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# Modelling of Cancer Treatment

Graeme Wake

in collaboration with

**Britta Basse, Auckland Medical School**

**Ronald Begg, University of Canterbury**

Bruce van-Brunt, Massey University

David Wall, University of Canterbury

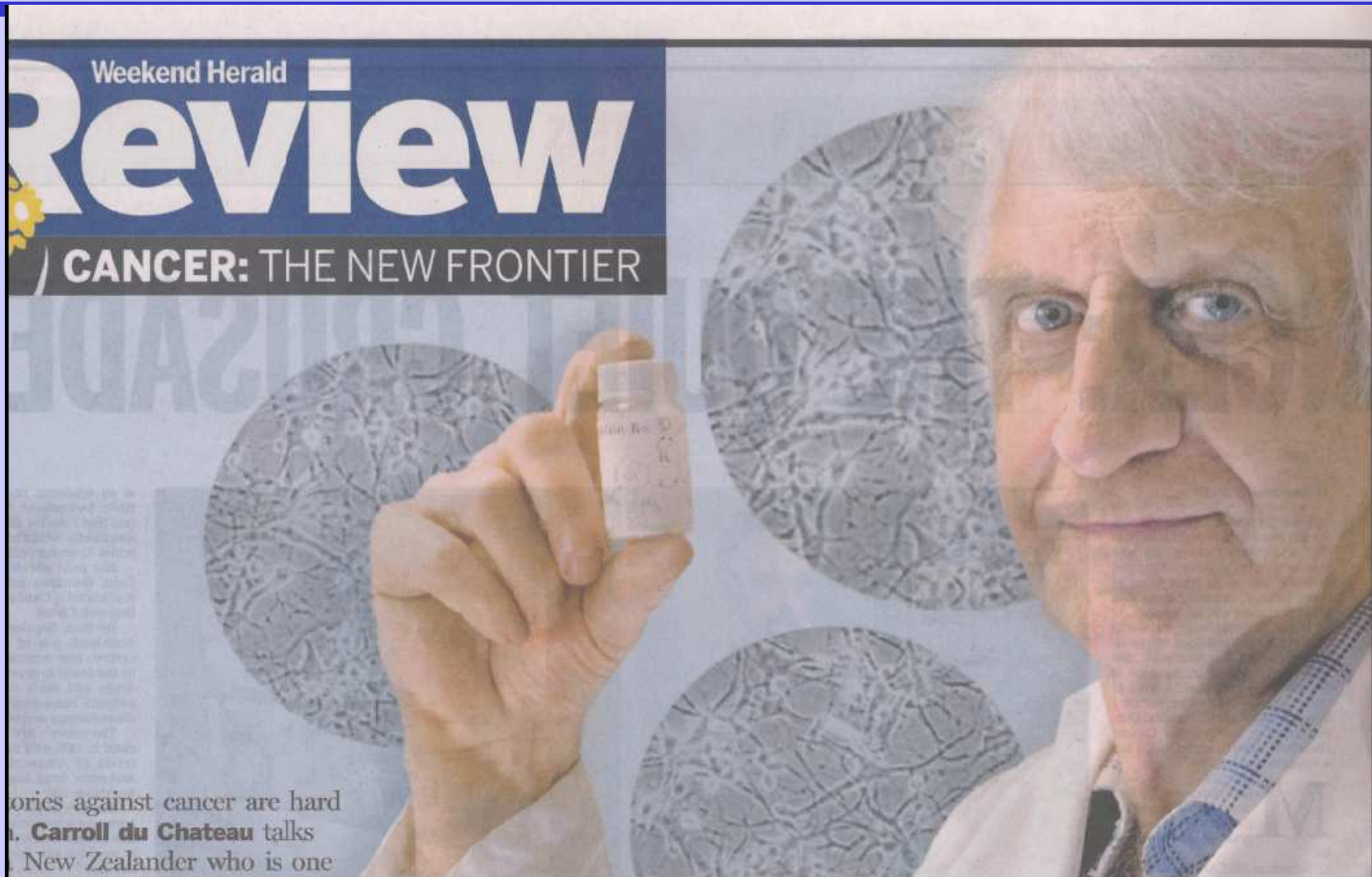
**Bruce Baguley, Auckland Medical School**

Elaine Marshall, Auckland Medical School

**Seminar**

**June-July 2008**

# Bruce Baguley; Auckland Cancer Society Professor



Graeme Wake, Centre for Mathematics in Industry, Massey University at Auckland, New Zealand



Graeme Wake, Centre for Mathematics in Industry, Massey University at Auckland, New Zealand

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- What is non-local Calculus?

Occurs when cause and effect are separated.

\* Earlier times affect the present.

\* Action at a distance.

- Test problems:

$$y'(t) = y(t-1), \quad y((-1,0]) = \text{given}.$$

$$y'(x) = \alpha y(\alpha x) - y(x), \quad y(0) = 0, \quad y \geq 0.$$

What is the solution?

# Outline

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1. History of the cell-growth model
2. The new mathematics!!!
3. Tumour cell growth
4. Further new maths
5. Modelling cancer treatment \*\*\*\* (main section)
6. Link to other approaches
7. Current work

# 1. History of the cell-growth model – personal.

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1988+: Horticulturists ask me to “provide an understanding of time-series data” which showed that cell populations, structured by size, evolved by simultaneously growing, dividing and dying, evolved to a

**“STEADY SIZE DISTRIBUTION”**

**SSD - first take-out.**

This data was for plant-root cells, maize etc.

This result was robust, independent of the initial condition, and in dynamical systems terms, was attracting.

# SSDs

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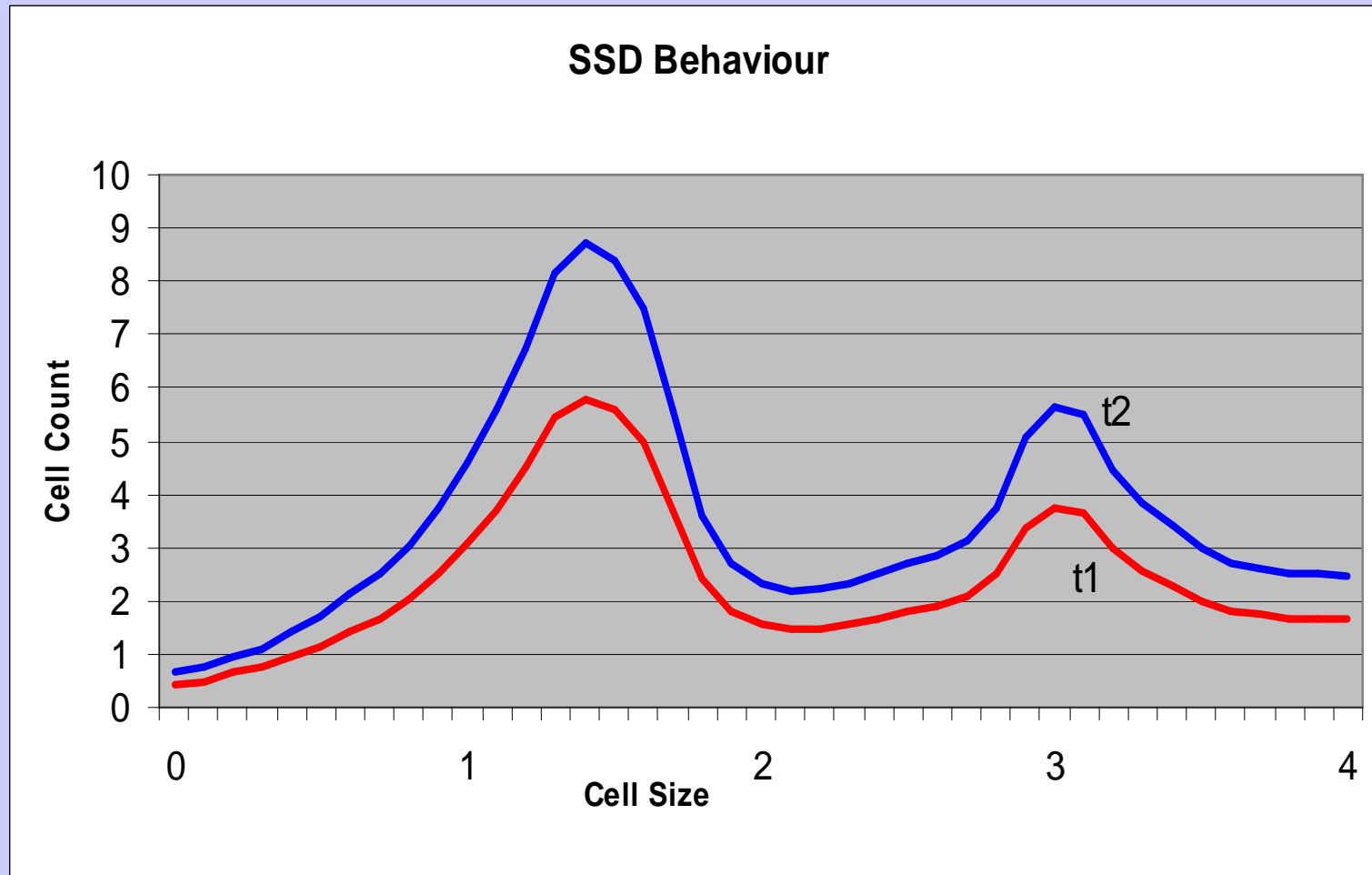
- What is a SSD? Introduce  $n(x,t)$ , the number density of a cell population cohort structured by attributes ( $x$ ) like:
  - \* size (say = DNA content) – this is us
  - \* age
  - \* time in a given phase....
  - etc

Then

$$\int_a^b n(x,t) dx = \# \text{ of cells (biomass) in size interval } [a,b], \text{ evolving in time.}$$



# SSD behaviour: $n(\cdot, t)$ evolves like:





# Core model

- This led to the first cell-growth model:

$$n_t = - (gn)_x + b\alpha^2 n(\alpha x, t) - bn - \mu n(x, t), \quad x, t > 0, \quad \alpha > 1. \text{ Why?}$$

↑
↑
↑
↑

growth
addition through division
loss through division
death

$$n(0, t) = 0, \quad n(\infty, t) = 0, \quad n(x, 0) \text{ given}, \quad n(x, t) \geq 0.$$

The terms are all local except “ $n(\alpha x, t)$ ”.

$$x = 0 \quad x/\alpha \quad \longleftarrow \quad x \quad \longleftarrow \quad \alpha x,$$

$\xrightarrow{\hspace{15em}} x$

# Key questions and answers

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- The question then is:

Are there solutions of the form

$$n(x,t) = N(t) y(x) = e^{-\lambda t} y(x) ? \quad \text{Q. sign of } \lambda????$$

**Yes there is:**

1. "A functional differential equation arising in modelling of cell growth". J Australian Math. Soc. Series B, Vol 30.424-435,1989 ( A J Hall and **G C Wake**).
2. "Functional differential equations determining steady size distributions for populations of cells growing exponentially",J.Austr.Math.Ser;B, Vol 31,434-453,1990 ( A J Hall and **G C Wake**).
3. "Steady size distributions for cells in one-dimensional plant tissues" Journal of Mathematical Biology. Vol 30. No 2. pp101-123. 1991 (A J Hall, **G C Wake** and P W Gandar).

**We then added dispersion, see later...**

4. "Functional differential equations for cell-growth models with dispersion" Comm. Appl. Anal. **4**, 2000, pp 561-574. (**G C Wake**, S Cooper, HK Kim, & B van-Brunt).

## 2. New Mathematics

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- The SSD is  $y(x)$ , which in the no-dispersive case satisfies the interesting non-local equation:

$$y'(x) = a\alpha y(\alpha x) - a y(x), \quad x > 0,$$

$$a = b\alpha/g,$$

$$y(0) = 0, \quad y(x) \geq 0, \quad \int_0^{\infty} y(x) dx = 1.$$

We prove in Reference 1 that this is well-posed and find  $y(x)$  Explicitly. How??

And the  $\lambda = \mu - b(\alpha-1) \lessgtr 0$ , in  $n(x,t) = e^{-\lambda t} y(x)$ .

*< Therefore this is a healthy growing cohort.*

*> Therefore this is a decaying cohort*

# Generic equation

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- This is akin to the “pantograph equation”
- Raised in the first MISG in Oxford in 1970
- Is generally ill-posed as an IVP:

“the future is dictating the past”

$$y'(x) = a\alpha y(\alpha x) - a y(x), \quad x > 0, \quad \alpha > 1,$$

$$y(0) = 0, \quad y(\infty) = 0.$$

- For us it is an “eigenvalue problem”, and we are at the principal eigenvalue.

# Normal Cells

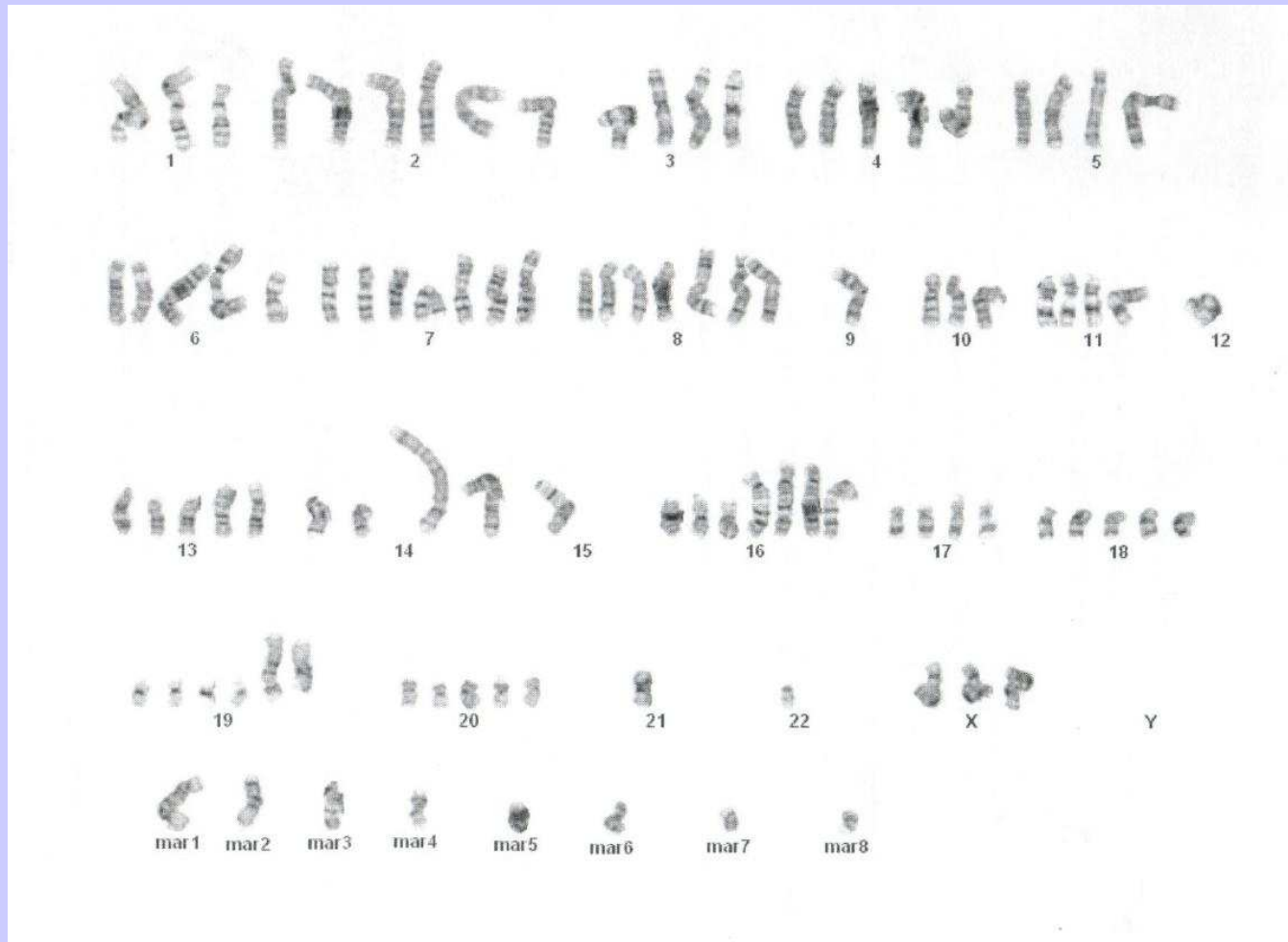
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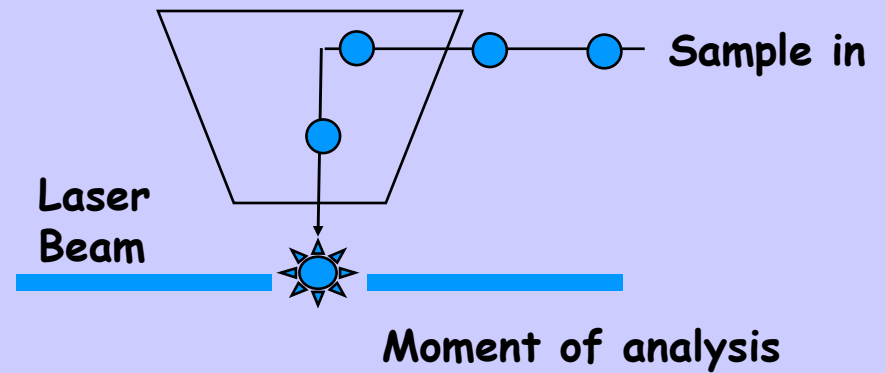
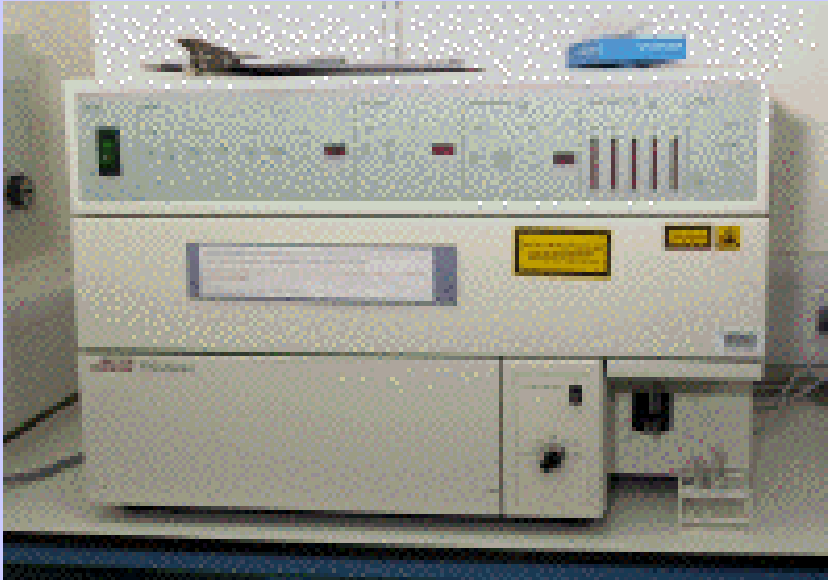
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# 3. Tumour cell growth



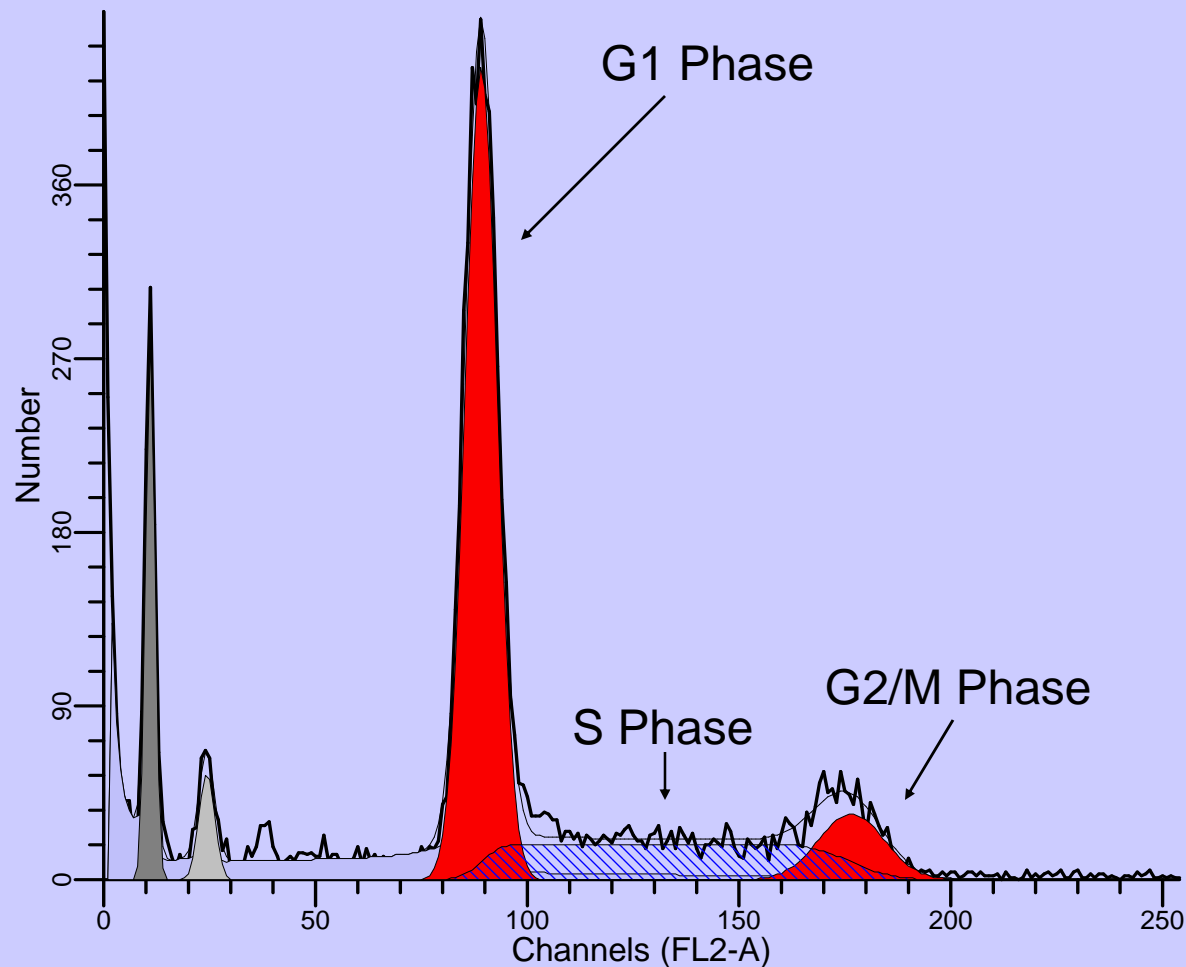
# Flow Cytometry

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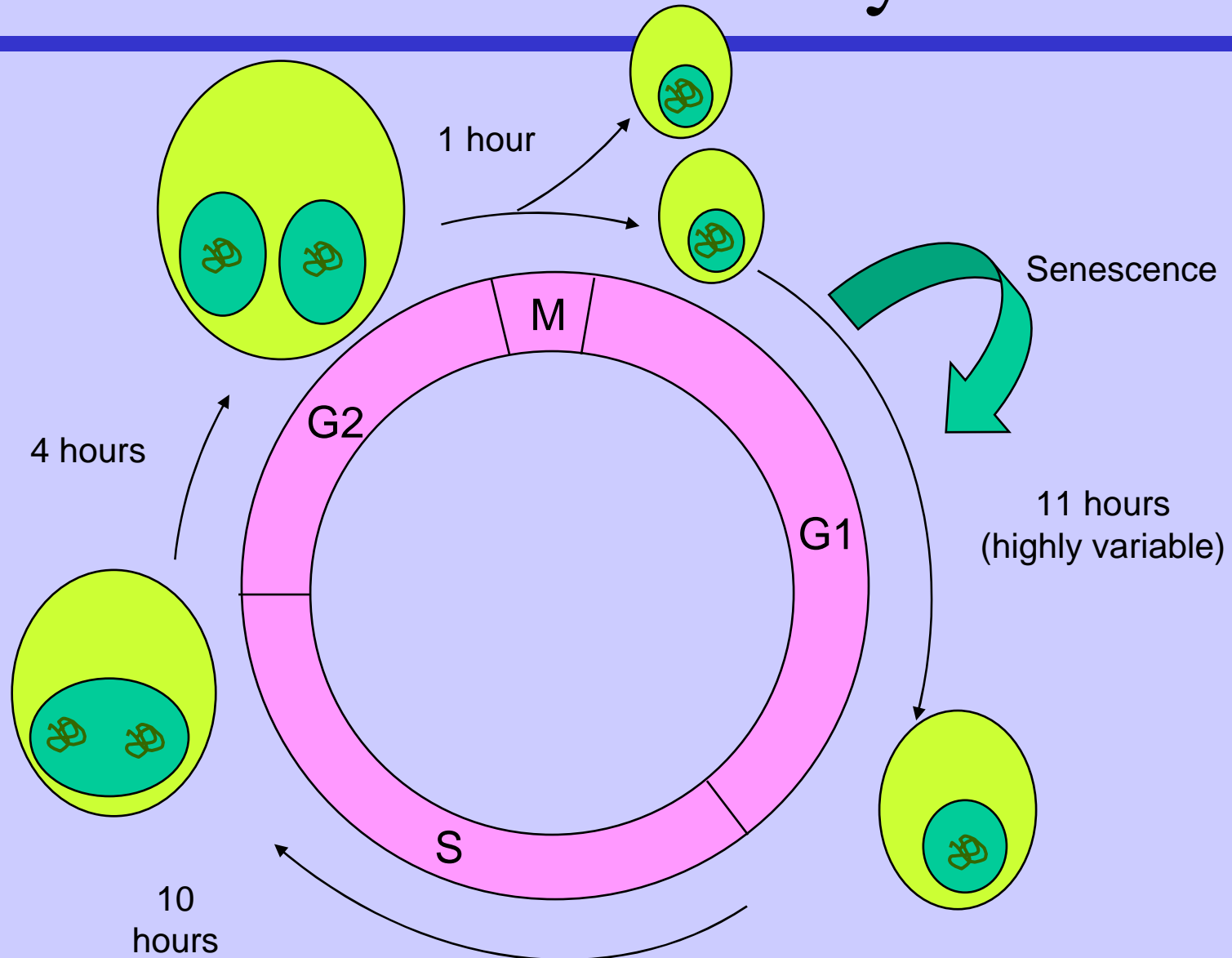




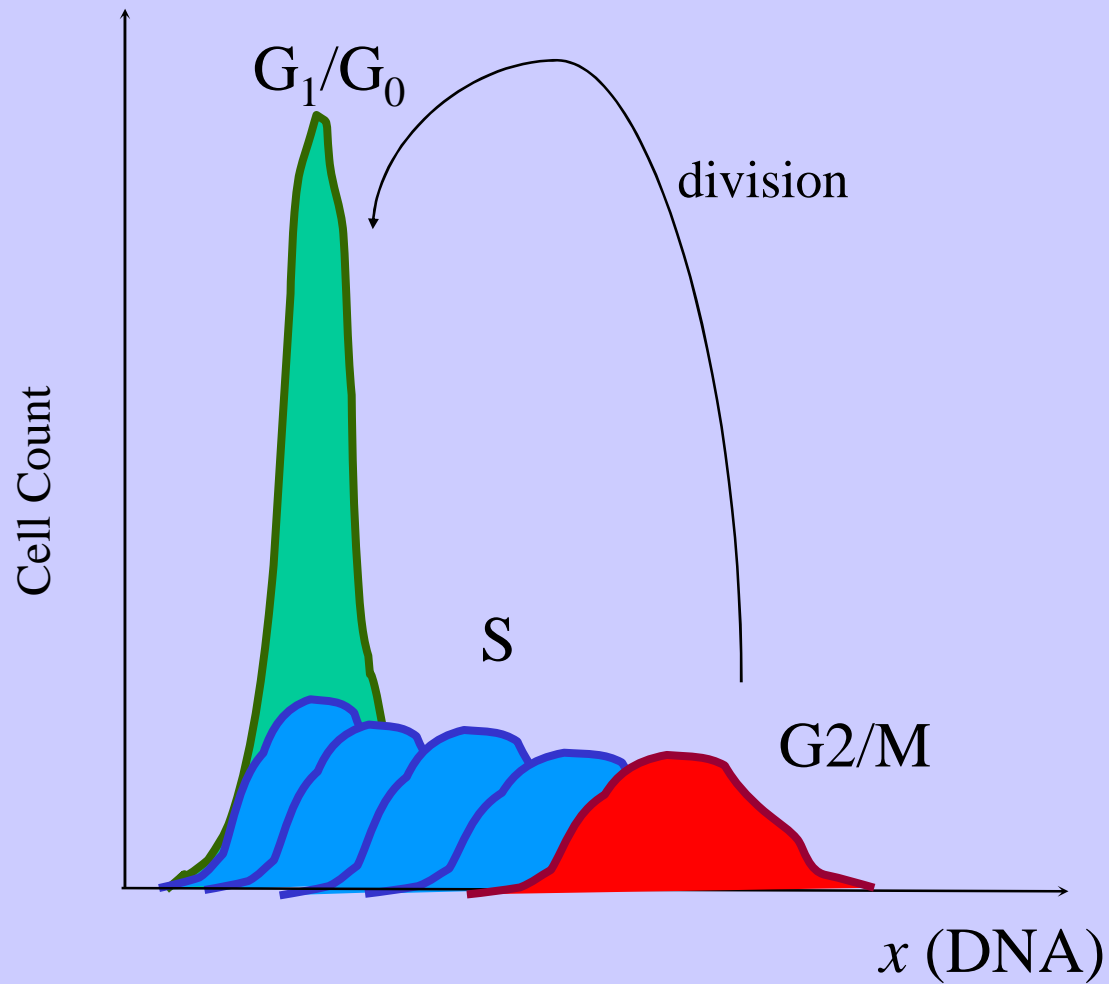
# Steady DNA Distribution (SDD)



# The Human Cell Cycle



# Idea



## Model of a cell line unperturbed by cancer therapy G1-phase

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$$\frac{\partial G_1(x,t)}{\partial t} = 4bM(2x,t) - k_1 G_1(x,t), \quad t > 0, 0 < x < L,$$
$$G_1(x, t=0) = G_{10}, \quad 0 < x < L,$$

# Dispersion

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- There is white noise particularly in the S-phase
- Growth  $dx = g dt + \sigma dX$

*Deterministic*      *White*  
*Growth*              *Noise*

- Gives Fokker-Planck Equation

$$S_t = (D S)_{xx} - (g S)_x + \dots \text{ etc.},$$

$$D = \sigma^2/2.$$

# Model of a cell line unperturbed by cancer therapy S-phase

$$\frac{\partial \bar{S}(x, t; \tau_s)}{\partial \tau_s} = D \frac{\partial^2 \bar{S}(x, t; \tau_s)}{\partial x^2} - g \frac{\partial \bar{S}(x, t; \tau_s)}{\partial x}, \quad t, \tau_s > 0, 0 < x < L,$$

$$\bar{S}(x, t; \tau_s = 0) = k_1 G_1(x, t), \quad t > 0, 0 < x < L,$$

$$\bar{S}(x, t = 0; \tau_s) = \bar{S}_{0\tau_s}, \quad \tau_s > 0, 0 < x < L,$$

$$D \frac{\partial \bar{S}}{\partial x}(x = 0, t; \tau_s) - g \bar{S}(x = 0, t; \tau_s) = 0, \quad t, \tau_s > 0,$$

$$D \frac{\partial \bar{S}}{\partial x}(x = L, t; \tau_s) - g \bar{S}(x = L, t; \tau_s) = 0, \quad t, \tau_s > 0,$$

$$S(x, t) = \int_0^{T_s} \bar{S}(x, t; \tau_s) d\tau_s$$

## Model of a cell line unperturbed by cancer therapy G2 & M-phase

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$$\frac{\partial G_2(x,t)}{\partial t} = \bar{S}(x,t;T_s) - (k_2 + \mu_{G_2})G_2(x,t), \quad t > 0, 0 < x < L,$$
$$G_2(x,t=0) = G_{20}, \quad 0 < x < L,$$

$$\frac{\partial M(x,t)}{\partial t} = k_2 G_2(x,t) - bM(x,t), \quad t > 0, 0 < x < L,$$
$$M(x,t=0) = M_0, \quad 0 < x < L,$$



$$\frac{\partial \bar{S}(x, t; \tau)}{\partial \tau} = D \frac{\partial^2 \bar{S}(x, t; \tau)}{\partial x^2} - g \frac{\partial \bar{S}(x, t; \tau)}{\partial x}, \quad \bar{S}(x, t; \tau = 0) = \bar{S}_0(x, t)$$

$$A < x < B$$

$$\bar{S}(x, t; \tau) = \int_A^B \bar{S}_0(x, t) \gamma(\tau, x, z) dz$$

$$-\infty < x < \infty$$

$$\gamma(\tau, x, z) = \frac{1}{2\sqrt{\pi D \tau}} e^{-(x-g\tau-z)^2/4D\tau}$$

$$0 < x < \infty$$

$$\bar{S}(x=0, t; \tau) = 0,$$

$$\gamma(\tau, x, z) = \frac{1}{2\sqrt{\pi D \tau}} e^{g(x-z-g\tau/2)/2D} \left( e^{-(x-z)^2/4D\tau} - e^{-(x+z)^2/4D\tau} \right)$$

$$0 < x < L$$

$$\bar{S}(x=0, t; \tau_s) = 0$$

$$\bar{S}(x=L, t; \tau_s) = 0$$

$$\gamma(\tau, x, z) = \frac{e^{g(x-z-g\tau/2)/(2D)}}{2\sqrt{\pi D \tau}} \sum_{n=-\infty}^{\infty} \left( e^{-(x-z+|n|2L)^2/4D\tau} - e^{-(x+z-|n|2L)^2/4D\tau} \right)$$

$$0 < x < L$$

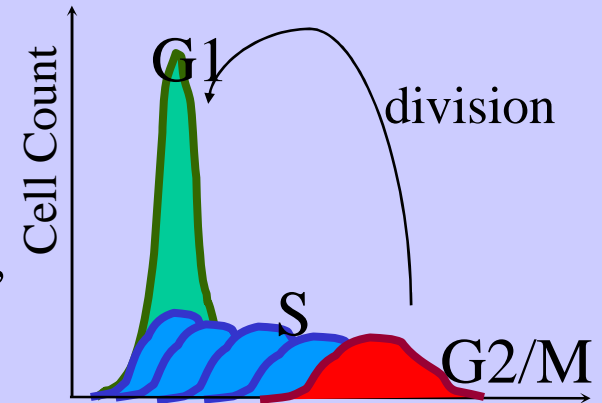
$$D \frac{\partial \bar{S}}{\partial x}(0, t; \tau_s) - g \bar{S}(0, t; \tau_s) = 0,$$

$$D \frac{\partial \bar{S}}{\partial x}(L, t; \tau_s) - g \bar{S}(L, t; \tau_s) = 0,$$

$$\gamma(\tau, x, z) \cong \frac{1}{2\sqrt{\pi D \tau}} e^{-(x-g\tau-z)^2/4D\tau}$$

# S-phase equations

$$\bar{S}(x, t; \tau_s) = \begin{cases} k_1 G_1(x - g \tau_s, t - \tau_s), & D = 0, \\ \int_0^L k_1 G_1(x, t - \tau_s) \gamma(\tau_s, x, z) dz, & D \neq 0, \end{cases}$$



$$\frac{\partial S(x, t)}{\partial t} = D \frac{\partial^2 S(x, t)}{\partial x^2} - g \frac{\partial S(x, t)}{\partial x} + k_1 G_1(x, t) - \bar{S}(x, t; \tau_s = T_s), \quad t > 0, 0 < x < L,$$

$$S(x, t = 0) = S_0, \quad 0 < x < L,$$

$$D \frac{\partial \bar{S}}{\partial x}(x = 0, t) - g \bar{S}(x = 0, t) = 0, \quad t > 0,$$

$$S(x, t) = k_1 G_1(x, t) + \int_0^{T_s} \int_0^L k_1 G_1(x, t - \tau_s) \gamma(\tau_s, x, z) dz d\tau_s,$$

## 4. Further new maths

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- *Finite Differences+convolution: Do get SSD's but slow.*
- Look for separable solutions.....

$$G_1(x, t) = N(t)y_1(x)$$

$$S(x, t) = N(t)y_s(x)$$

$$G_2(x, t) = N(t)y_2(x)$$

$$M(x, t) = N(t)y_M(x)$$

If solutions are attracting then  
the  $y$ 's are the SSDs in each phase.

# Equations for $y_1$ the SSD: G1-phase

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Delay equation equation:

$$y_1(x-1) = \Lambda y_1\left(\frac{x}{2}\right), \quad x > 0, \quad D=0,$$

Solution:  $y_1(x) = \delta(x-1)$ ,  $\Lambda = 1/2$ . **Check.**

Exam Question on generalised functions?

Fredholm integral equation (non-symmetric):

$$\int_0^L \gamma(T_s, 2x, z) y_1(z) dz = \Lambda y_1(x), \quad x > 0, \quad D \neq 0$$
$$\Lambda = F(\lambda) = \frac{(\lambda + k_1)(\lambda + k_2)(\lambda + b)e^{\lambda T_s}}{4bk_1k_2}$$

# SDD Solutions

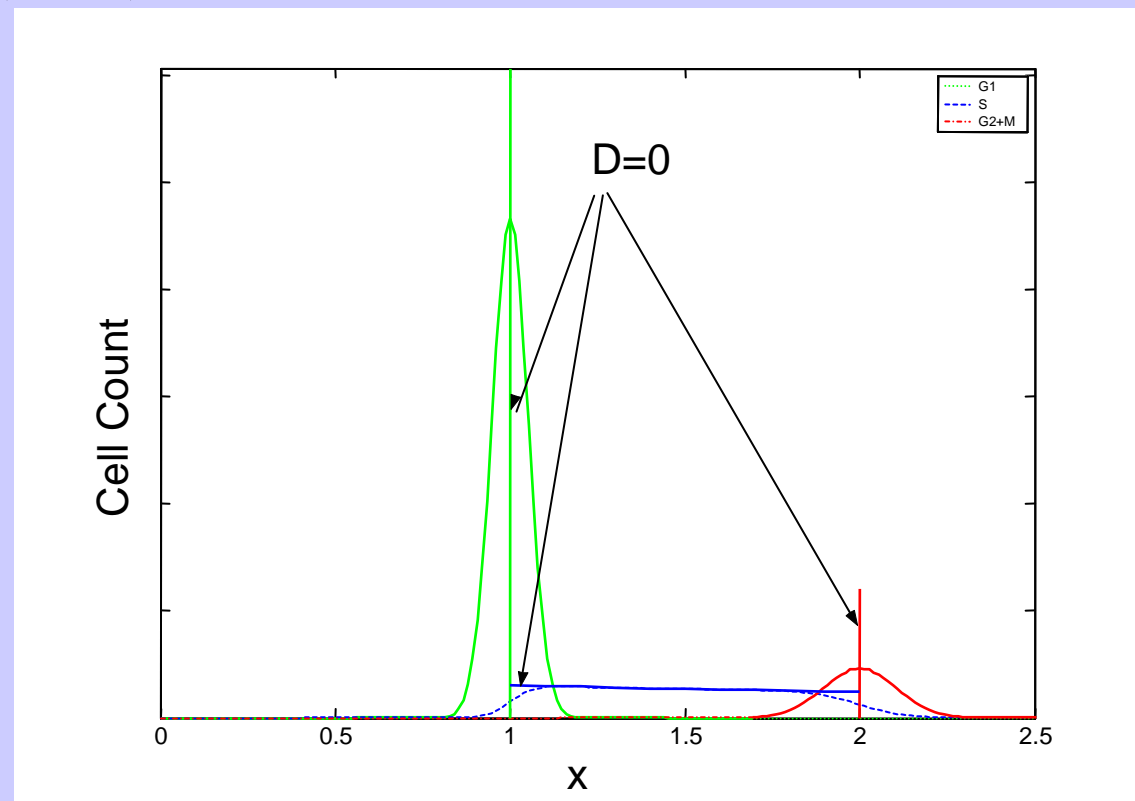
- *Delay equation ( $D=0$ ): (unique) Point distributions.*

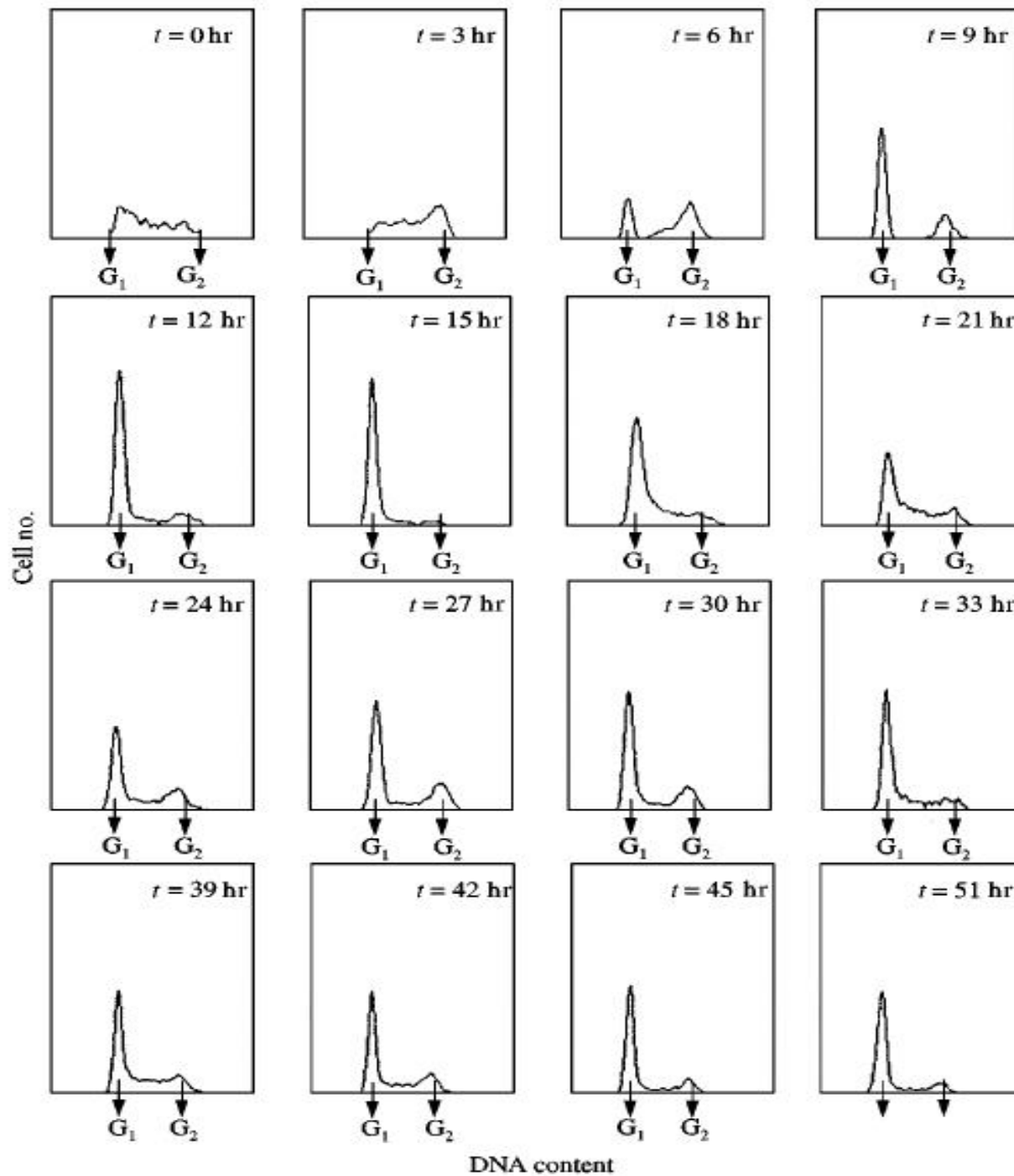
$$y_1(x) \sim \delta(x-1), \quad y_2(x), y_M(x) \sim \delta(x-2),$$

$$y_S(x) \sim H(x-2) - H(x-1)$$

- *Fredholm Integral Equation ( $D \neq 0$ ):*

*Numerical methods.  
Get 1 eigenfunction  
(there could be  
others)*





Chiorino et al.  
*Desynchronization Rate in Cell Populations: Mathematical Modeling and Experimental Data*, J. theor. Biol. (2001) **208**, 185-199

# References: Cancer treatment

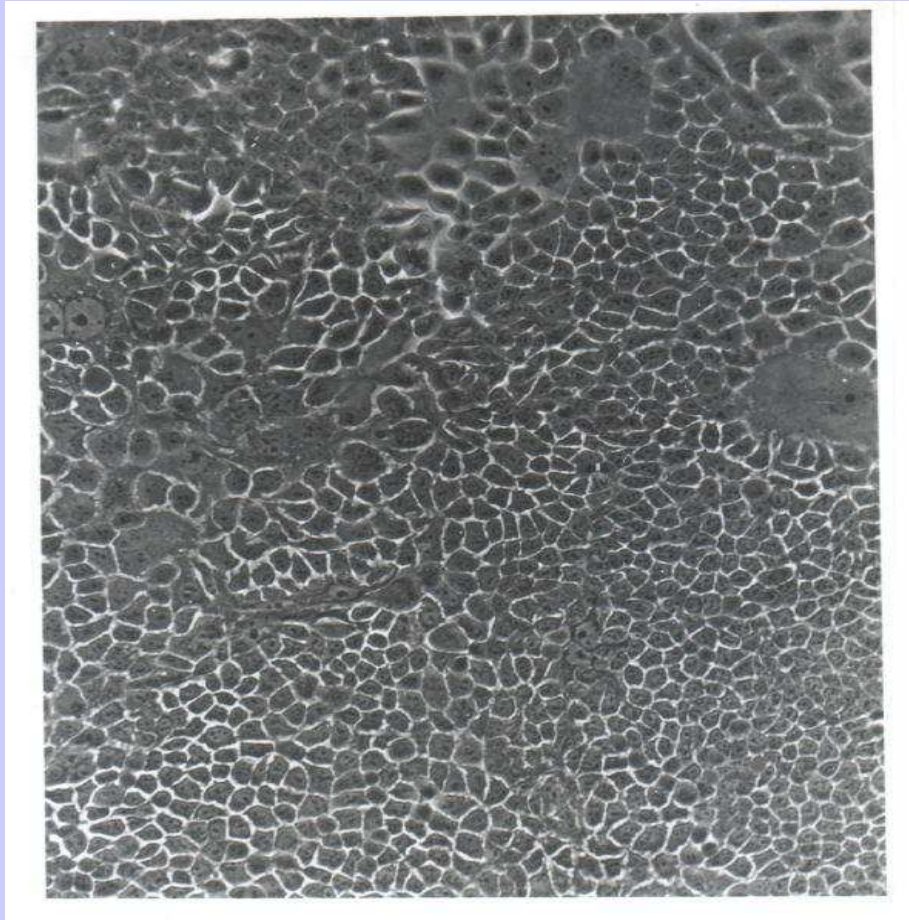
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- Basse B, Baguley BC, Marshall WR, Joseph B, van-Brunt B, **Wake GC**, & Wall DJN "Modelling cell death in human tumour cell lines exposed to the anticancer drug paclitaxel", Journal Mathematical Biology, **49**, 2004, 329-357.
- Basse B, Baguley BC, Marshall E, **Wake GC** & Wall DJN) "Modelling cell population growth with applications to cancer therapy in human cell lines", Prog Biophysics Mol Biol, **85**, 2004, pp 353-368.



## 5. Modelling Cancer Treatment: Cell Lines

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# Medical options

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- Cure by
  - \* poison = chemotherapy
  - \* burn = radiotherapy
  - \* cut = surgery

# Taxol effect: 6 weeks “in vivo”

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Model of a cell line  
perturbed by paclitaxel

- Halts cell division
- Can induce cell death in G2/M

$$\frac{\partial G_1(x,t)}{\partial t} = -k_1 G_1(x,t), \quad t > 0, 0 < x < L,$$
$$G_1(x, t=0) = G_{10}, \quad 0 < x < L,$$

$$\frac{\partial M(x,t;\tau_M)}{\partial t} = -\mu_M M(x,t;\tau_M), \quad t > 0, 0 < x < L,$$
$$M(x,t;\tau_M=0) = k_2 G_2(x,t), \quad t > 0, \quad 0 < x < L,$$
$$M(x,t=0;\tau_M) = M_0, \quad \tau_M > 0, \quad 0 < x < L,$$

Model of a cell line  
perturbed by paclitaxel cont...

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$$\frac{\partial A(x,t)}{\partial t} = \frac{g_A \partial A(x,t)}{\partial x} + \int_0^\infty \mu_M M(x,t;\tau_M) d\tau_M, \quad t > 0, 0 < x < L,$$

$$A(x=0,t) = A(x=L,t) = 0, \quad t > 0,$$

$$A(x,t=0) = A_0, \quad 0 < x < L,$$

# Parameter Fitting

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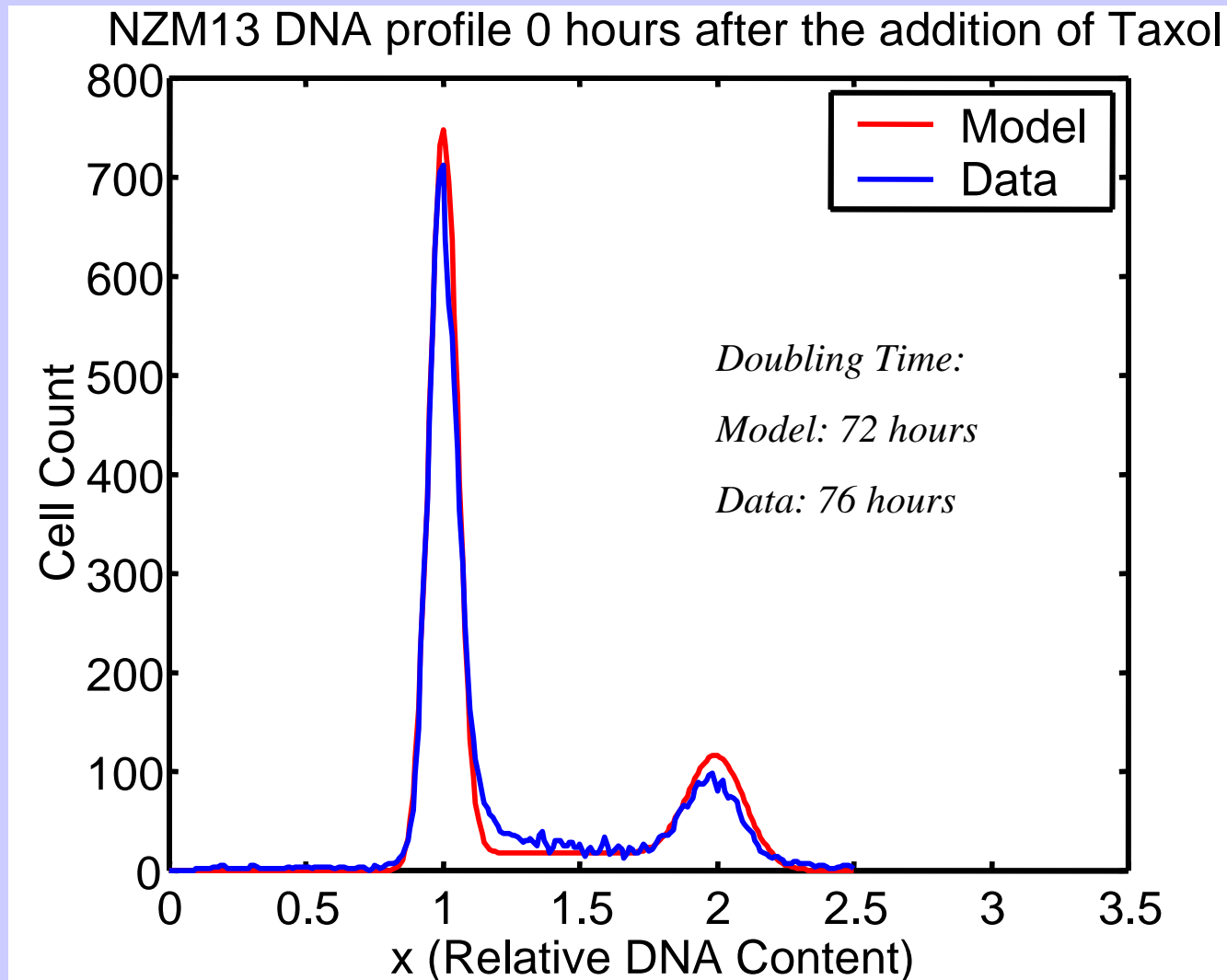
- Minimise  $\sum_{j=1}^J \left( T(\mathbf{x}, t_j) - D(\mathbf{x}, t_j) \right)^2$   
 $\beta = (k_1, k_2, T_M, \mu_M, g_A)$ 
  - *Choose a parameter set*
  - *Find the model of an unperturbed cell line SDD*
  - *Use finite differences and convolution to solve the model of a cell line perturbed by taxol*
  - *Calculate the objective function value*



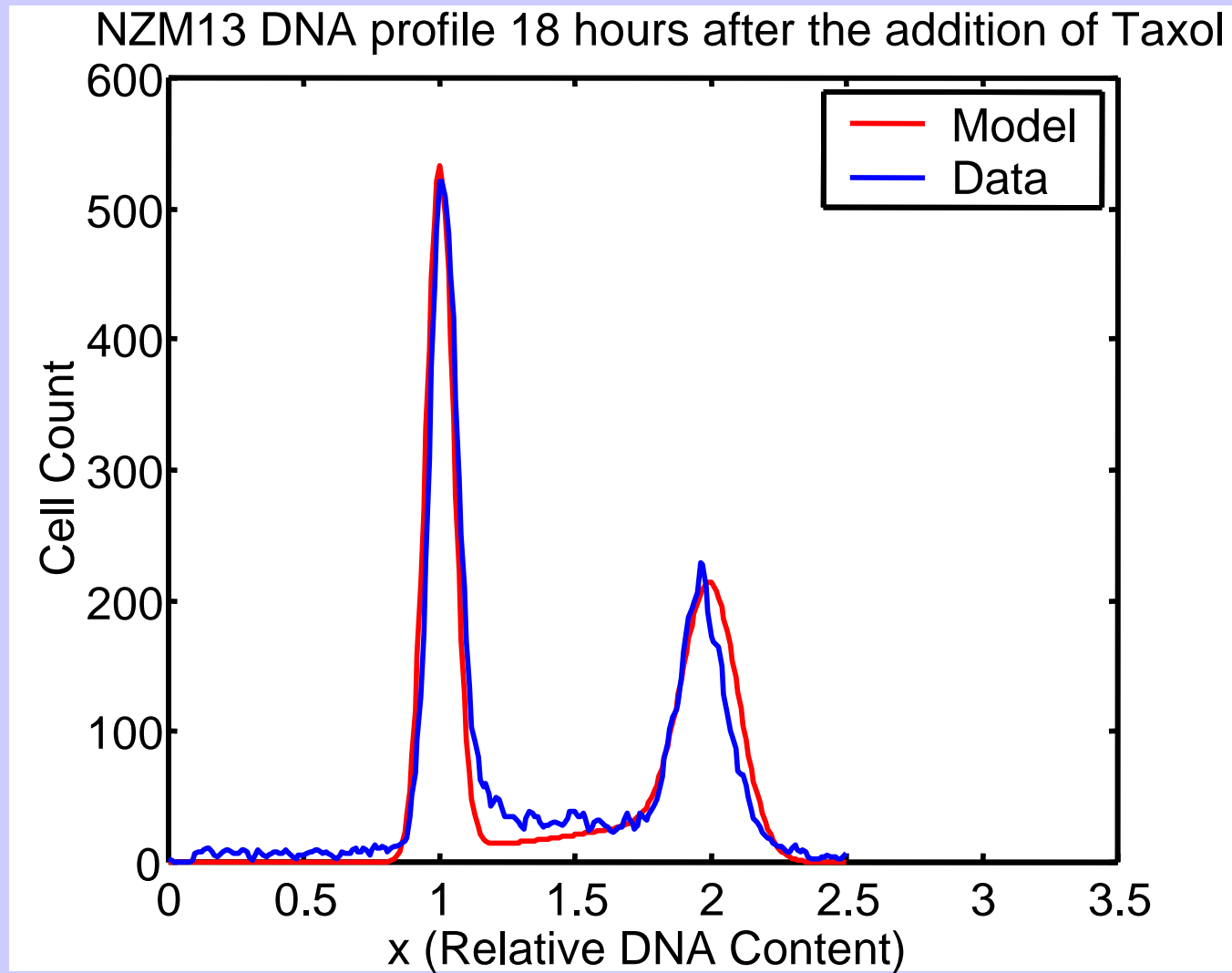
# References: Cell-growth; Compartments

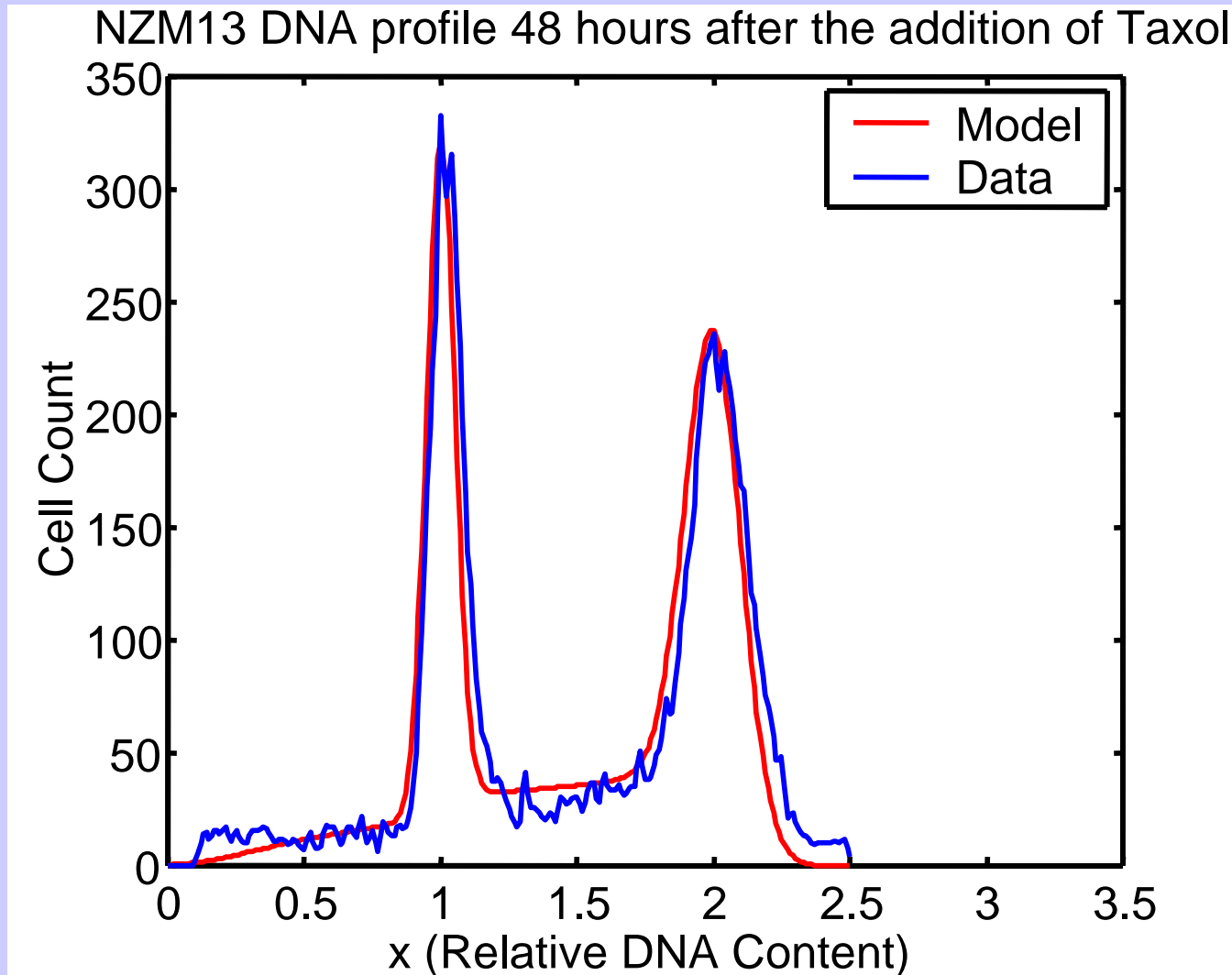
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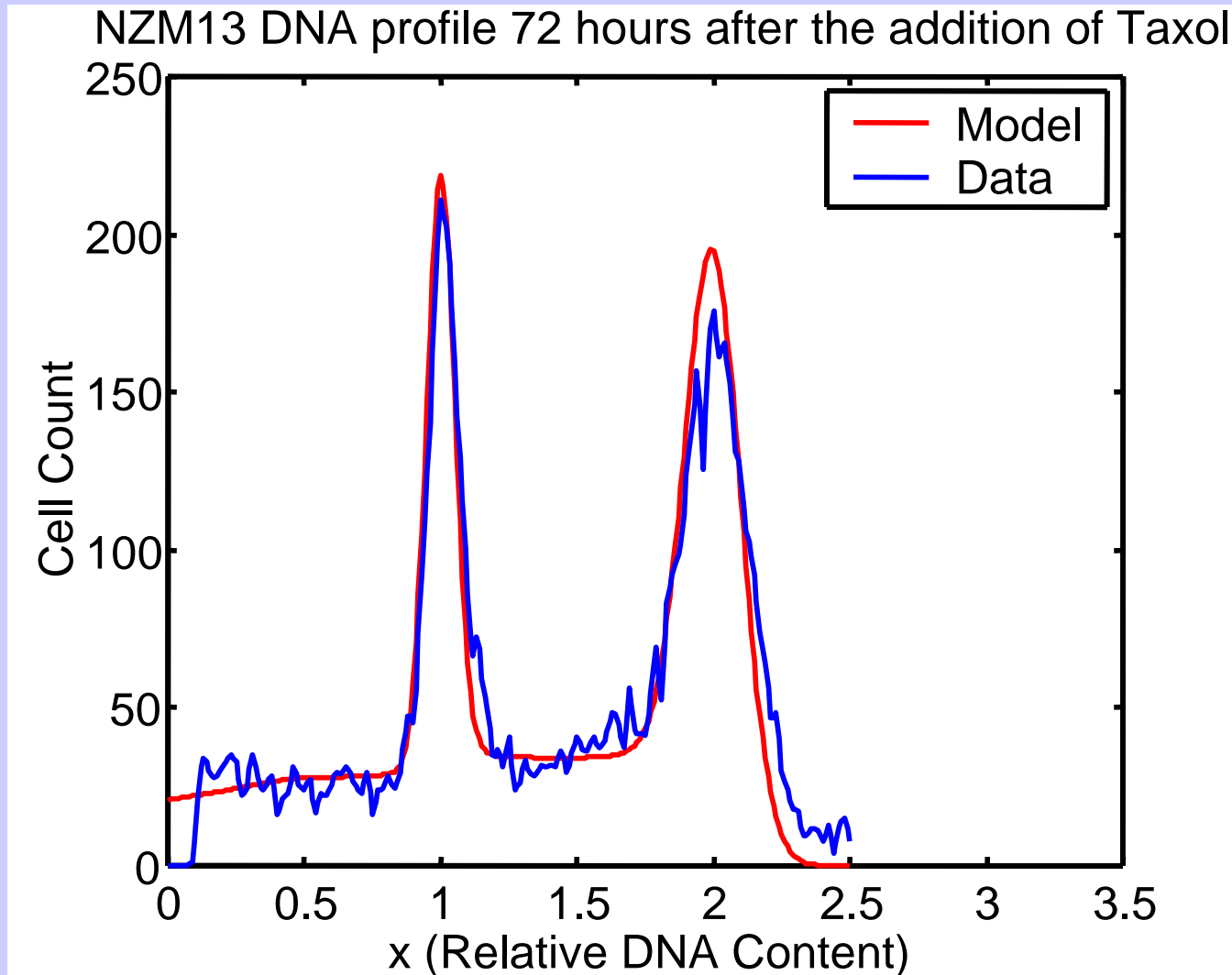
- Basse B, Baguley BC, Marshall ES, Joseph WR, van Brunt B **Wake GC** & Wall DJN. “A mathematical model for analysis of the cell cycle in cell lines derived from human tumours”, J Math Biol 2003, **47**, pp 295-312.
- Basse B, **Wake GC**, Wall DJN, & van-Brunt B. “On a cell-growth model for plankton” Mathematical Medicine and Biology: A Journal of the IMA; **21**, 2004, pp 49-61.

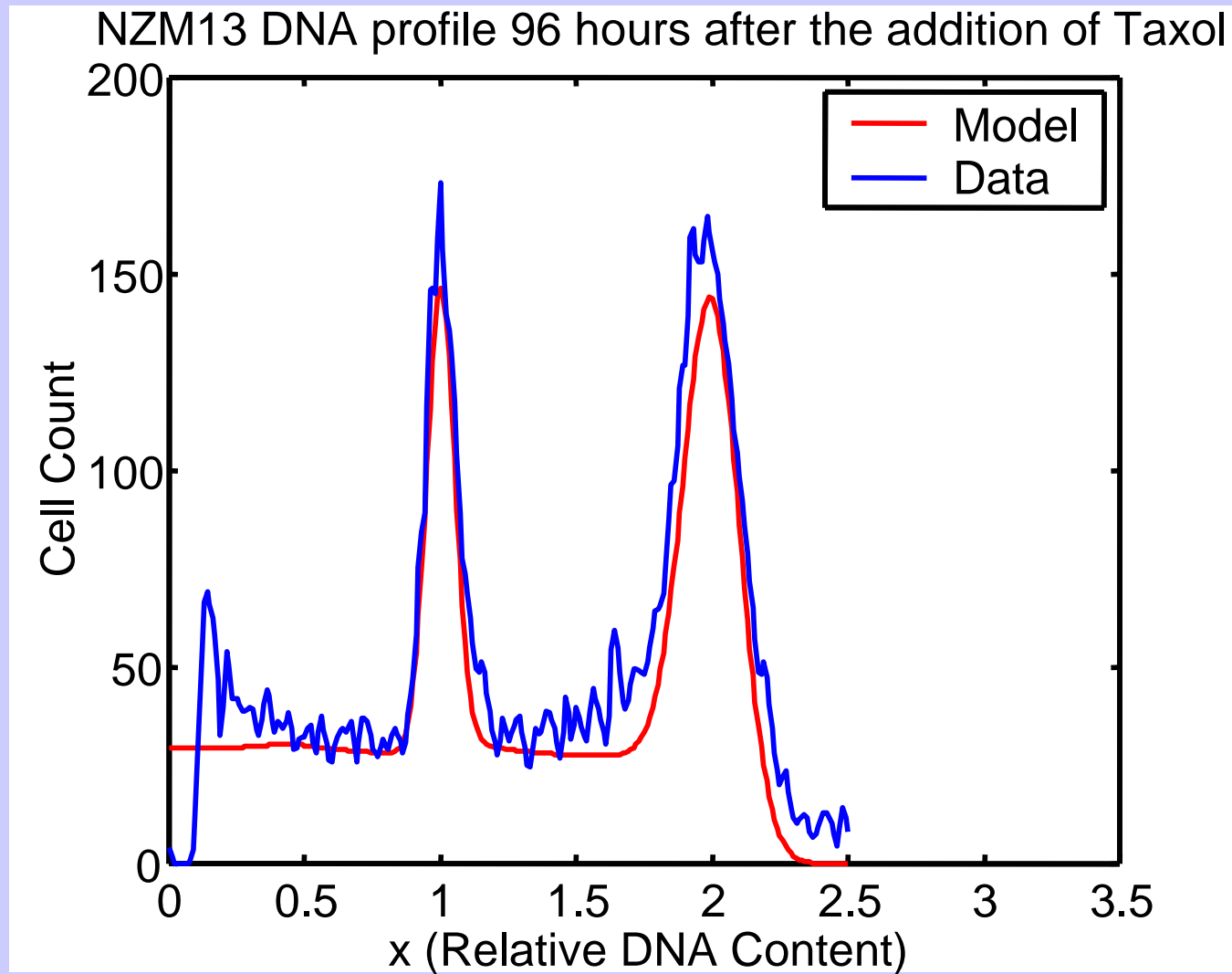










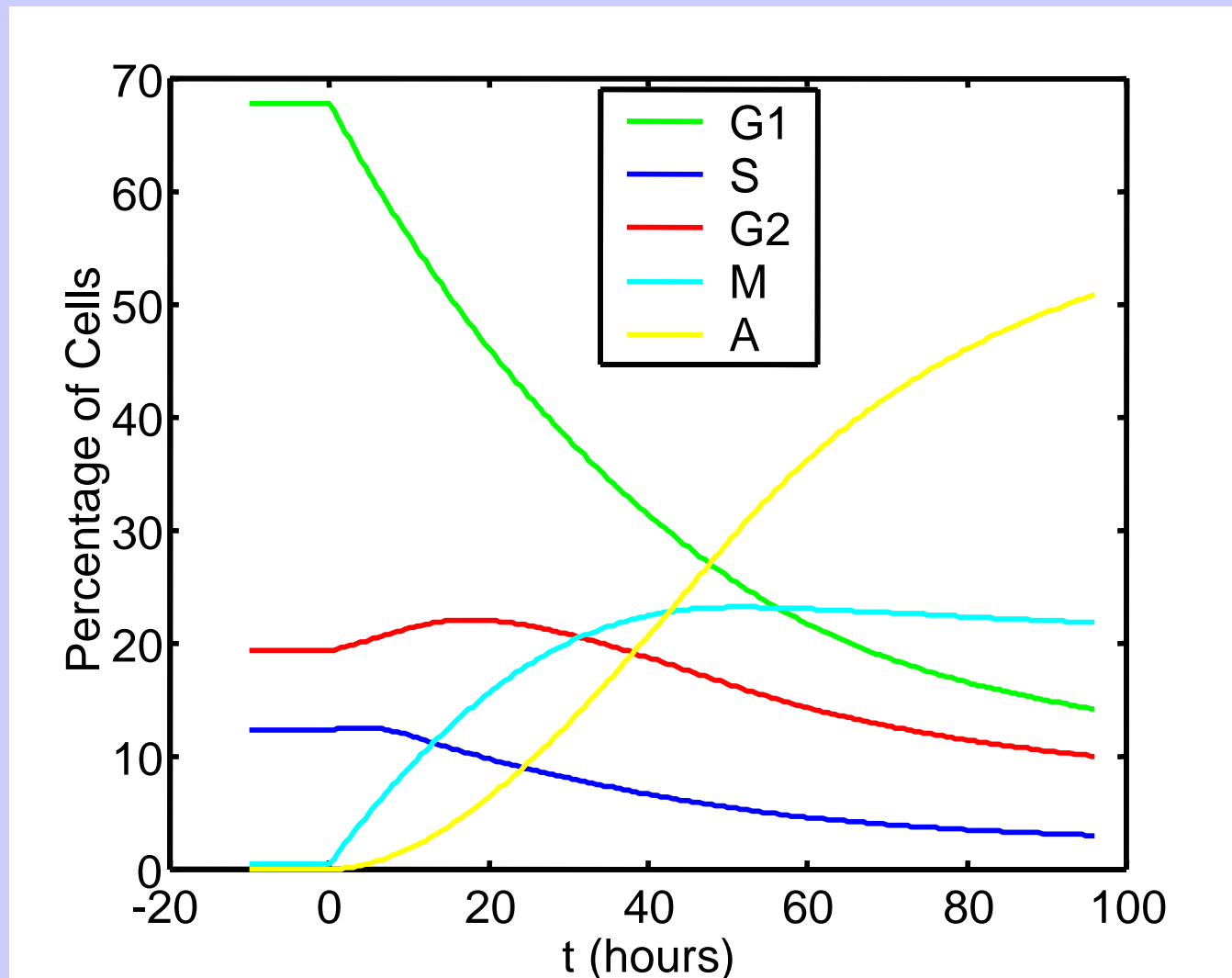


# Model Outputs

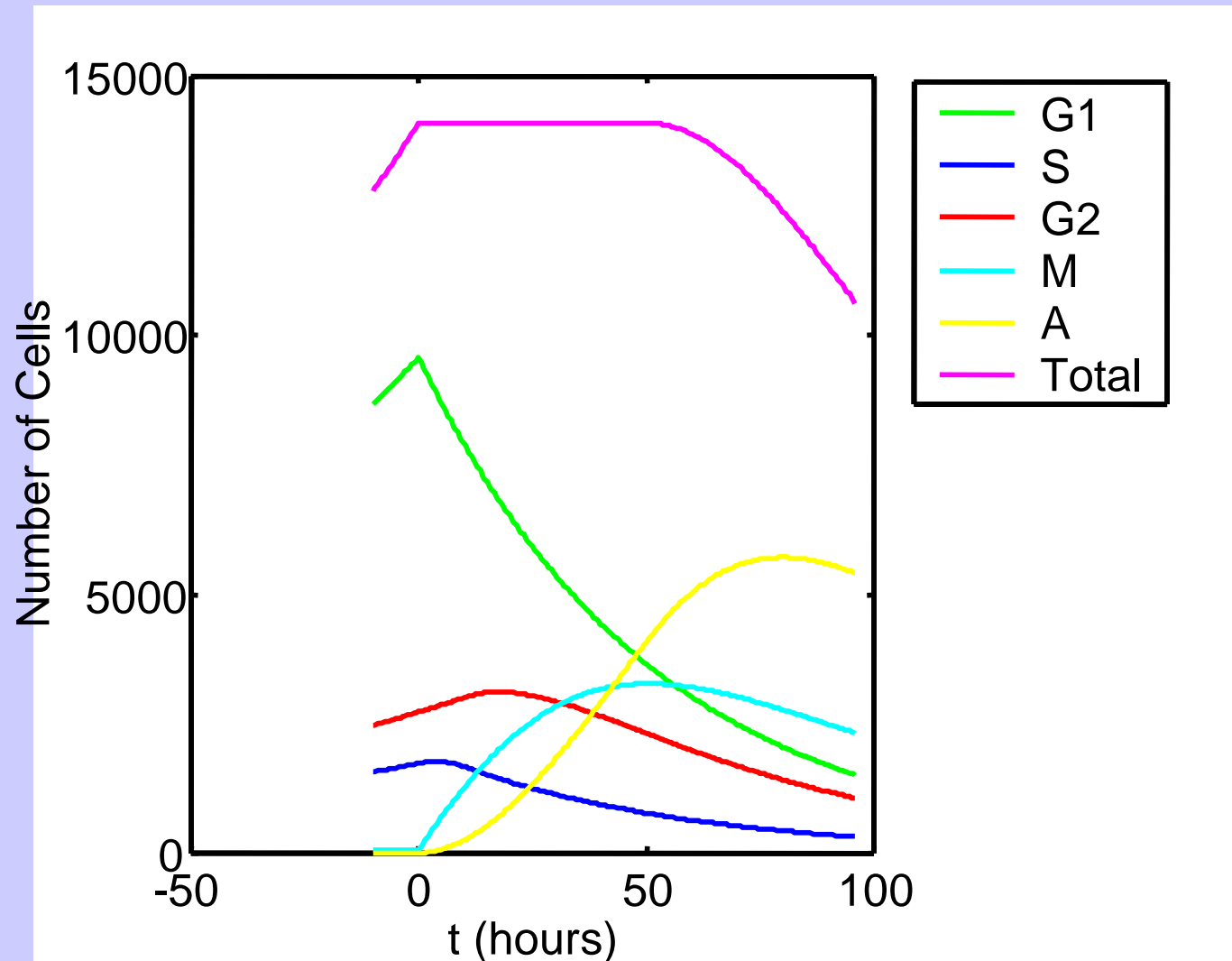
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- DNA profiles
- Percentages in each phase over time
- Absolute numbers in each phase over time
- $T_M$ , the time in M-phase before the onset of apoptosis
- $\mu_M$ , the eventual transition rate from M-phase to A-phase
- $g_A$ , the degradation rate in A-phase
- the time it takes for a cell to degrade
- the rate of eventual cell loss from A-phase

# Percentages in each phase

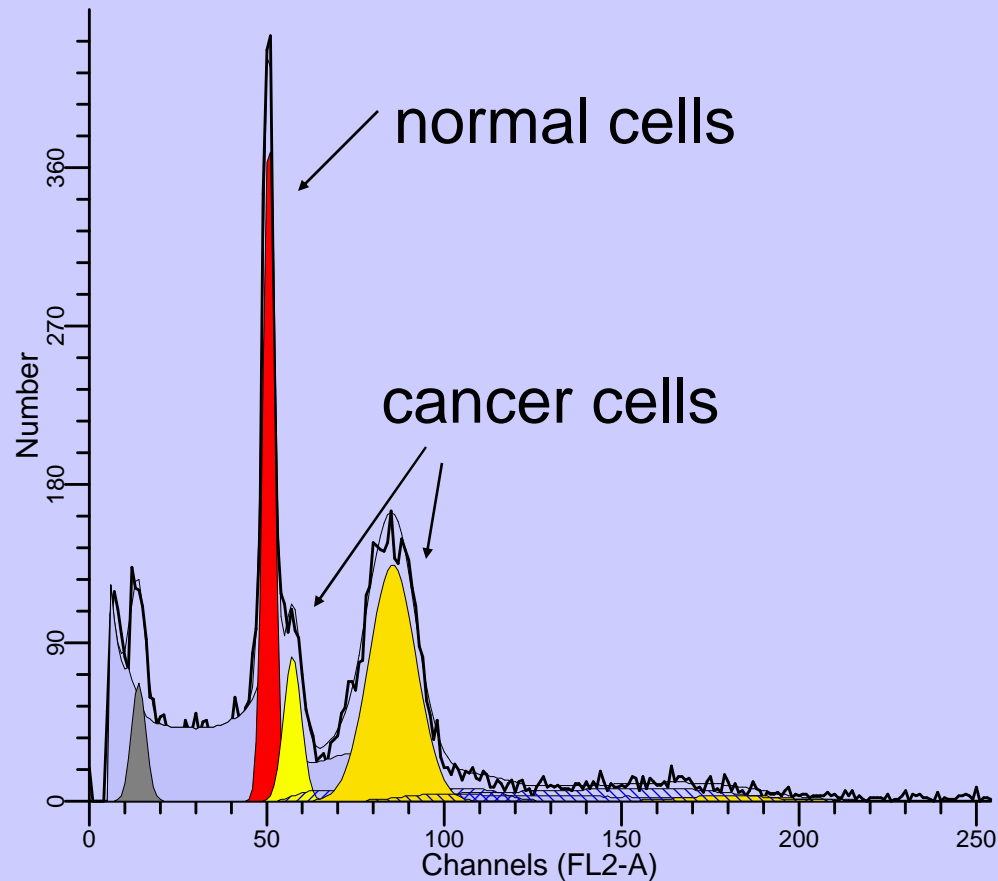


# Absolute Cell Number



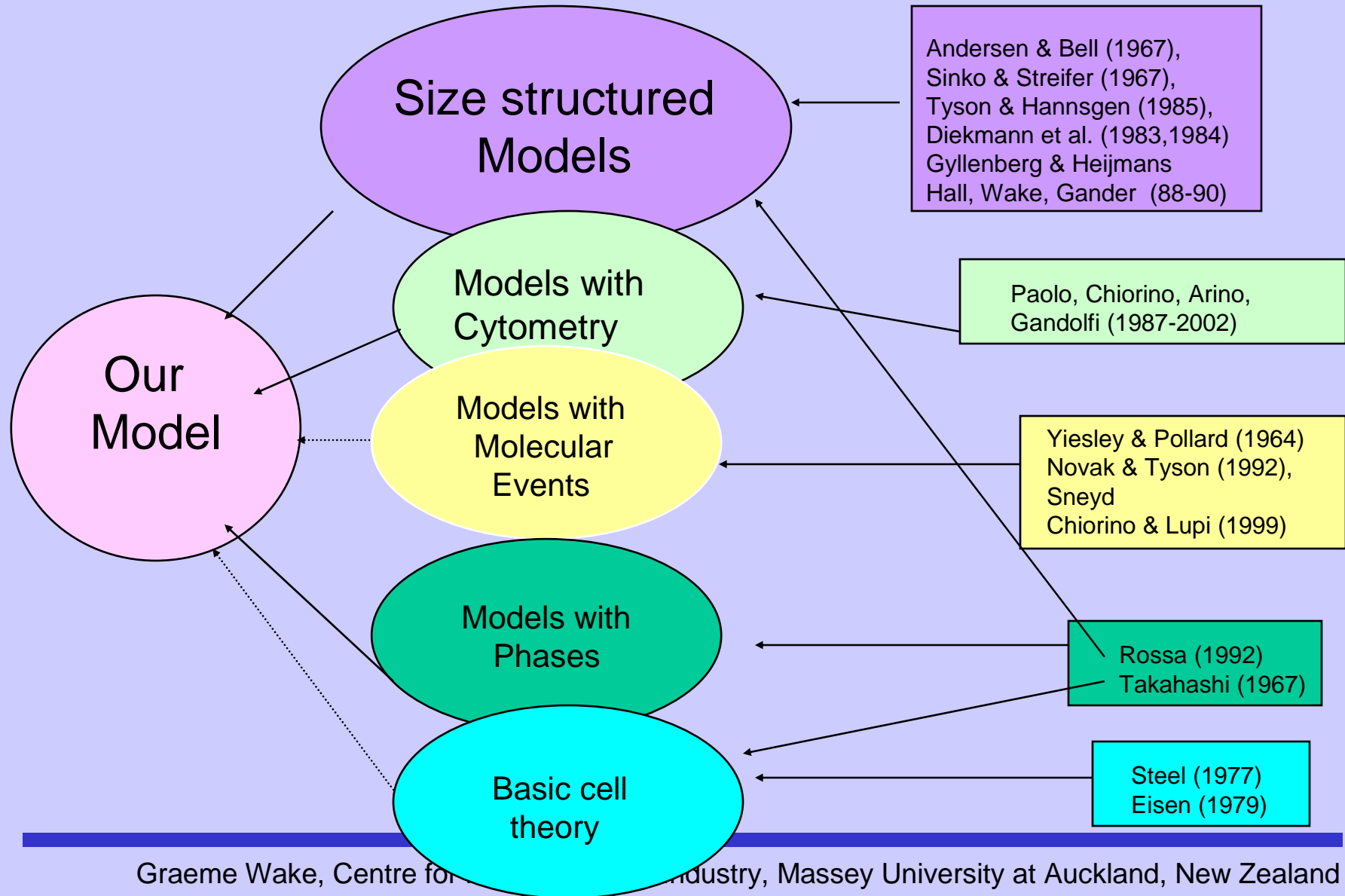
# Typical Clinical Sample

Patient with metastatic malignant melanoma





# 6. Other Models



# 7(a) Current work -Basse

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- *Uniqueness of parameter fitting*
- *Fit other data, compare results*
- *Non-constant degradation rate*
- *Change the concentration of paclitaxel in cell lines*
- *Effects of paclitaxel in vitro (mice) (doesn't stay in the system)*
- *Look at patient tumours*
- *See: Basse B, Baguley BC, Marshall ES, Wake GC & Wall DJN "Modelling the flow of cytometric data obtained from unperturbed human tumour cell lines: Parameter fitting and comparison", Bulletin of Math Biology, **67**, No 4, 2005, 815 - 830.*

## 7(b) Current work: Transient Models

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- Does the transient problem have a solution that (globally) is attracted to SSD behaviour?
- Yes!!!!!!

# The model; Single compartment only

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$$n_t = (Dn)_{xx} - (gn)_x - \mu n + \alpha^2 B(\alpha x)n(\alpha x, t) - B(x)n(x, t)$$

$$(D(0)n(0, t))_x - g(0)n(0, t) = 0, \quad t > 0$$

$$n(x, t) \rightarrow 0, \quad x \rightarrow \infty \qquad n_x(x, t) \rightarrow 0, \quad x \rightarrow \infty$$

- $n(x, t)$  = density of cells of 'size'  $x$  at time  $t$ .
- Size is measured by DNA content
- $B(x)$  is the rate of cell division at size  $x$ .  $B(x) = b\delta(x - l)$
- A cell of size  $x$  divides into  $\alpha$  daughter cells of size  $\frac{x}{\alpha}$
- Of primary interest in this presentation is fixed-size division:

# SSD behaviour

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- SSD = Steady Size-Distribution
- A model displays SSD behaviour when the shape of the cell-size distribution remains constant while the overall population grows or decays
- This corresponds to a separable solution of the model
- Growth rate is usually exponential
- SSDs are observed to occur in physical cell-cohorts

# The $D=0$ case: no dispersion

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Problem:  $n_t = -gn_x - \mu n + \alpha b \delta(x-l/\alpha)n(l^-,t) - b \delta(x-l)n(l^-,t),$   
 $n(x,0) = n_0(x), \quad x > 0,$   
 $n(0,t) = 0, \quad t > 0.$

- The limiting shape of the SSDs for the dispersive case as  $D \rightarrow 0$  with the requirement of continuity from the left, is a global attractor in this model (in a sense): basically it is the hull that exhibits SSD behaviour

$$n(x,t) e^{\lambda t} \sim \max_{t \geq 0} n(x,t) = y(x) \text{ as } D \rightarrow 0$$

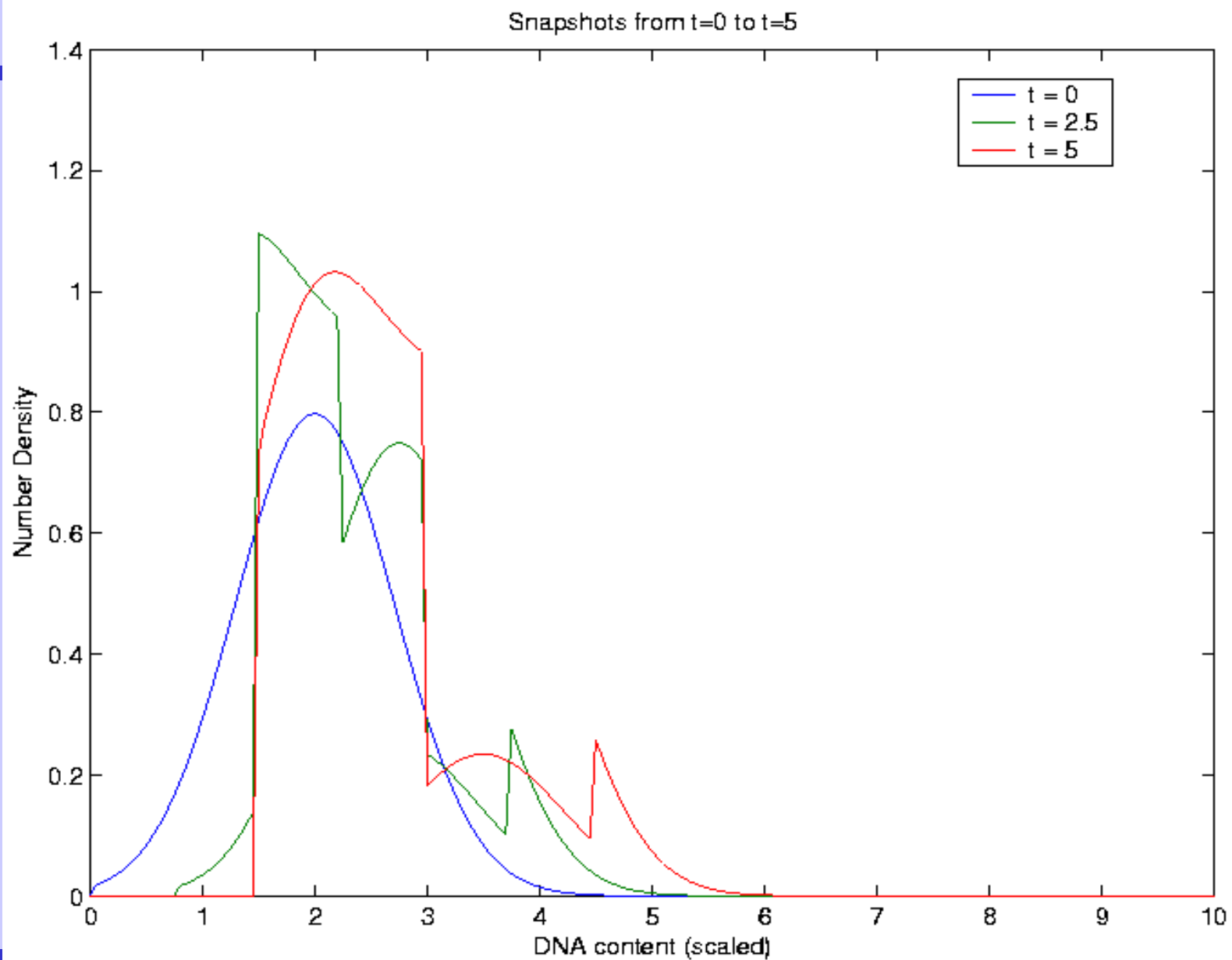
## The solution to the D=0 case: follow characteristics

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$$n_1(x, t) = n_0(x - gt)e^{-\mu t} H(x - gt), \quad t > 0. \quad (\alpha = 2)$$

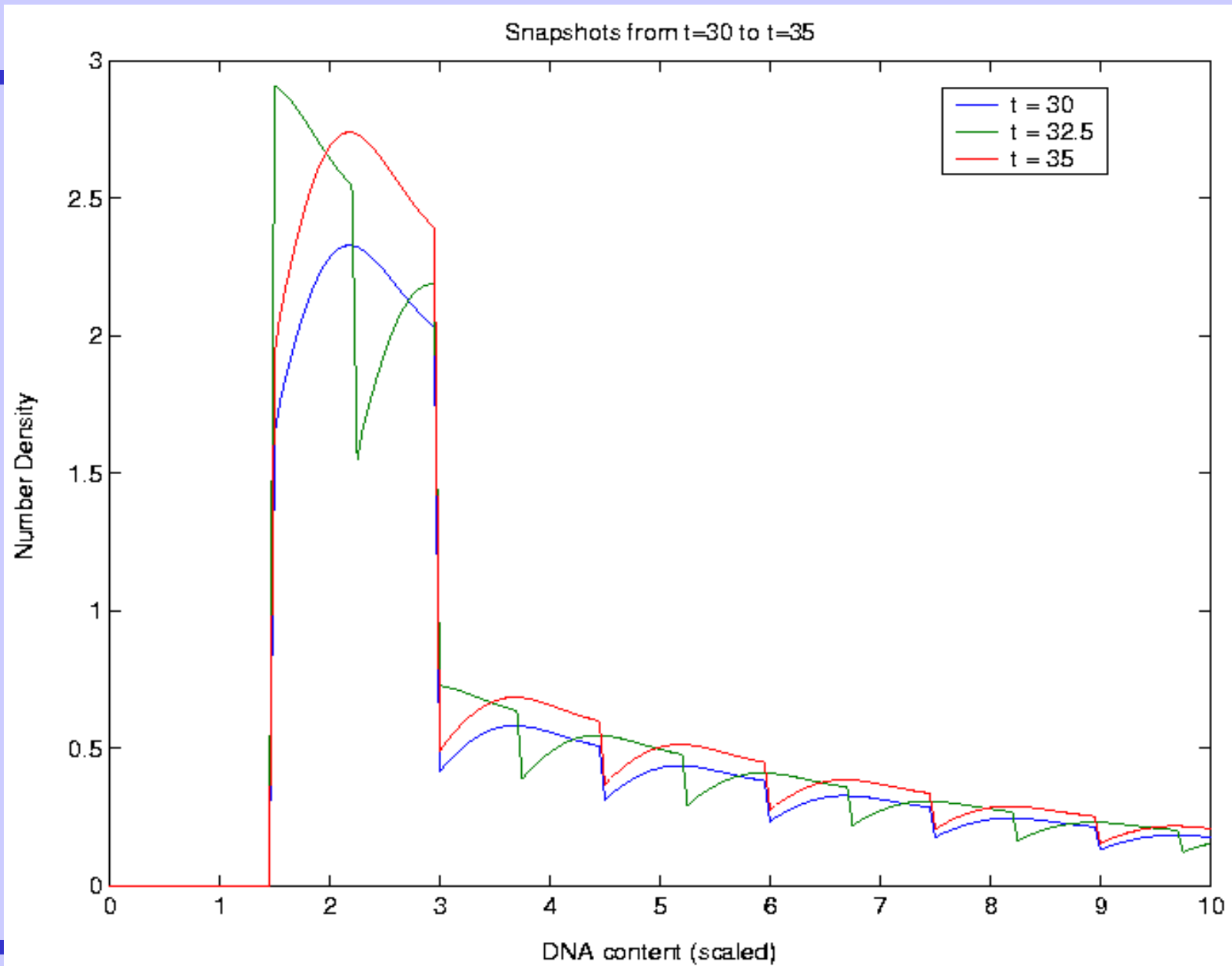
$$n_2(x, t) = \begin{cases} n_0(x - gt)e^{-\mu t}, & \frac{l}{2} < x - gt < l, \\ e^{-\mu t} \left[ \lambda^m n_0\left(x - gt + \frac{ml}{2}\right) + \lambda^{m+1} n_0\left(x - gt + \frac{(m+1)l}{2}\right) \right], & \frac{-ml}{2} < x - gt < \frac{-(m-1)l}{2} \end{cases}$$

$$n_3(x, t) = \begin{cases} n_0(x - gt)e^{-\mu t}, & l < x - gt, \\ \left(1 - \frac{b}{g}\right) n_2(x, t), & x - gt < l. \end{cases} \quad \left( \lambda = \frac{\alpha b}{g} \right)$$



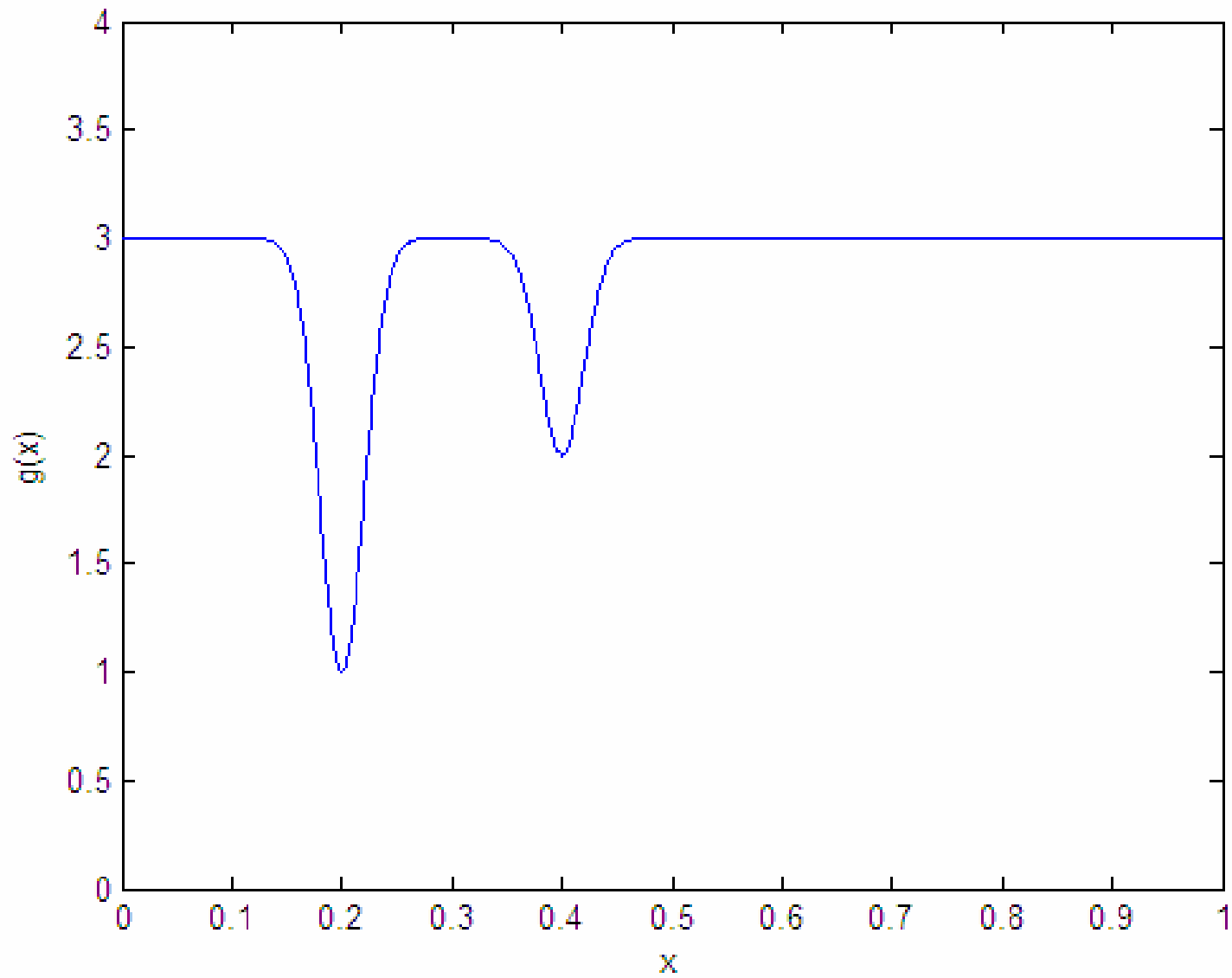
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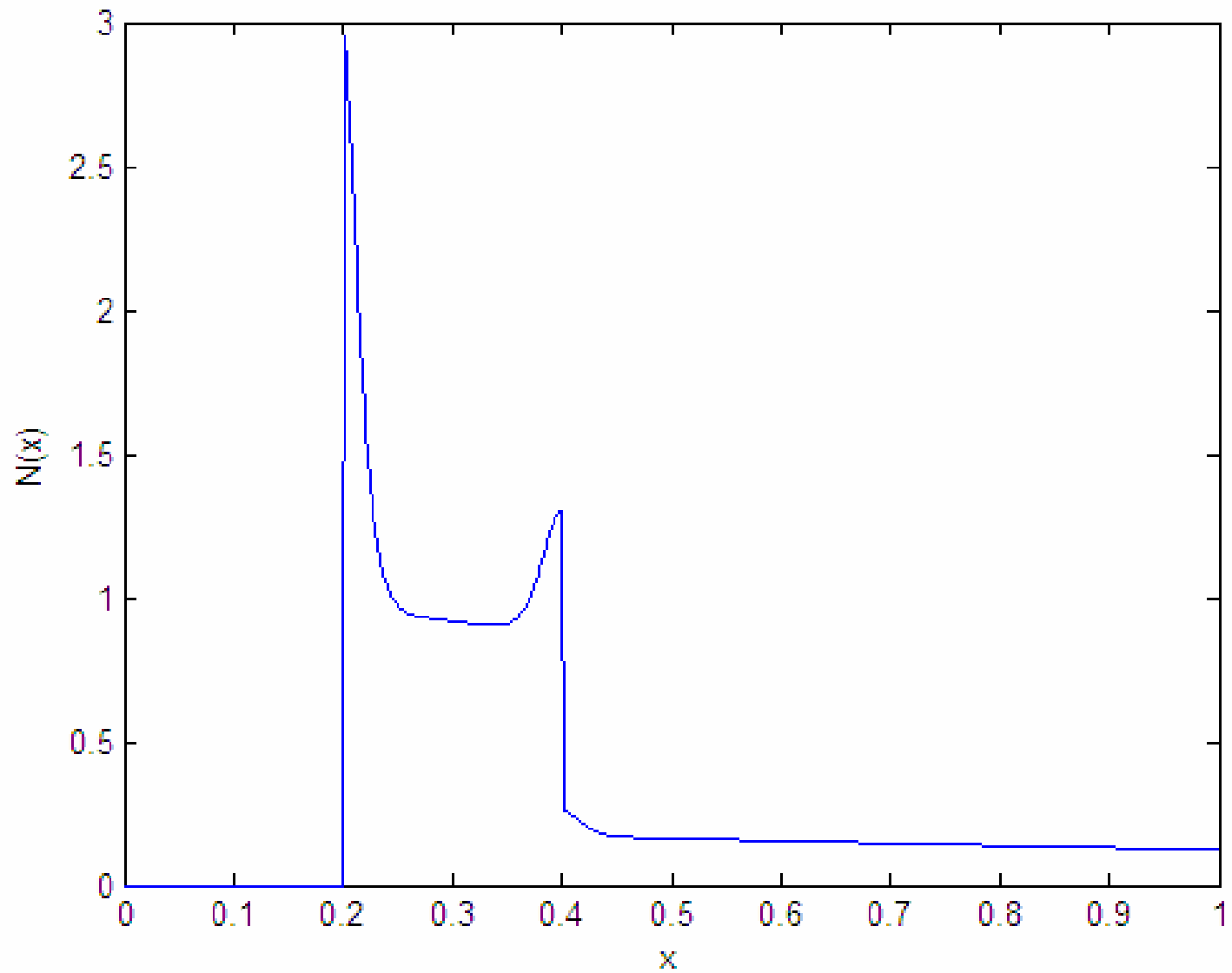


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## A variable growth function



## The corresponding hull



# Summary

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- With  $D=0$ , fixed size division the solution exhibited periodic exponential growth
- The hull is a global attractor with the same shape as the limiting SSD as  $D$  tends to zero
- A variable growth rate allows the shape of the hull to match observations
- Upper/lower solutions were used to prove a convergence result in a special case (possibly useful in other cases?)

# References

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R. Begg, G. C. Wake, D. J. N. Wall, *On a functional equation model of transient cell growth*, Math. Med. Biol. **22** (4) (2005) 371-390 (D= 0).

R. Begg, G. C. Wake, D. J. N. Wall, *On the stability of steady size-distributions for a stochastic cell-growth process*, J. Differential and Integral Equations: Vol **21**, Nos. **1-2**, 2008, pp 1-24.(D >0)

# The transient problem $D > 0$ , fixed size division

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- The problem is:

$$\frac{\partial}{\partial t}n(x,t) = D\frac{\partial^2}{\partial x^2}n(x,t) - g\frac{\partial}{\partial x}n(x,t) + \alpha^2 b\delta(\alpha x - l)n(\alpha x, t) - b\delta(x - l)n(x, t) - \mu n(x, t),$$

$$n(x, 0) = n_0(x), \quad n_0 \in (C \cap L^1 \cap L^\infty)[0, \infty)$$

$$Dn_x(x, t) - gn(x, t)|_{x=0} = 0,$$

$$n(x, t) \rightarrow 0, \quad x \rightarrow \infty, \quad t > 0$$

$$n_x(x, t) \rightarrow 0, \quad x \rightarrow \infty, \quad t > 0.$$

# SSD equation

- This is found to be

equation,

$$\begin{cases} y''(x) - \gamma y'(x) + \alpha^2 \beta \delta(\alpha x - l) y(\alpha x) - (\beta \delta(x - l) + \lambda) y(x) = 0, \\ y \in (C \cap W^{2,1} \cap L^\infty)[0, \infty), \\ y'(0) - \gamma y(0) = 0, \\ y'(x), y(x) \rightarrow 0, \quad x \rightarrow \infty. \end{cases} \quad (3.1.7)$$

where  $W^{2,1}[0, \infty)$  is the Sobolev space of functions in  $L^1[0, \infty)$  who have weak derivatives up to order 2 also in  $L^1[0, \infty)$  (where we consider the  $\delta$ -distribution to be in  $L^1[0, \infty)$  for now);  $\gamma = g/D$ ,  $\beta = b/D$  and  $\lambda$  is an eigenvalue of the operator

$$y(\cdot) \mapsto y''(\cdot) - \gamma y'(\cdot) + \alpha^2 \beta \delta(\alpha \cdot - l) y(\alpha \cdot) - \beta \delta(\cdot - l) y(\cdot).$$

If such an eigenvalue exists then there is a separable solution,  $N(t)y(x)$

- Trick is to use the dual problem (Ronald Begg found this), where  $m(x,t) = n(x,t)e^{\lambda t}$

$$\begin{cases} \psi''(x) + \gamma\psi'(x) + \alpha\beta\delta(x-l)\psi\left(\frac{x}{\alpha}\right) - (\beta\delta(x-l) + \lambda)\psi(x) = 0 \\ \psi'(0) = 0, \quad 0 < \psi(x) \in (C \cap W^{2,1} \cap L^\infty)[0, \infty), \quad \int_0^\infty \psi(x)y(x) dx = 1, \end{cases}$$

has two very useful properties which help in proving the stability of the SSD  $y$ .

*The following convergence result holds:*

$$\int_0^\infty \psi(x)|m(x,t) - ky(x)| dx \rightarrow 0, \quad t \rightarrow \infty.$$



## Wish list....

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- Extend to splitting at any size.....
  - although the Basse et.al multi-compartment model collapsed to a single compartment seems to be best match by this:  
“splitting at any size”.

Some preliminary results are in:

Begg R, Wake GC & Wall DJN “The steady-states of a multi-compartment, age-size distribution model of cell-growth”; *Euro J of Appl Maths*, Accepted February 2008.

# Summary...

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- We have a generic set of simple (?)models for cellgrowth/division
- It can be, and is being used to underpin decision support
- There is plenty of “new mathematics” here

# Questions/Comments

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- Thank you for your attention!!!!

