

Using mathematics to understand why insulin resistance is bad for you, but can be good for your cells

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Type 2 diabetes



Figure: <http://www.cdc.gov/diabetes/pubs/statsreport14/diabetes-infographic.pdf>

What is type 2 diabetes?

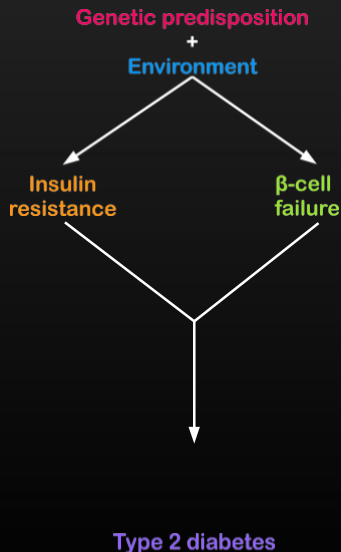
Two major players:

- glucose main energy source for most cells
- insulin produced by pancreatic β cells; signals cells to take up glucose from blood

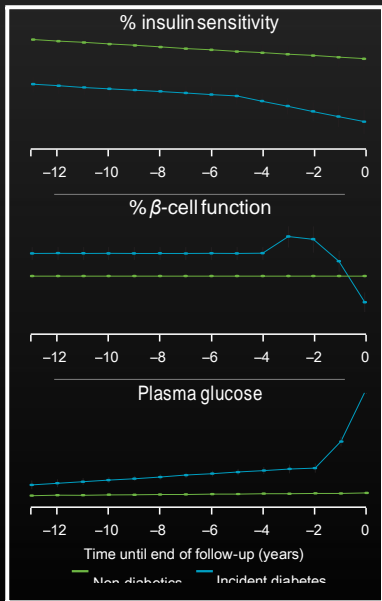
The disease:

- defined by severe hyperglycemia
- caused by combination of
 - insulin resistance
 - β -cell failure
- influenced by genetics and environment
-

characterized by insufficient insulin



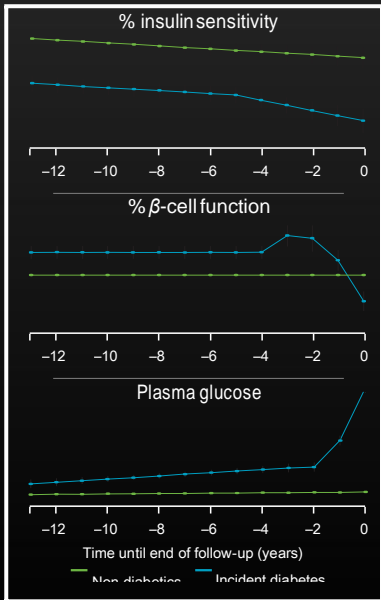
Type 2 diabetes dynamics



- decline in insulin sensitivity* with time
- severe insulin resistance in diabetics
- β -cell compensation for insulin resistance
- β -cell failure initiates diabetic hyperglycemia

* insulin sensitivity $\frac{1}{\text{insulin resistance}}$

Type 2 diabetes dynamics



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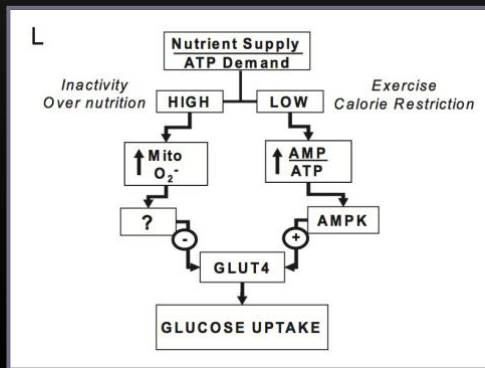
* insulin sensitivity $\frac{1}{\text{insulin resistance}}$

The problem: precise mechanisms of the development of insulin resistance and β -cell dysfunction are unclear.

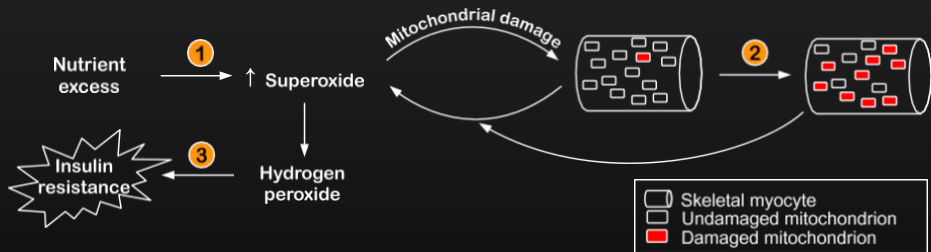
Insulin resistance is a cellular antioxidant defense mechanism

Kyle L. Hoehn^{a,1,2,3}, Adam B. Salmon^{b,1}, Cordula Hohnen-Behrens^a, Nigel Turner^a, Andrew J. Hoy^a, Ghassan J. Maghzal^c, Roland Stocker^c, Holly Van Remmen^b, Edward W. Kraegen^a, Greg J. Cooney^a, Arlan R. Richardson^b, and David E. James^{a,2}

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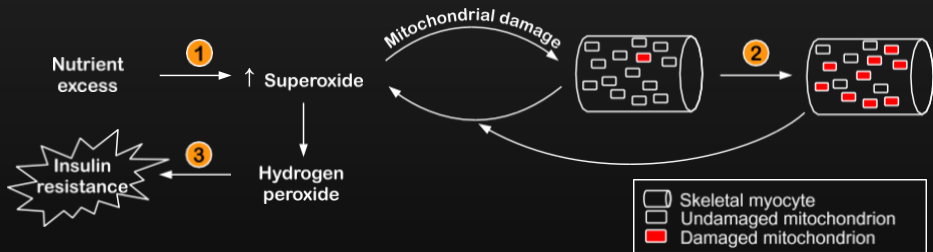


Skeletal muscle insulin resistance



Oxidative stress accumulation of reactive oxygen species,
e.g., superoxide, hydrogen peroxide

Skeletal muscle insulin resistance

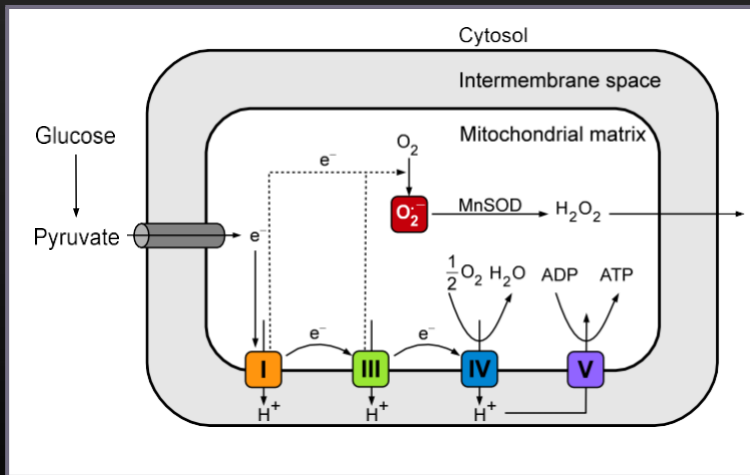


1 superoxide production

2

3

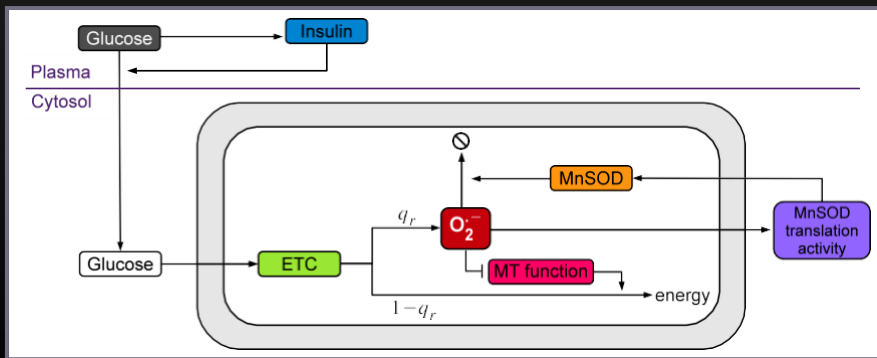
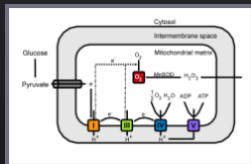
Subsystem I: superoxide production



- e^- : electron
- H^+ : proton
- O_2^- : superoxide

- MnSOD: antioxidant, manganese superoxide dismutase
- H_2O_2 : hydrogen peroxide

Subsystem I: superoxide production



■ ETC: electron transport chain

■ MT: mitochondrial

Subsystem I equations

? ΔG reference parameter for food intake, with σ an increasing function of ΔG .

? F mitochondrial function variable; form specified in feedback coupling.

Plasma glucose:

$$\frac{dG}{dt} = \underbrace{|\{z\}^\sigma}_{\text{food intake}} + \underbrace{|\{z\}^{hg}}_{\text{production}} - \underbrace{k_g G}_{\text{insulin-independent uptake}} - \underbrace{\frac{SGI}{K_1 K_2}}_{\text{insulin-dependent uptake}}$$

Plasma insulin:

$$\frac{dI}{dt} = \underbrace{h_i B \frac{G^2}{G + G_h}}_{\text{production}} - \underbrace{k_i I}_{\text{clearance}} |\{z\}$$

Intracellular glucose:

$$\frac{dG_i}{dt} = \underbrace{\gamma_1 \frac{SGI}{K_1}}_{\text{uptake from plasma}} - \underbrace{k_{qz} G_i}_{\text{glucose processing}}$$

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Intracellular glucose:
$$\frac{dG_i}{dt} = \underbrace{v_1 \frac{S G I}{\{Z\}}}_{\text{uptake from plasma}} - \underbrace{k_{g_i} G_i}_{\text{glucose processing}}$$

ETC activity:
$$\frac{dC}{dt} = \underbrace{v_2 k_{g_i} G_i (C_{tot} - C)}_{\text{activation}} - \underbrace{k_c C}_{\text{production}} (1 - q_r) F + \underbrace{q_r}_{\text{superoxide production}} \{Z\}$$

Superoxide:
$$\frac{dR_s}{dt} = \underbrace{k q_c C}_{\text{production}} - \underbrace{k_{rs} R_s}_{\text{removal}} \delta \{Z\} - \underbrace{\delta \{Z\}}_{\text{inactivation}}$$

Antioxidant:
$$\frac{dA_s}{dt} = \underbrace{q_a E}_{\text{production}} - \underbrace{q_a k_{rs} R_s A_s}_{\text{superoxide toxicity}}$$

Antioxidant production:
$$\frac{dE}{dt} = \underbrace{v_2 k_{g_i} G_i}_{\text{activation}} \{Z\} - \underbrace{k_{eg} E}_{\text{deactivation}} \{Z\}$$

Mitochondrial dysfunction: assumptions

1 Homogeneity:

—→ altered respiratory activity

- damaged mitochondrial lipids/proteins

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Feedback model I: Mitochondrial Inefficiency Model (MIM)

$$\frac{dL}{dt} = \xi(1-L) \frac{X^2}{X^2 + \lambda^2}, \text{ where } X = \frac{R_s/(R_s + A_s)}{R_{s0}/(R_{s0} + A_{s0})} - 1.$$
$$\Rightarrow F_{MIM} = 1 - L$$

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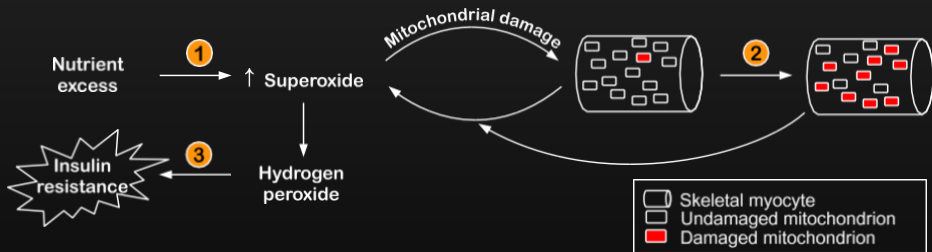
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2 Heterogeneity:

→ abnormal population dynamics

- 'sufficient' mutant mtDNA clonal expansion
- mitochondrial swelling and membrane permeability
- stress from accumulation of damaged content

Skeletal muscle insulin resistance



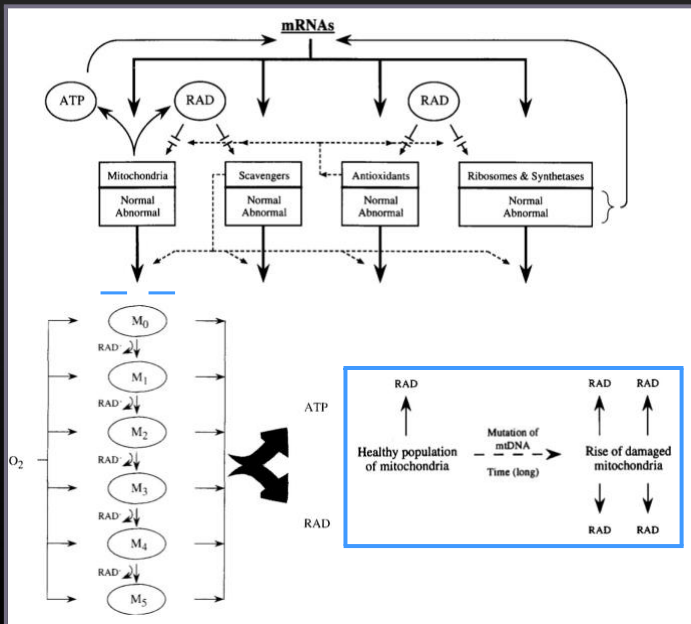
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2 mitochondrial selection

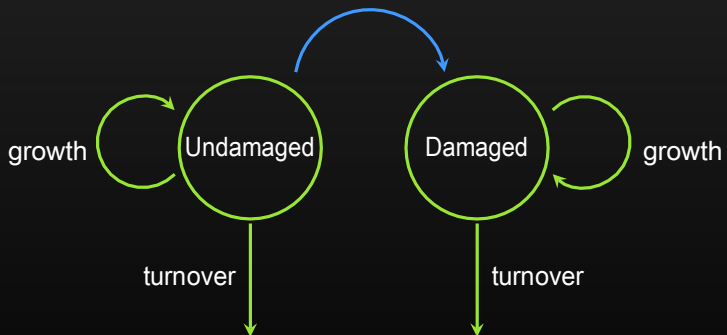
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MARS: A network theory of aging

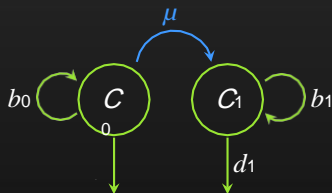
Mitochondria
A aberrant proteins
Radicals
Scavengers



Subsystem II: mitochondrial selection



Modeling mitochondrial selection: setup



Two-state stochastic model

$M_0(t)$:= the number of healthy (class C_0) mitochondria at time t

$M_1(t)$:= the number of damaged (class C_1) mitochondria at time t

$M_0(t) + M_1(t) = K$ for all t ; K is constant

Parameters

- selection parameters $\left(\begin{array}{l} s_r := \text{fractional replicative difference} \\ s_m := \text{fractional turnover difference} \end{array} \right) \in (-1, 1)$

- $b_1 = (1 + s_r)b_0$

- $d_1 = (1 + s_m)d_0$

“Null selection” \Rightarrow

$$s_m = s_r = 0$$

Modeling mitochondrial selection: state transitions

Assume that each mitochondrial turnover event results in a growth event.

Transition matrix

$$A = \begin{array}{c} \begin{array}{cc} \text{"} & \# \end{array} \\ \begin{array}{cc} \Delta M_0 & \begin{array}{cc} \text{"} & \# \end{array} \\ \Delta M_1 & \begin{array}{cc} \text{"} & \# \end{array} \end{array} \begin{array}{ccc} 1 & -1 & 0 \\ -1 & 1 & 0 \end{array} [A_1 \ A_2 \ A_3] \end{array}$$

Transition probabilities

$$\begin{array}{l} p_i \Pr(A_1/M_0 \ i) \\ q_i \Pr(A_2/M_0 \ i) \\ \Rightarrow \Pr(A_3/M_0 \ i) \end{array} \begin{array}{l} \frac{d_1(K-i)}{d_0 i + d_1(K-i)} \\ \frac{(1-\mu)b_0 i}{b_0 i + b_1(K-i)} \\ \frac{d_0 i + d_1(K-i)}{d_0 i + d_1(K-i)} \\ \frac{\mu b_0 i + b_1(K-i)}{b_0 i + b_1(K-i)} \\ 1 - p_i - q_i \end{array} \begin{array}{l} \{z\} \\ \{z\} \\ \{z\} \\ \{z\} \end{array} \begin{array}{l} \text{death from } C_1 \\ \text{birth to } C_0 \\ \text{death from } C_0 \\ \text{transition/birth to } C_1 \end{array}$$

Mean time to total damage

Let $T_i :=$ the expected time to total damage starting from i healthy mitochondria.

Let $E_i :=$ the mean waiting time between events for $M_0 = i$, i.e.

$$E_i = [d_0 i + d_1 (K - i)]^{-1}.$$

$$T_i = q_i (T_{i-1} + E_i) + (1 - p_i - q_i) (T_i + E_i) + p_i (T_{i+1} + E_i)$$

$$\Rightarrow -E_i = \underbrace{q_i T_{i-1}}_{\text{lose a healthy one}} - \underbrace{(p_i + q_i) T_i}_{\text{no change}} + \underbrace{p_i T_{i+1}}_{\text{gain a healthy one}}$$

Solution

$$T_i = T_0 + \sum_{j=0}^{i-1} \eta_{j+1} + \sum_{m=0}^{K-i-2} \sum_{n=m+1}^{K-i-1} \eta_{K-m} \rho_{K-n}$$

for $i = 1, \dots, K-1$, with $\eta_i = \frac{E_i}{q_i}$, $\rho_i = \frac{p_i}{q_i}$, $T_0 = 0$ and $T_K = \eta_K + T_{K-1}$.

With null selection and constant μ : $T_K \approx 400$ years.

Superoxide-to-damage feedback

- damage transition: $\mu(t) := \mu_0 \left(1 + \rho \left(\frac{R_s(t)}{R_{s0}} - 1 \right) \right)^h$

- probability distribution: $\pi_j(t) := \Pr(M_1 = j)$

Master equation:

$$\begin{aligned} \frac{d\pi_0}{dt} &= -\hat{q}_0 \pi_0 + \hat{p}_1 \pi_1, \dots, \\ \frac{d\pi_j}{dt} &= \hat{q}_{j-1} \pi_{j-1} - (\hat{q}_j + \hat{p}_j) \pi_j + \hat{p}_{j+1} \pi_{j+1}, \dots, \\ \frac{d\pi_K}{dt} &= \hat{q}_{K-1} \pi_{K-1} - \hat{p}_K \pi_K \end{aligned}$$

- damage likelihood:

$$D(t) = \Pr(M_1 \geq 1) = \frac{1}{K} \sum_{j=1}^K \pi_j(t)$$

Feedback models I – IV: specifying F

Complex: $\frac{dC}{dt} = v_2 k_{gi} G_i (C_{tot} - C) - k_c C [(1 - q_r) F + q_r]$

1

$$F_{MIM} = 1 - L$$

2

$$F_{DMM} = 1 - D$$

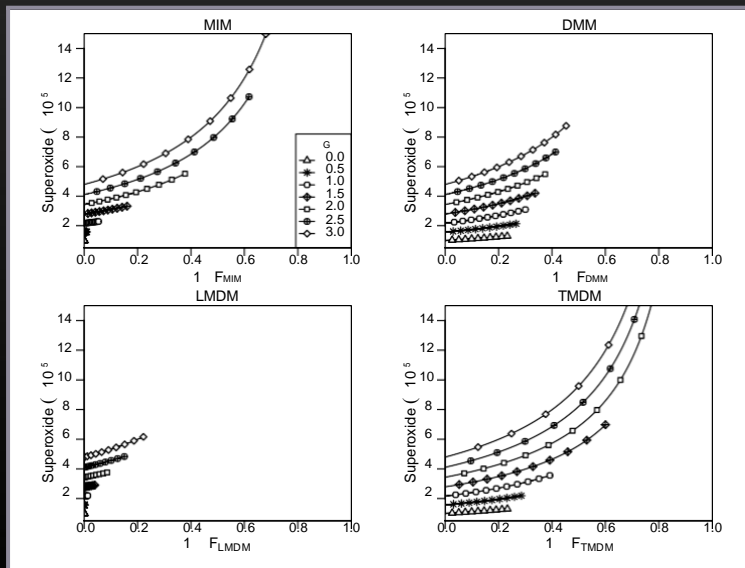
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$$F_{LMDM} = (1 - D) + D \cdot (1 - L)$$

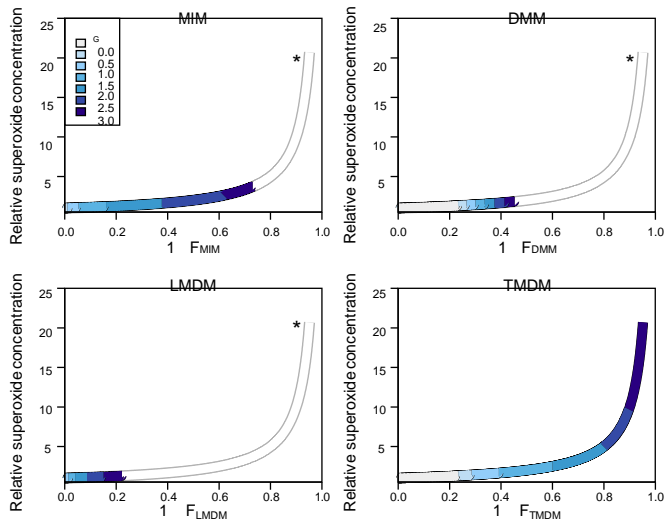
4


$$F_{\text{TMDM}} = (1 \ -L)(1 \ -D)$$

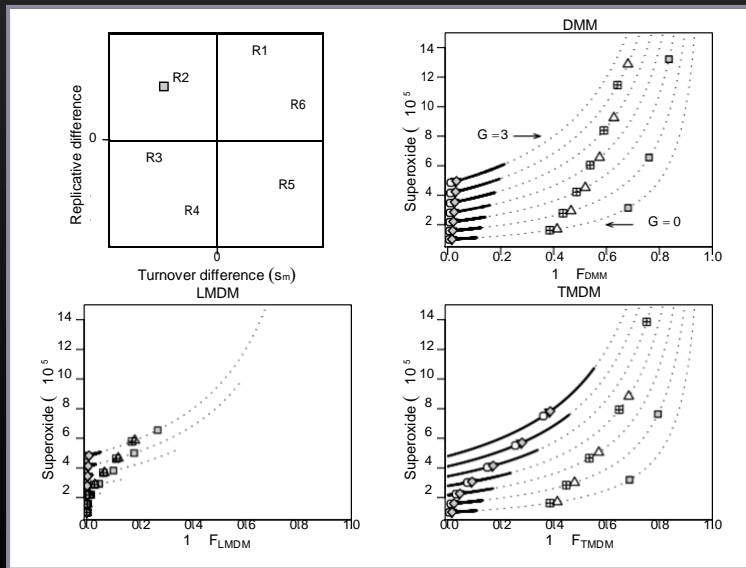
Results I: null selection



Results II: assessing the timing of dysfunction

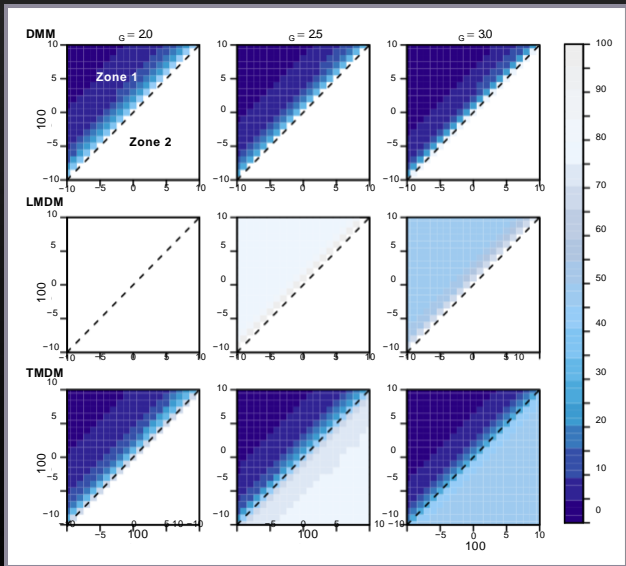


Results III: response to mitochondrial selection



Recall: $s_r = (b_1/b_0) - 1$; $s_m = (d_1/d_0) - 1$

Results IV: response to selection parameters



- compute age at which superoxide concentration exceeds threshold of $10^{-4} \mu\text{M}$
- Zone 1: $s_r > s_m$
Zone 2: $s_r < s_m$
- physiological restriction: $s_m > 0$

Why insulin resistance is bad for you, good for your cells

In susceptible individuals:

Intracellular response

↑ glucose uptake (good for you, bad for the cell)

↑ superoxide production

mitochondrial dysfunction

↑ oxidative stress

↓ glucose uptake (bad for you); ↓ superoxide production (good for the cell)

stress signal activation; impaired insulin signaling

Systemic response

↓ glucose uptake (bad for you)

↑ glucose uptake

Thank you!

Acknowledgements



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The logo for North Carolina State University, consisting of a red rectangle with the text 'NC STATE UNIVERSITY' in white.

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Department of Mathematics

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References

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