

# How modelling can help understanding environmental triggers of disease

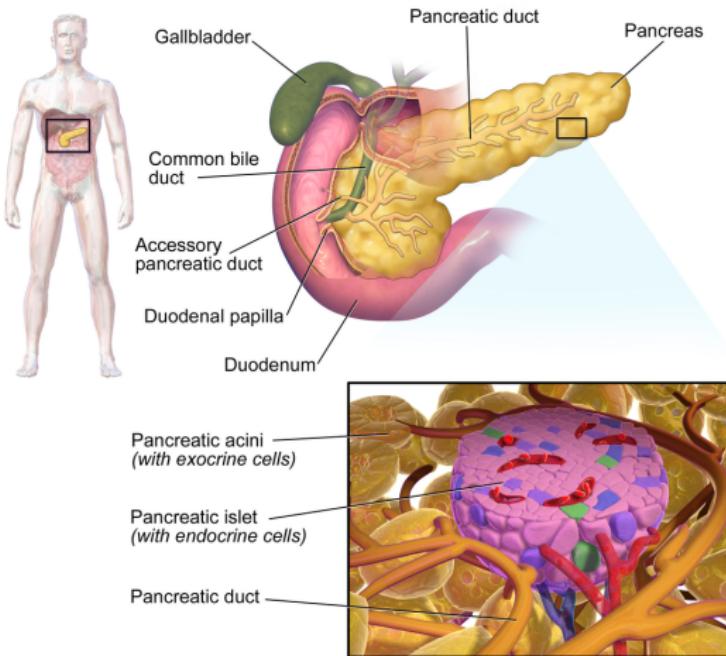
Stan Marée

Computational and Systems Biology

John Innes Centre

Norwich, January 26, 2017

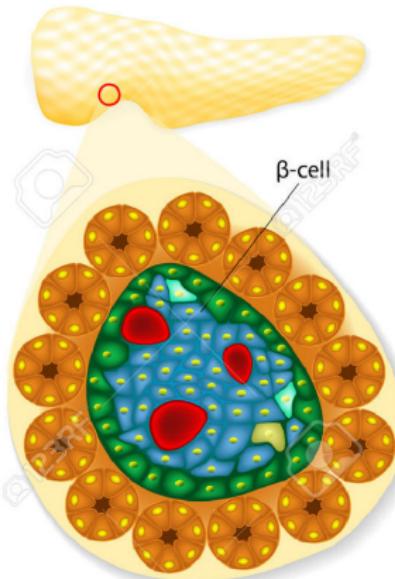
# Type I diabetes



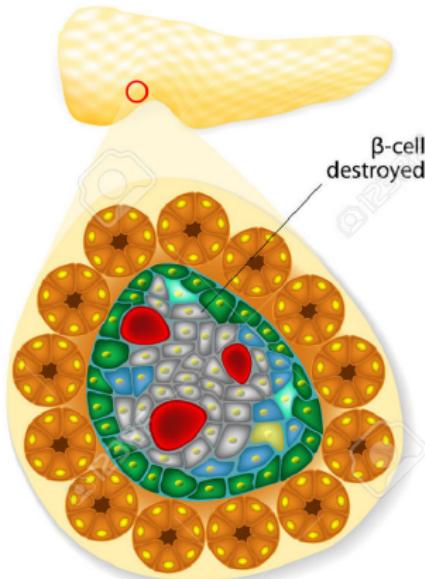
## Pancreatic Tissue

## ISLETS OF LANGERHANS

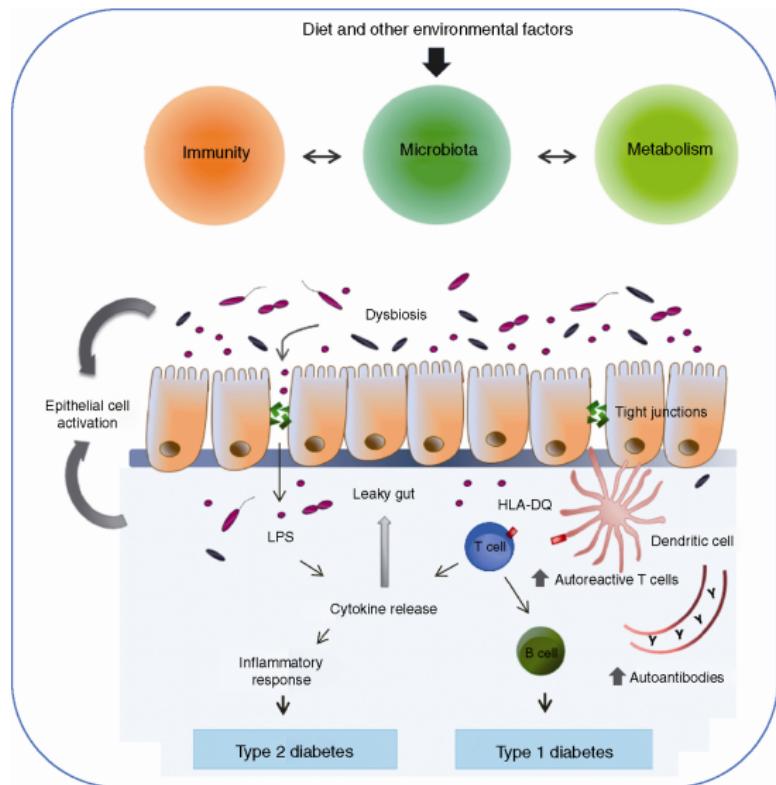
Healthy pancreas



Diabetes mellitus type 1



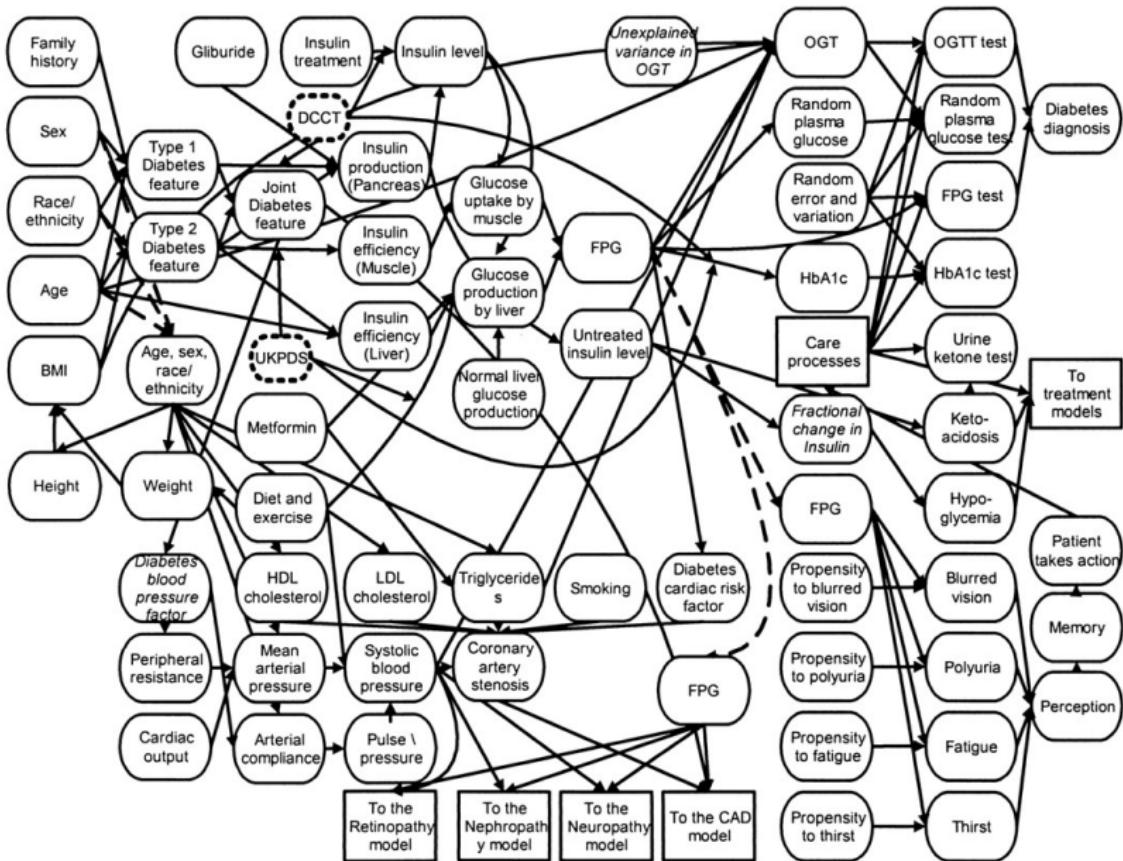
# Innate immunity and intestinal microbiota in the development of Type 1 diabetes

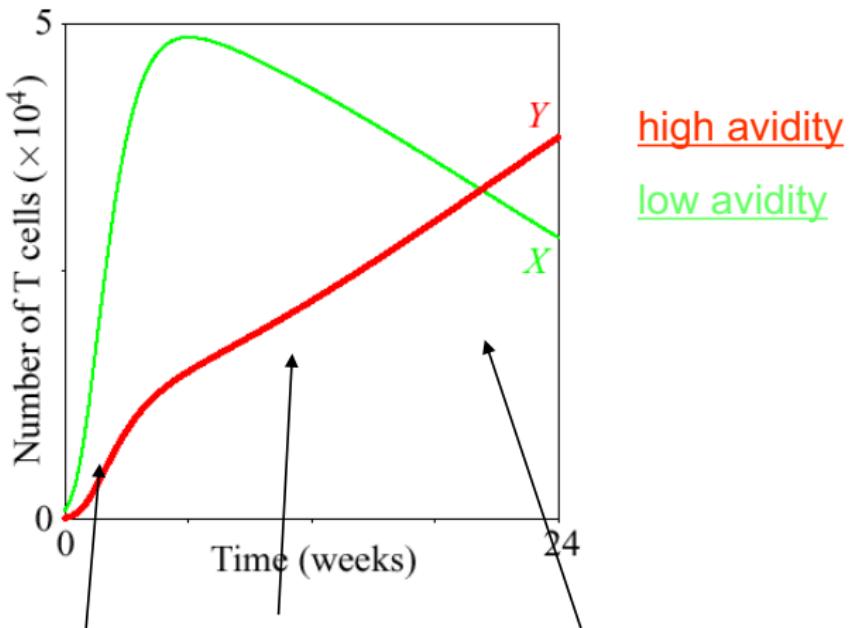


# NOD mouse model



# Trump modelling





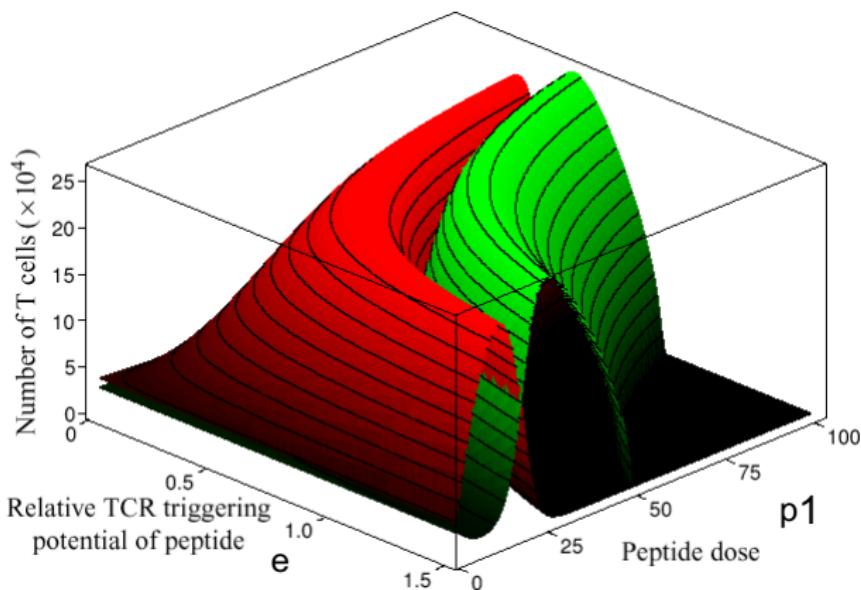
Weak competition

memory pool turnover

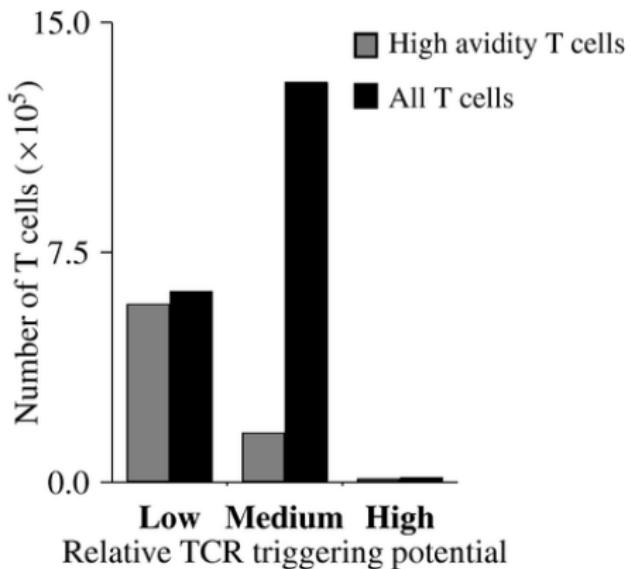
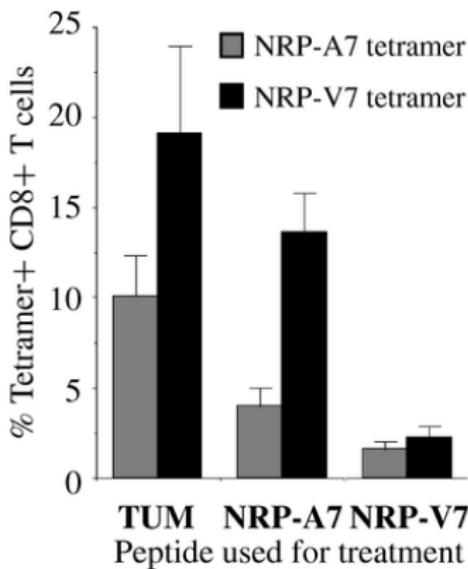
strong competition for sites on APC's;

# Gaining insights

Relative size of high and low avidity populations depends on both peptide dose  $p$  and affinity (triggering potential), e. Small changes in either can be very damaging.



# Gaining insights

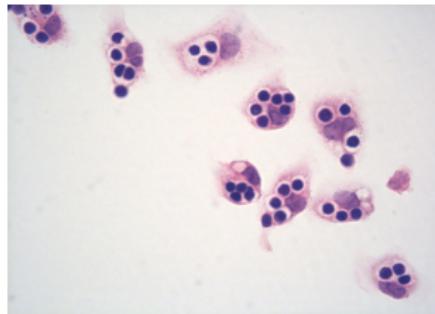
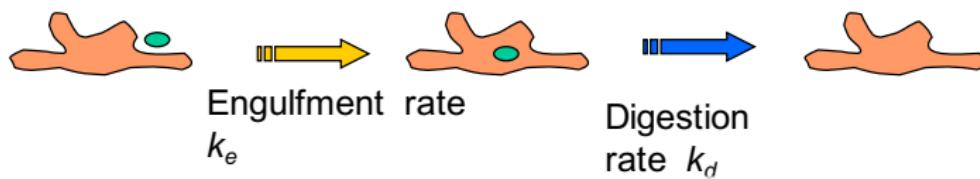


# Macrophage issues



Isolate immune  
scavenger cells  
(macrophages)

How do macrophages from healthy animals compare to those prone to autoimmune diabetes?

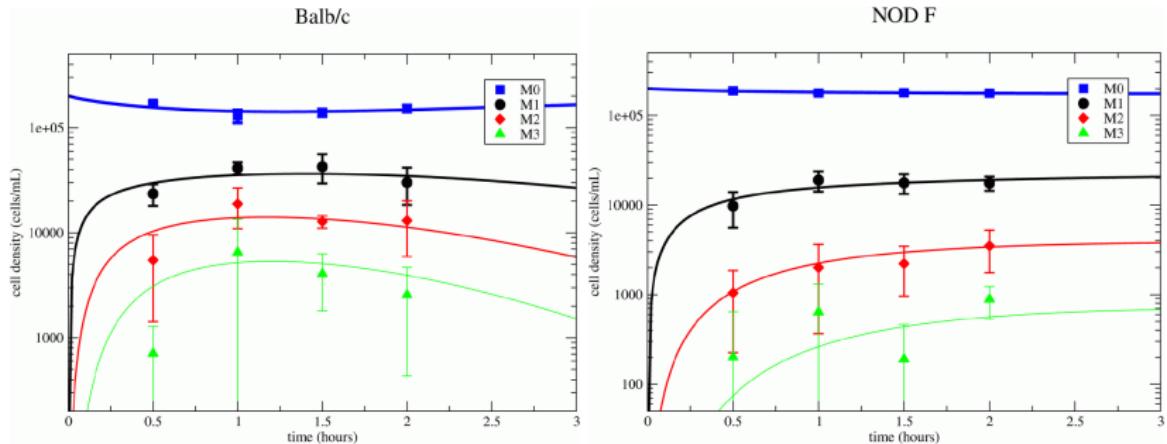


Expts:

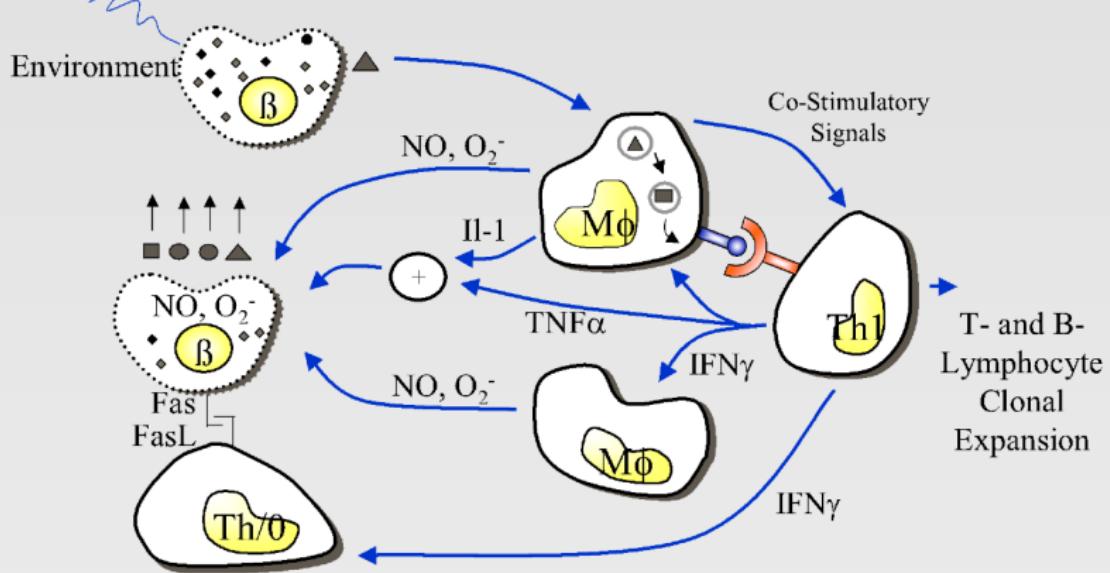
Count number of macrophages with  
1, 2, 3, ... etc engulfed cells.

# Quantifying the macrophage effect

We also use the full model differential equations and an optimization method (Marquardt Levenberg) to fit parameter values to the model and its variants.

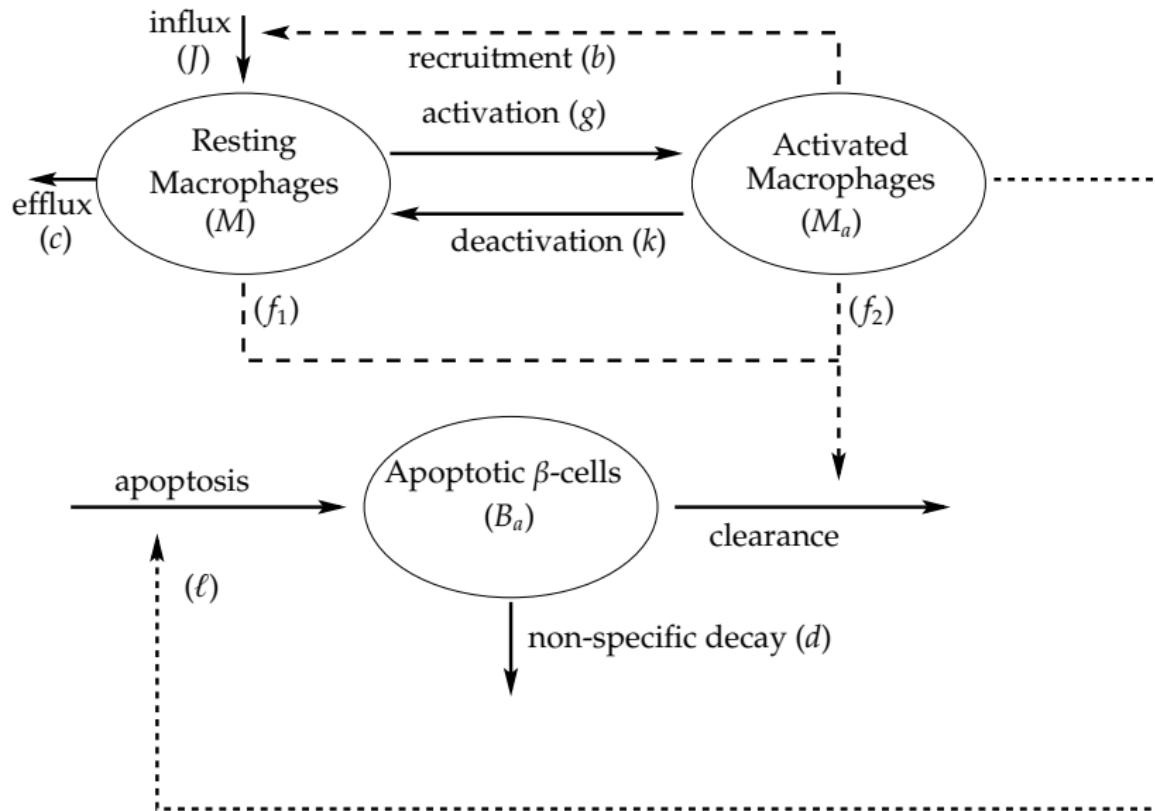


## Copenhagen Model of Type 1 Diabetes



J. Nerup, original presentation

# Basic model

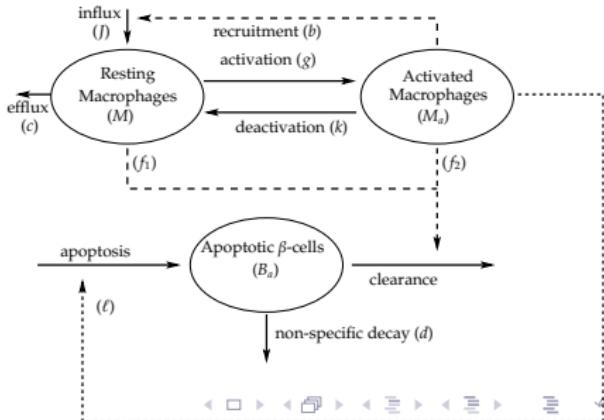


# Corresponding equations

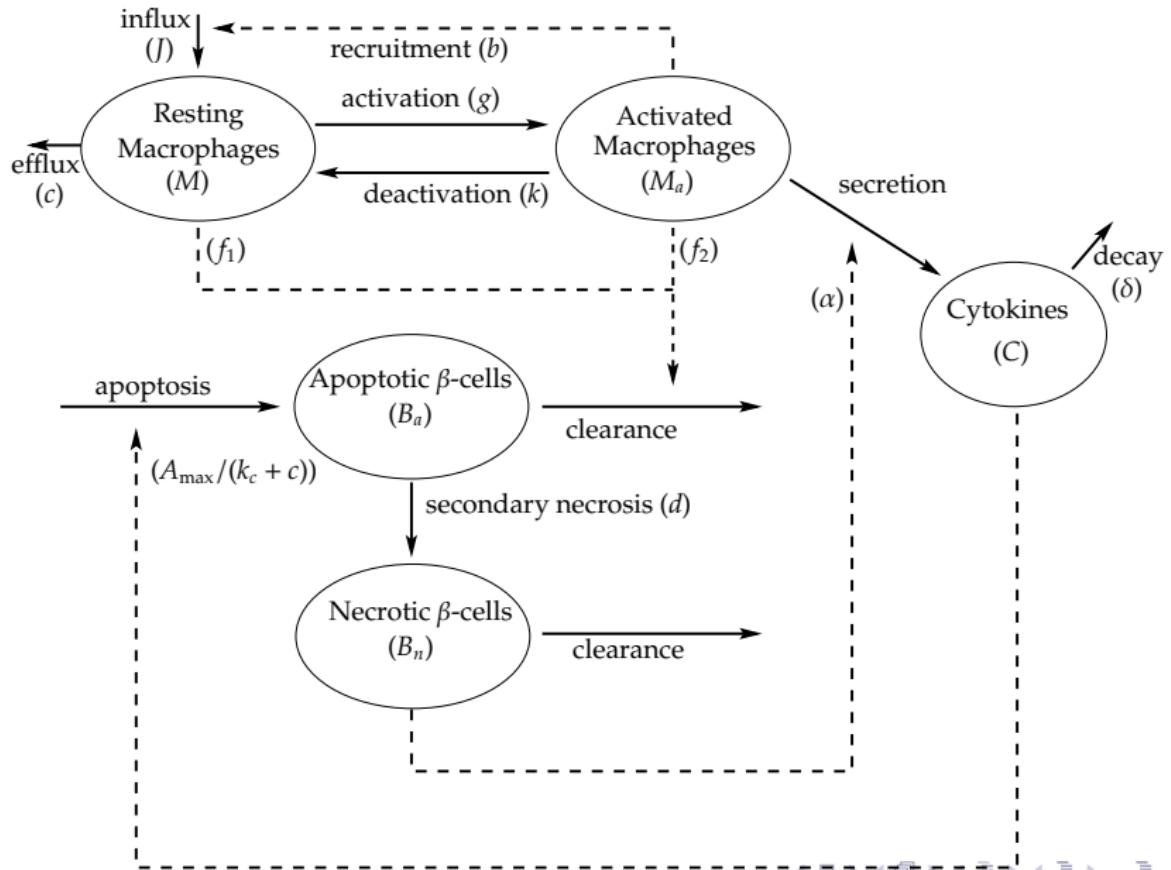
$$\frac{dM}{dt} = J + (k + b)M_a - cM - gMB_a - e_1M(M + M_a)$$

$$\frac{dM_a}{dt} = gMB_a - kM_a - e_2M_a(M + M_a)$$

$$\frac{dB_a}{dt} = W(t) + \ell M_a - f_1 MB_a - f_2 M_a B_a - dB_a$$



# Extended model



# Corresponding equations

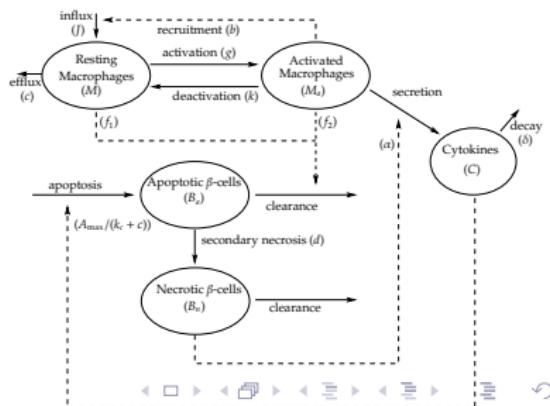
$$\frac{dM}{dt} = J + (k + b)M_a - cM - gMB_a - e_1M(M + M_a)$$

$$\frac{dM_a}{dt} = gMB_a - kM_a - e_2M_a(M + M_a)$$

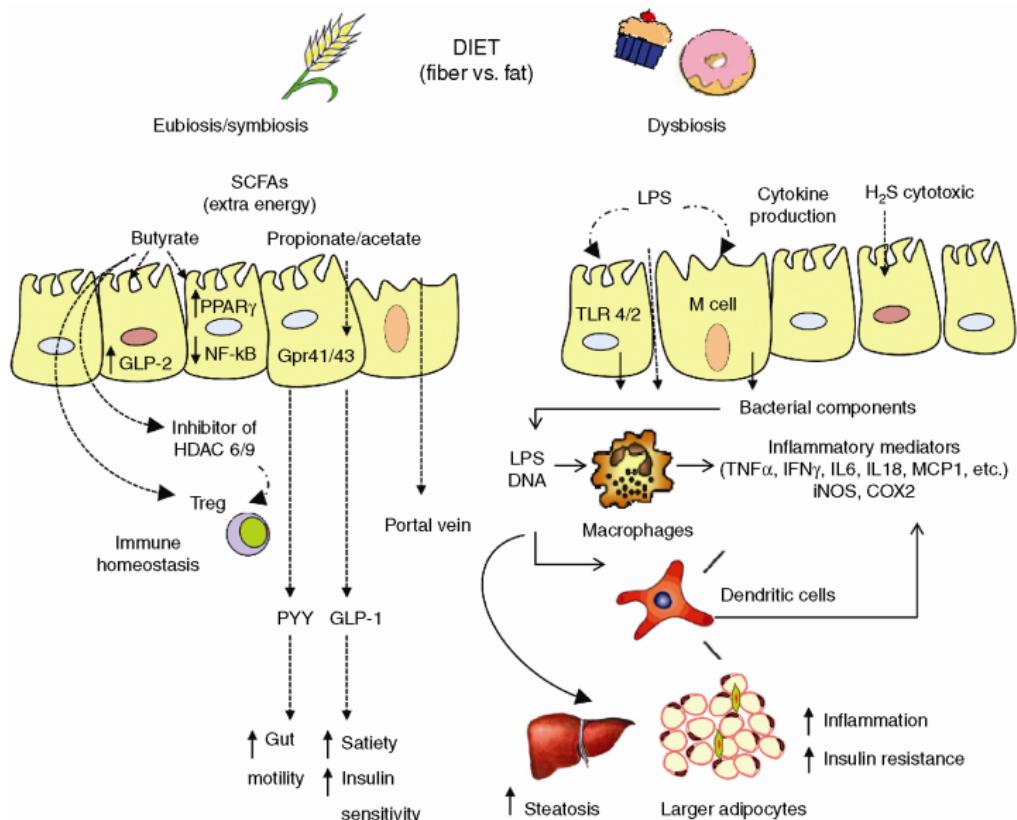
$$\frac{dB_a}{dt} = \frac{A_{max}C}{k_c + C} - f_1MB_a - f_2M_aB_a - dB_a$$

$$\frac{dB_n}{dt} = dB_a - f_1MB_n - f_2M_aB_n$$

$$\frac{dC}{dt} = I + \alpha B_n M_a - \delta C$$



# Cytokine level (e.g., IL1- $\beta$ ) and the gut



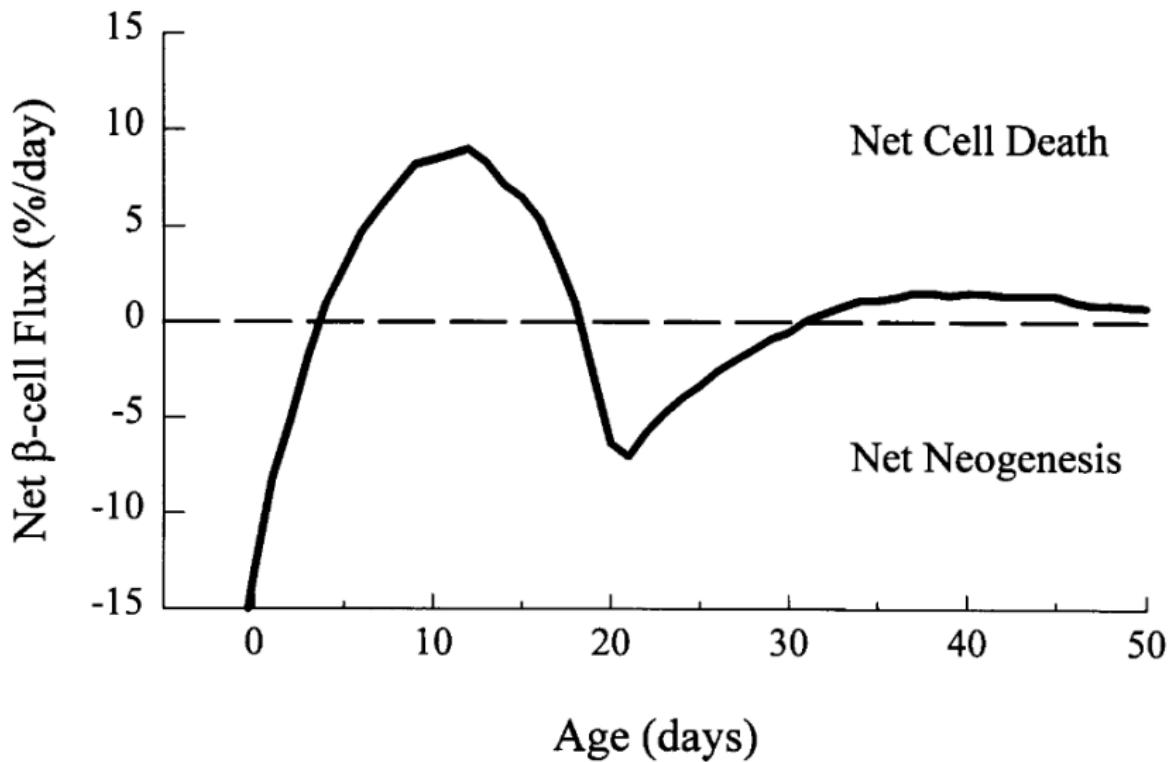
# Realistic, carefully measured parameters

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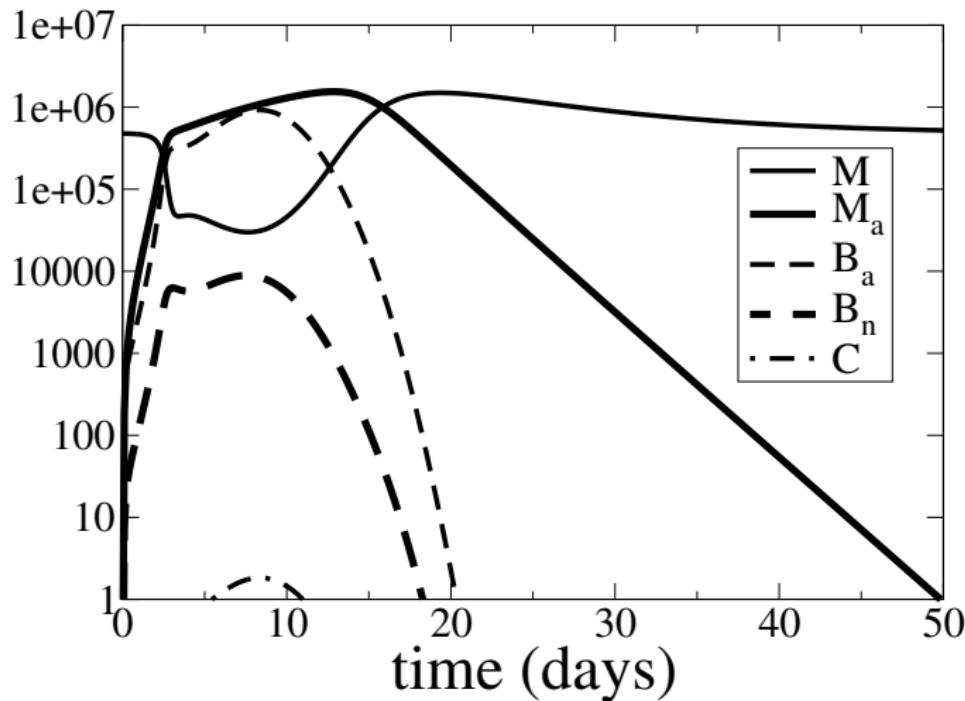
parameter	meaning	Balb/c	NOD	units
$J$	normal macrophage ( $M$ ) influx	5		$\times 10^4 \text{ cells ml}^{-1} \text{ d}^{-1}$
$c$	macrophage egress rate	0.1		$\text{d}^{-1}$
$b$	recruitment rate of $M$ by $M_a$	0.09		$\text{d}^{-1}$
$\ell$	$B_a$ apoptosis induced per $M_a$	0.41		$\text{d}^{-1}$
$d$	$B_a$ non-specific decay rate	0.5		$\text{d}^{-1}$
$k$	$M_a$ deactivation rate	0.4		$\text{d}^{-1}$
$g=f_1$	basal phagocytosis rate per $M$	2	1	$\times 10^{-5} \text{ ml cell}^{-1} \text{ d}^{-1}$
$f_2$	activated phagocytosis rate per $M_a$	5	1	$\times 10^{-5} \text{ ml cell}^{-1} \text{ d}^{-1}$
$e_1 = e_2$	anti-crowding terms		1	$\times 10^{-8} \text{ cell}^{-1} \text{ d}^{-1}$

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# Apoptotic wave

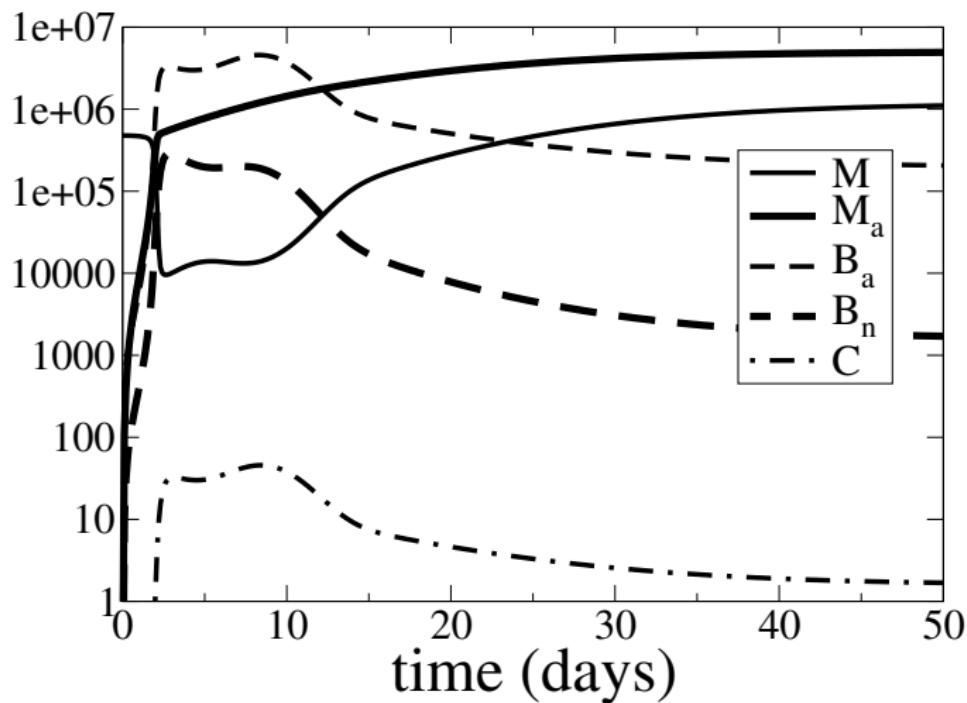


## Balb/c



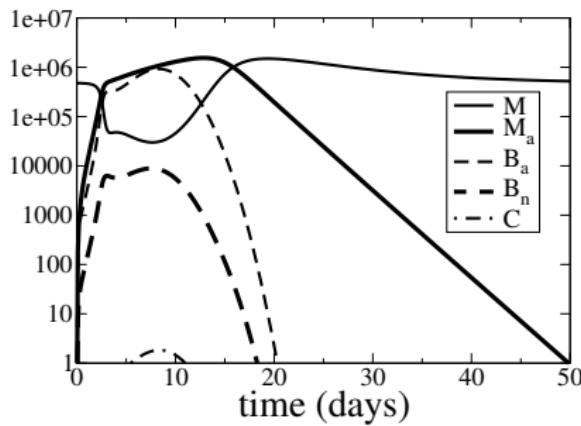
# Time dynamics, NOD mice

NOD

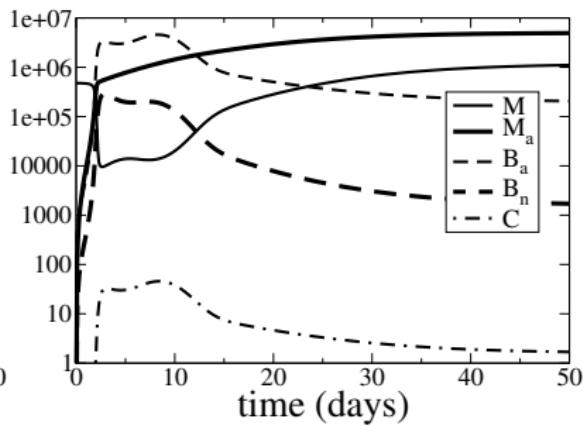


# Running a model vs understanding

Balb/c



NOD



# Simplifications:

- variables behave in the same manner: merge them
- variables change very slowly: treat them as a parameter
- variables change very fast: make a Quasi-Steady-State assumption

# Acknowledgements

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