sequence Analysis - Comparison





Figure 12.11 Genome structure comparison using gMAP. (a) Comparison of entire genomes; (b) comparison of selection region 3, from (a); (c) detailed comparison of selected region 3, from (b)



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Transcriptome Analysis Bacteria grown Bacteria grown Genome sequence aerobically anaerobically 4000 ORFs identified mRNA extracted mRNA extracted cDNA amplified cDNA amplified and labelled and labelled 4000 PCR products with dye 1 with dye 2 spotted on a glass slide Mix and hybridize to the array 00000000 00000000 00000000 \bigcirc 00000 00000 Gene expressed in both (yellow) 00000 \odot Gene expressed aerobically only (red) 00 00 000000 000 Gene expressed anaerobically only (green)

Microarray







At the top are dried microdrops, each of which contains immobilized DNA strands from a different gene (*A*-*F*). These are hybridized with a mixture of fluorescence-labeled DNA samples obtained by reverse transcription of cellular mRNA. Competitive hybridization of red (experimental) and green (control) label is proportional to the relativeabundance of each mRNA species in the samples. The relative levels or red and green fluorescence of each spot are assayed by microscopic scanning and displayed as a single color. Red or orange indicates overexpression in the experimental sample, green or yellow-green underexpression in the experimental sample and yellow equal expression.

In sample 2, relative to sample 1, Gene D is moderately overexpressed, Gene E is equally expressed, and Gene F is moderately underexpressed.





Small part of a yeast DNA chip showing 1764 spots, each specific for hybridization with a different mRNA sequence. The color of each spot indicates the relative level of gene expression in experimental and control samples. The complete chip for allyeast open reading frames includes over 6200 spots. [Courtesy of Jeffrey P. Townsend, Duccio Cavalieri and the Harvard Center for Genomics Research]



RNA sequencing

RNA-seq: \rightarrow Transcriptome analysis

Isolation of mRNA (polyA+ with eukaryotes)

Reverse transcription (random primers or oiligo dT primers,

fragmentation, massive sequencing technology (NGS)

RNA-seq: \rightarrow Non-coding RNAs \rightarrow miRNA

Isolation of size selected RNA

Reverse transcription with specific primers



https://upload.wikimedia.org/wikipedia/commons/0/01/RNA-Seq-alignment.png











Glick et al., Molecular Biotechnology – Principles and Applications of Recombinant DNA, 4th ed., ASM Press

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FIGURE 6.22 RNA interference using siRNA (left) and miRNA (right).

Molecular Biotechnology, Carolyn A. Dehlinger, Jones and Bartlett Learning

Proteome Analysis



Figure 14.6 Two-dimensional gel electrophoresis



How to Do it? Three Types of Simulation







Atomic Scale 0.1 - 1.0 nm Coordinate data Dynamic data 0.1 - 10 ns Molecular dynamics

Meso Scale 1.0 - 10 nm Interaction data Kon, Koff, Kd 10 ns - 10 ms Mesodynamics

Continuum Model 10 - 100 nm Concentrations Diffusion rates 10 ms - 1000 s Fluid dynamics



How To Do it? (Computationally)





Some Problems...

- Almost all simulation systems are ultimately based on solving either:
- ordinary differential equations (ODEs),
- partial differential equations (PDEs)
- or stochastic differential equations (SDEs)
- Differential equations are "hard" to work with
- when simulating spatial phenomena,
- when dealing with discrete events (binding, switching), non continuous variables (low copy number) or
- when key parameters are unknown or unknowable





Some Problems...

- DEs are notorious for instabilities or situations where small rounding errors lead to singularities or chaotic behavior
- DE methods are not conducive to visualization or interactive "movies"
- DE methods require considerable mathematical skill and understanding (not common among biologists)
- DE methods don't easily capture stochasticity or noise (common in biology)
- Issue of realism cells don't do calculus



Is There a Better Way?

- Sidney Brenner calls it "biological arithmetic not calculus"
- Needs to accommodate the discrete (binding, signaling) and continuous (substrate concentration) nature of many cellular phenomena
- Two new approaches which avoid DEs
- Petri Nets (stochastic and hybrid)
- Cellular automata or agent based methods



46 Petri Nets

- A directed, bipartite graph in which nodes are either "places" (circles) or "transitions" (rectangles)
- A Petri net is marked by placing "tokens" on linked or connected places
- When all the places have a token, the transition "fires", removing a token from each input place and adding a token to each place pointed to by the transition (its output places)
- Petri nets are used to model concurrent systems, particularly network protocols w/o differential eqs.
- Hybrid petri nets allow modelling of continuous and discrete phenomena



Petri Nets

Petri net (also known as a **place/transition net** or **P/T net**) is one of several mathematical modeling languages for the description of distributed systems.

A Petri net consists of places (P), transitions (T), and arcs.

Arcs run from a place to a transition or vice versa, never between places or between transitions

 → Petri nets offer a graphical notation for stepwise processes that include choice, iteration, and concurrent execution.
 → Petri nets have an exact mathematical definition of their execution semantics, with a well-developed mathematical theory for process analysis. Graphically, places in a Petri net may contain a discrete number of marks called tokens. Any distribution of tokens over the places will represent a configuration of the net called a marking



Petri Nets





λ phage control circuit

Predicted protein expression



Petri Nets – Limitations

- Not designed to handle spatial events or spatial processes easily
- Stochasticity is "imposed", it does not arise from underlying rules or interactions
- Does not reproduce physical events (brownian motion, collisions, transport, binding, etc.) that might be seen in a cell
- Petri Nets are more like a plumbing and valving control system



50 Cellular Automata

- Computer modelling method that uses lattices and discrete state "rules" to model time dependent processes
- a way to animate things
- No differential equations to solve, easy to calculate, more phenomenological
- Simple unit behavior -> complex group behavior
- Used to model fluid flow, percolation, reaction + diffusion, traffic flow, pheromone tracking, predator-prey models, ecology, social nets
- Scales from 10⁻¹² to 10⁺¹²

Cellular Automata



Can be extended to 3D lattice





Dynamic Cellular Automata

- A novel method to apply Brownian motion to objects in the Cellular Automata lattice (mimics collisions)
- Takes advantage of the scale-free nature of Brownian motion and the scale-free nature of heterogeneous mixtures to
 - allow simulations to span many orders of time (nanosec to hours) and space (nanometers to meters)





- Java application that uses Dynamic Cellular Automata (DCA) to model motions, interactions, transport and transformations at the meso-scale (10⁻⁸ to 10⁻⁶ m)
- Uses a square, 2D lattice to model processes, lattice squares are equivalent to 3x3 nm regions
- Molecular objects are moved randomly and interactions determined according to a set of interaction rules that are only applied when objects are in contact (collision detection)



⁵⁴ Sim Cell Interactions

- Five different types of molecules or objects allowed in SimCell:
- 1) small molecules,
- 2) soluble proteins,
- 3) membrane proteins,
- 4) DNA molecules, and
- 5) membranes
- Protein-ligand interactions reduced to relatively small number of possibilities
- Touch and Go (T&G)
- Bind and Stick (B&S)
- Transport (TRA)



Sim Cell Interactions

Touch & Go





Sim Cell Interactions





Sim Cell Interactions





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Sim Cell and Cell Simulation

- Ideal for model checking and validation
- Conceptually equivalent to spatially dependent stochastic
 Petri nets

• Universally applicable: Enzyme kinetics, diffusion, excluded volume, binding, vesicle fusion, osmotic lysis, osmotic pressure, genetic circuits, metabolism, transport, repression, signalling, cell division, embryo gene expression...

All from one tool!



- Systems Biology requires the integration of
 - data archiving,
 - experimentation and
 - novel computational approaches

• There is a clear need for bioinformatics to step up from the static "stamp collection" phase to thinking about systems in dynamic/interactive/integrated terms

New tools are needed to make this possible
 – consider DCA & Petri Nets



⁶⁰ Can we 'compute' a cell?



Inverse Metabolic Engineering MOL.921 Molecular Biotechnology II



Elementary Modes: Example





Schuster, S., Hilgetag, C., Woods, J.H. & Fell, D.(1996)

F.Srienc, University of Minnesota





Elementary Modes: Example









 $M_3 = (0 - 1 \ 0 \ 1 \ 1 \ 1)^T$

F.Srienc, University of Minnesota

Data aquisition

Single Unit Data $\leftarrow \rightarrow$ Comprising Data Sets

Single Pathways

 $\leftarrow \rightarrow$ Integrated Networks

Single Regulatory Units

Static Systems $\leftarrow \rightarrow$ Dynamic Systems