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# Modelling Methods: High Throughput Time-Course Data

Jaroslav Stark Centre for Integrative Systems Biology at Imperial College (CISBIC) and Department of Mathematics

### **Overview**

- Dynamic v Static
- Forward and Inverse Systems Biology
- Model Identifiability
- Example: p53 Activity



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- Dynamic v Static
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#### **Perturbations and Responses**

- To Understand Biological Systems We Need To:
  - Perturb the System
  - Observe Resulting Effect





### **Static v Dynamic**

- Static Perturbation
  - eg Gene Knockout
  - Measure Long Term Effect
- Dynamic Perturbation
  - eg Immune Activation by Pathogen
  - Measure Dynamic Response





#### **Time Course Data**

- To Understand Dynamic Response We Need Time-Course Data
- Increasingly Can Obtain Many Simultaneous
   Measurements of Different Variables
  - eg Microarray Time-Coures





#### **Time Series**

- Vast Body of Methods:
  - Signal Processing
  - Statistics
  - Nonlinear Dynamics
- Systems Biology Presents a New Challenge
  - Very Few Time Points
  - Many Simultaneous Variables





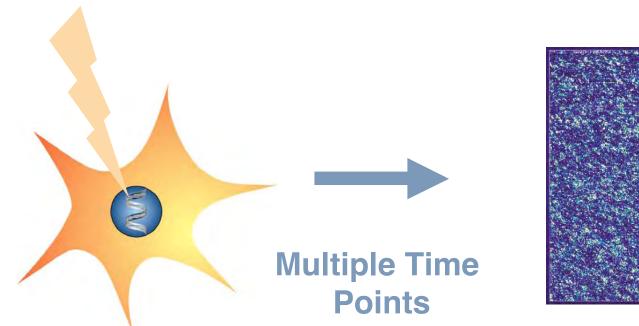
### **High-Throughput Time-Course Data**

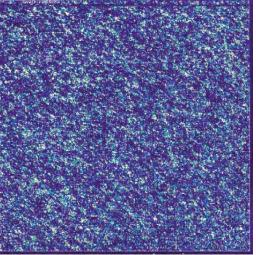
- Methods in Use
  - Very Simplistic By Time Series Standards
- Traditional Time-Series Methods
  - Typically Do Not Work With "Short Fat" Data



### **Example: DNA Damage Time-Course**

#### Irradiate Cells and Take Microarray Time Series



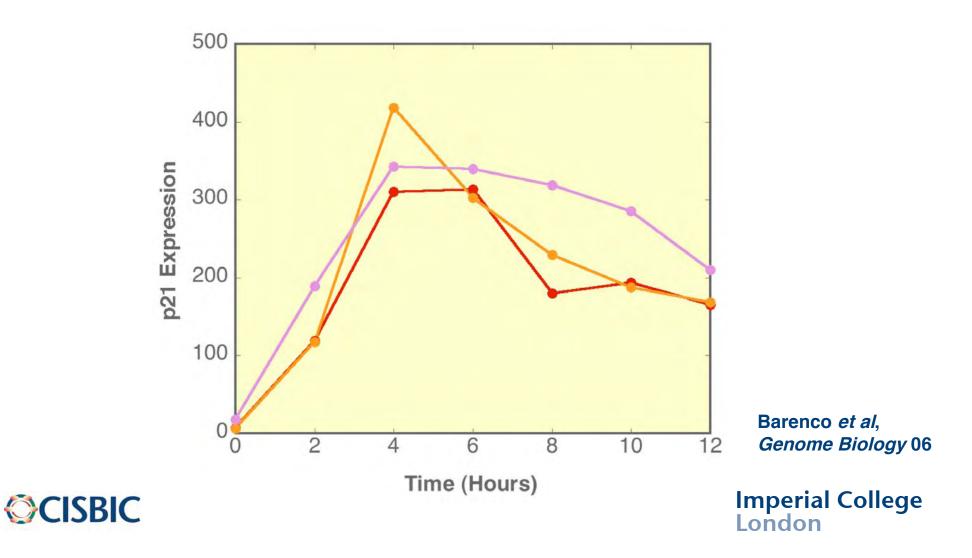




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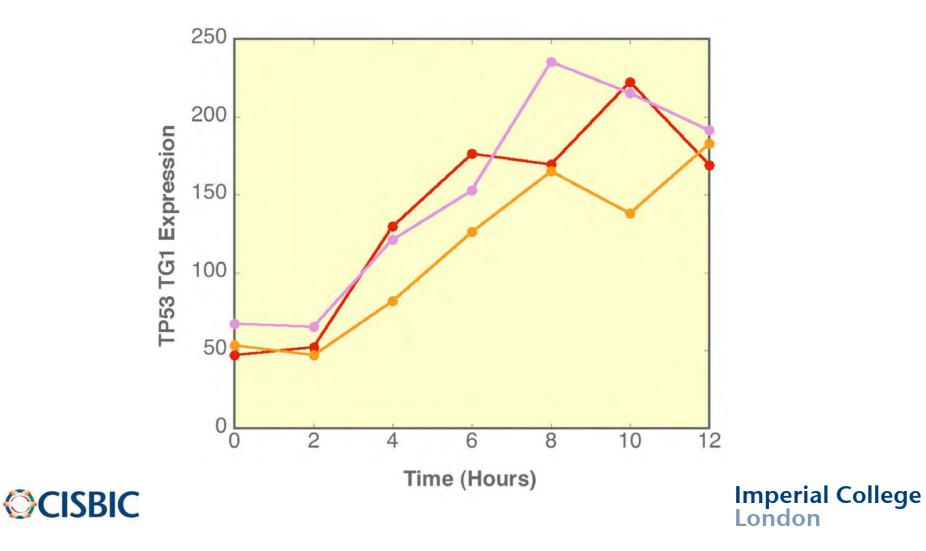
### **Measured Data For One Gene (p21)**

• Three Replicates



#### We Measure 19,000 Genes Simultaneously

Different Genes Have Different Dynamics



#### **Usual Analysis**

- Clustering or Principal Component Analysis
- Basic Principle: Group Genes With Similar Patterns of Activity





### **Usual Analysis**

- Clustering or Principal Component Analysis
- Basic Principle: Group Genes With Similar Patterns of Activity
- Key Assumption

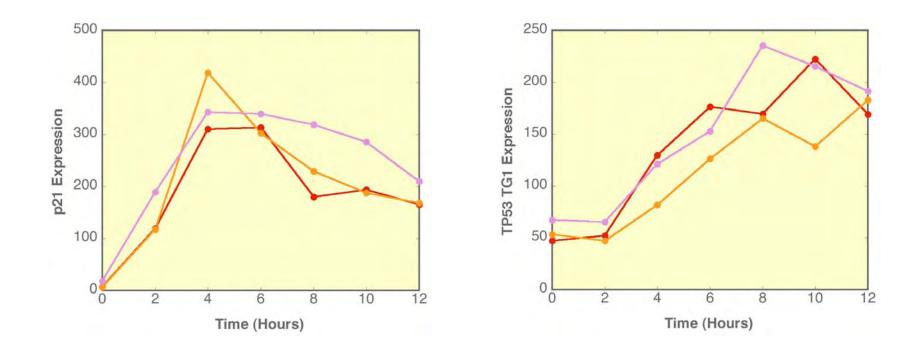
Similar Pattern = Similar Regulation Different Pattern = Different Regulation





#### But ...

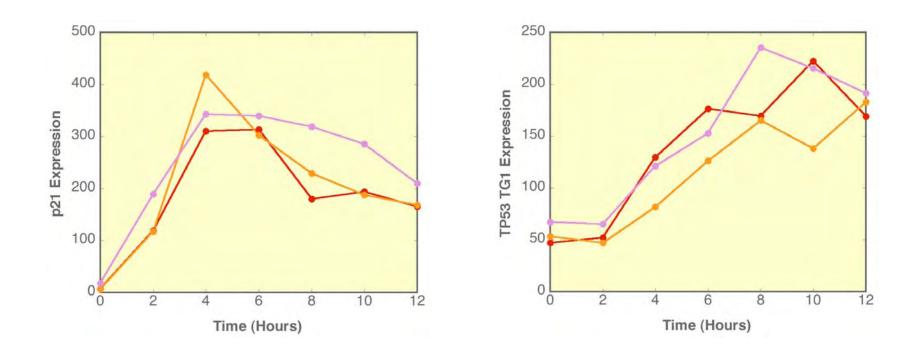
p21 and TP53 TG1 Are Both p53 Targets





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#### • Similar Production, Different Degradation





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### **Usual Analysis: Major Flaw**

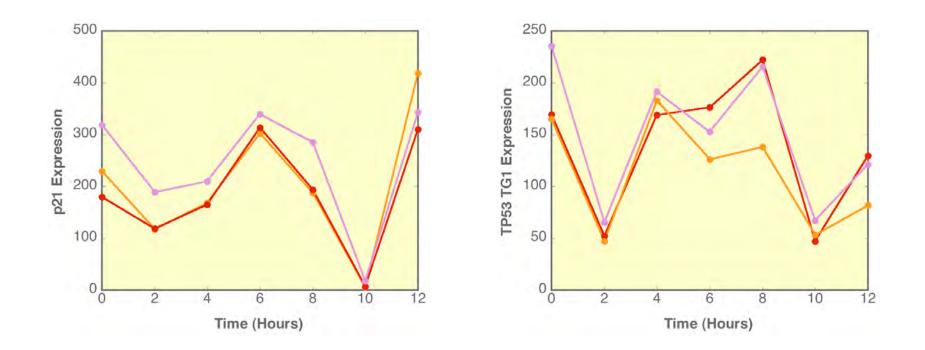
- Clustering or Principal Component Analysis
   Discards Important Information
- If We Scramble Data in Time
  - Recover Same Groups





#### **Permuted Time Labels**

• Different Patterns, Same Groups





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### **Usual Analysis: Major Flaw**

- Clustering or Principal Component Analysis
   Discards Important Information
- If We Scramble Data in Time
  - Recover Same Groups
- Standard Analysis Ignores Temporal Relationships





#### **Better Analysis**

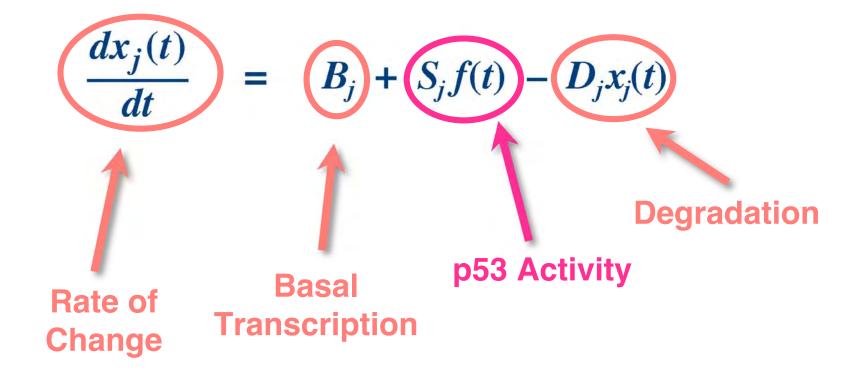
- Need Models That Explicitly Incorporate Time
- Data Is Very Crude
  - Use Very Simple Models





#### **Example: p53 Model**

 Ordinary Differential Equation for Expression Level x<sub>i</sub>(t) of Gene j

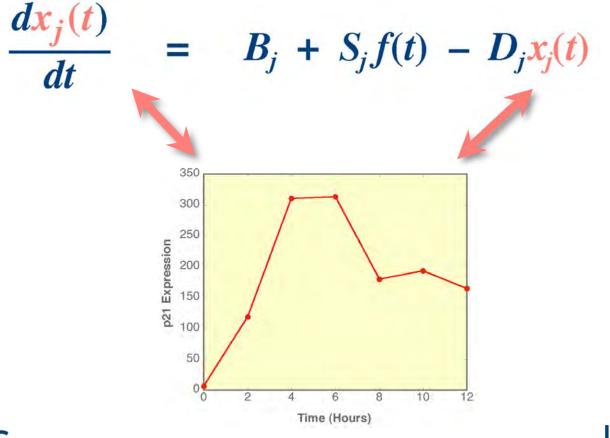




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#### **Problem: p53 Activity Is Not Measured**

Microarray Only Measures Expression x<sub>i</sub>(t)







### **Problem: p53 Activity Is Not Measured**

- p53 Activity Not Directly Controlled By Transcription
- p53 Activity Controlled By
  - Phosphorylation
  - Tetramerization
  - Ubiquitination
  - Translocation to Nucleus
- This Will Not Show Up in Microarray Data





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#### **Paradox of Molecular Biology**

- It Possesses Vast Amounts of Information
   About Individual Molecules
- It Needs to Integrate This Into An Understanding of the Whole System





## **Systems Biology: Forward Approach**

- Integration of Known Components
  - High Throughput Experiments
    - Measure All Potential Components
    - Measure All Potential Interactions
  - Modelling
    - Characterize Common Patterns
    - Suggest Functional Roles
    - "Understand System"





### **Paradox of Molecular Biology**

- It Possesses Vast Amounts of Information
   About Individual Molecules
- It Needs to Integrate This Into An
   Understanding of the Whole System
- Many Potential Experiments or Treatments
   Are Difficult or Impossible
  - Combinatorics of Multiple Gene Knockouts

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Dose-Time Course Experiments



#### **Three Approaches**

- Pretend Data is Correct and Complete
- Wait Until Data is Correct and Complete
- Develop Modelling Methods to Overcome
   Limitations





### **Systems Biology: Inverse Approach**

- Discovery of Hidden Effects
  - Models Based on Incomplete Information
  - Comparison to Incomplete Data
  - Prediction of New Components or Interactions



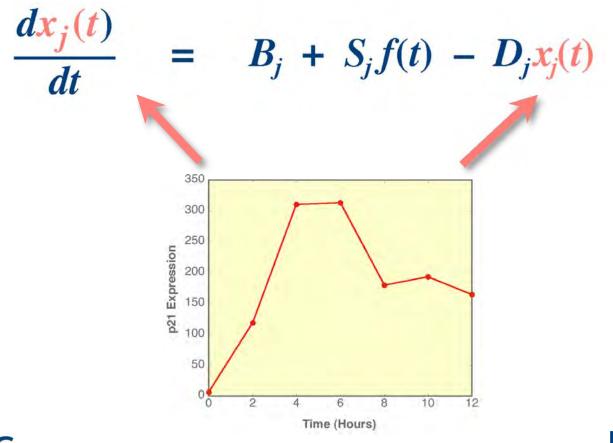
## **Systems Biology: Inverse Approach**

- Discovery of Hidden Effects
  - Models Based on Incomplete Information
  - Comparison to Incomplete Data
  - Prediction of New Components or Interactions
- Modelling is Useful With Limited Data



#### **Estimate p53 Activity**

Use 5 Known Targets to Fit This Equation

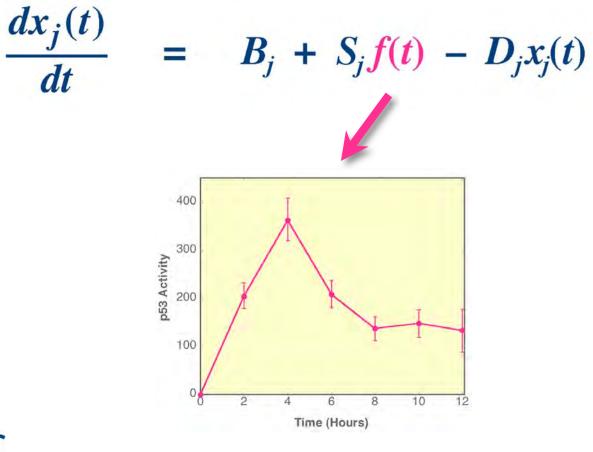






#### **Estimate p53 Activity**

Estimate Unknown p53 Profile







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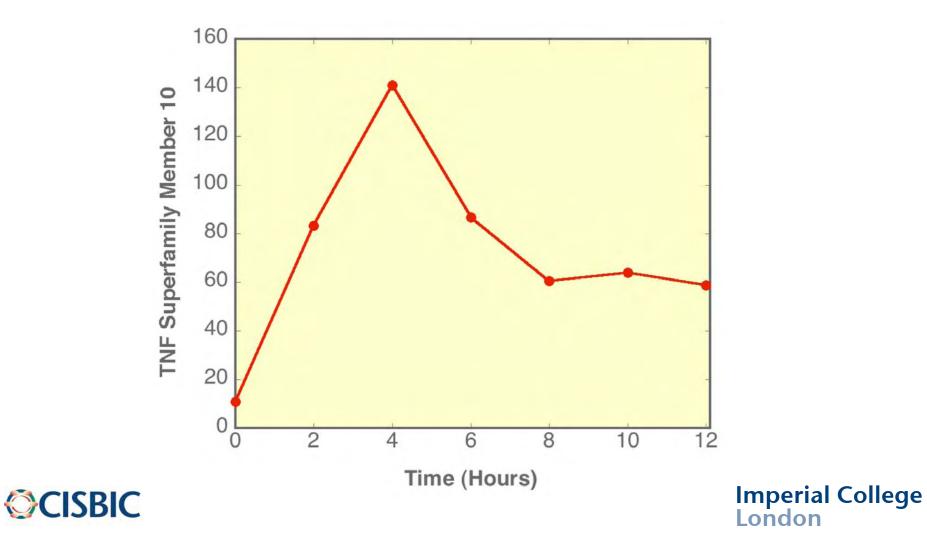
 Run Model at Two Different Parameter Combinations

$$\frac{dx_j(t)}{dt} = B_j + S_j f(t) - D_j x_j(t)$$

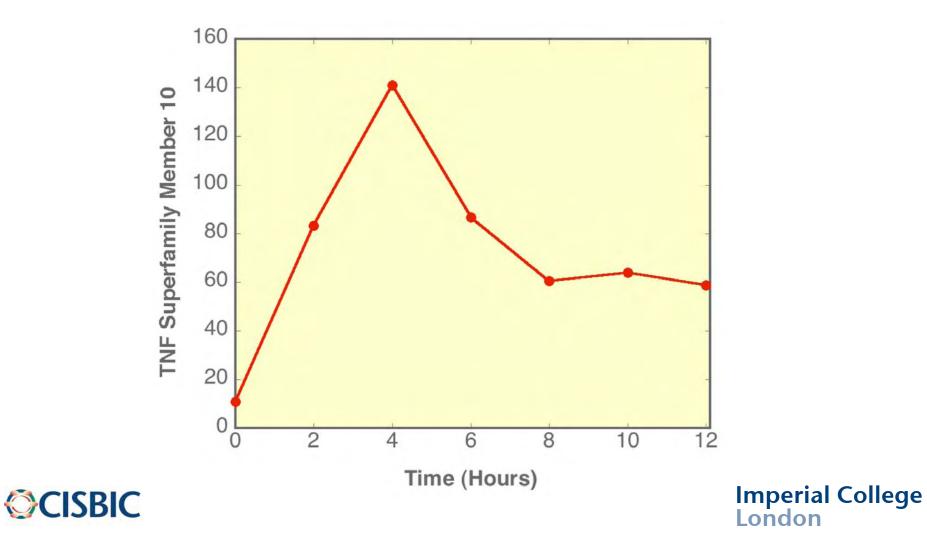




Parameter Combination 1



Parameter Combination 2



 Different Parameter Combinations Give Same Solution

$$\frac{dx_j(t)}{dt} = B_j + S_j f(t) - D_j x_j(t)$$





Trade-Off Between Basal and Induced Activity

$$\frac{dx_j(t)}{dt} = B_j + S_j f(t) - D_j x_j(t)$$





Trade-Off Between Basal and Induced Activity

 $B_j + S_j f(t)$ 





Trade-Off Between Basal and Induced Activity

 $(B_j + 1) + S_j f(t)$ 





Trade-Off Between Basal and Induced Activity

$$(B_j + 1) + S_j (f(t) - 1/S_j)$$





### **Model Non-Identifiability**

- Well Known Problem in Statistics and Control
  Theory
- Sontag, J Nonlin Sci, 12, (2002), 553-583.
  - Classic Operon Model
- Can Be Overcome by Careful Choices of Normalization
  - Reduces Number of Parameters to Be Estimated

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Tutorial

