First meeting of the NeuroMod Institute 1-2 July 2019, Fréjus

Modelling of physiological and pathological states in neuroscience: exchanges among theoreticians and experimentalists

B. Cessac (Inria), M. Mantegazza (IPMC)

In a virtuous cycle, modelling can use data obtained by experiments to capture essential features and organizing principles of biological systems at multiple scales, and provide hypotheses that can be directly tested with experiments. This can allow to gain important information on both physiological functions and pathological dysfunctions. However, the virtuous cycle is not always ignited and sometimes experimentalists are sceptical of modelling data, feeling that they are too far away from real biology to be informative. There is a delicate trade off between the theoreticians' quest of organizing mechanisms and principles obtained by simplifying reality, and the experimentalists will of staying as close as possible from the system under study. We will discuss what experimentalists expect from theoreticians and vice-versa, what are some new contemporary challenges in neuroscience, and how modelling could contribute to address them.

The modeller point of view

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A set of equations:

Variables (quantity evolving in time and space), Parameters (quantity tuned "by hand"),

Causal interactions between variables depending on parameters,

Space time evolution,

Scales

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$$\begin{cases} C\frac{dV}{dt} = -g_L(V - V_L) - g_{Na}^M m^3 h (V - V_{Na}) - g_K^M n^4 (V - V_K) + i_{ext} \\ \frac{dn}{dt} = -\beta_n n + \alpha_n (1 - n) = \frac{n_\infty(V) - n}{\tau_n(V)} \\ \frac{dm}{dt} = -\beta_m m + \alpha_m (1 - m) = \frac{m_\infty(V) - m}{\tau_m(V)} \\ \frac{dh}{dt} = -\beta_h h + \alpha_h (1 - h) = \frac{h_\infty(V) - h}{\tau_h(V)}. \end{cases}$$

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Studying equations

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 $C\frac{dV}{dt} = -g_{L}(V - V)$ Mathematics is not (only) at tool, it provides a conceptual framework which guides but also constrains the analysis. $\frac{dn}{dt} = -\beta_{n}n + \alpha_{n}(1-n) = \frac{n \otimes (V) - n}{\tau_{n}(V)}$ $\frac{dm}{dt} = -\beta_{n}m + \alpha_{m}(1-m) = \frac{n \otimes (V) - m}{\tau_{n}(V)}$

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expect.

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Simulation, is not a magic stick, don't simulate if you have no idea of what to expect.

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Studying equations

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Each equation imposes implicit constraints on the model-description

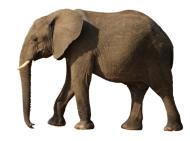
What is not a model ?

A model is not reality

Simplify to understand



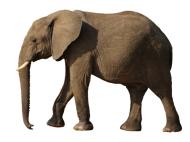




What is their common feature ?

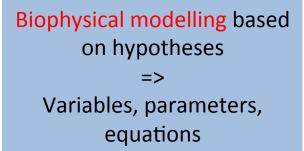






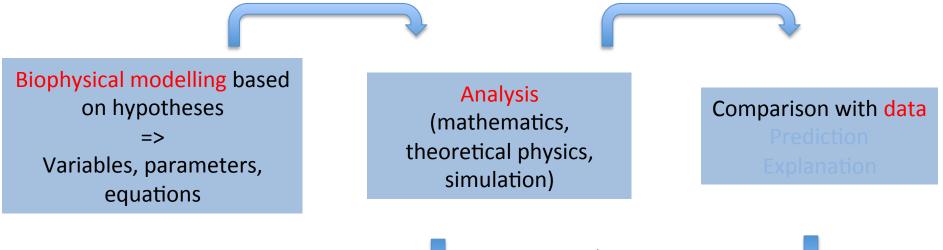
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Simplify to understand



Analysis (mathematics, theoretical physics, simulation)









Data

Modellers are not data analysers.

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Models may help unravelling this structure

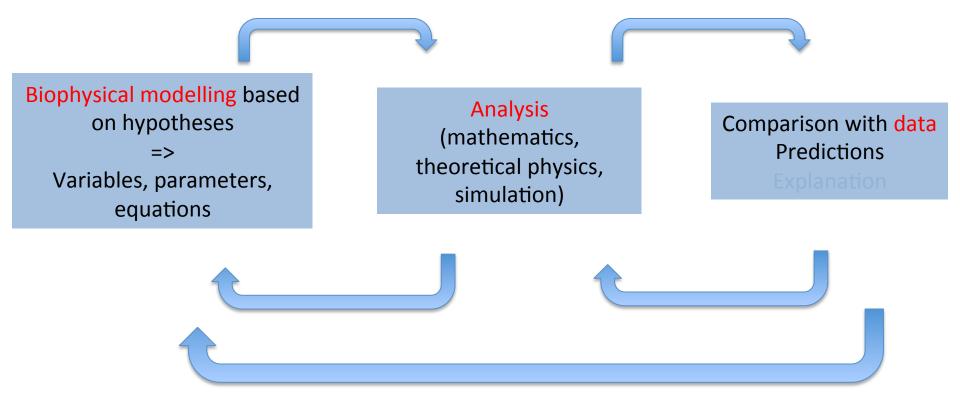
Goals of modellers

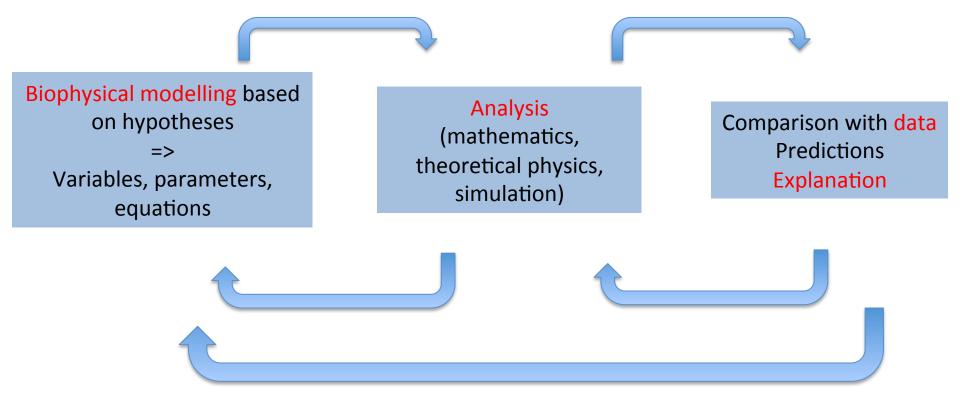
The goal is not to tell biologists something they already know

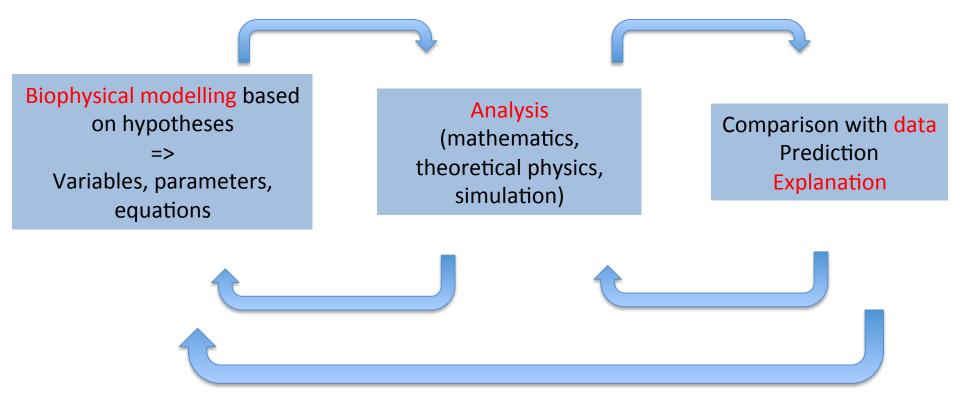
Not to reproduce experiments / "reality" => just a step

Make predictions

Explain







To predict is not to explain (R. Thom)

Generic mechanisms, organizing principles

A neurobiologist point of view

A neurobiologist point of view *

* Disclaimer: I think modeling is scientifically useful, not just a way to obtain funding...

The best model of a cat is a cat, possibly the same cat.



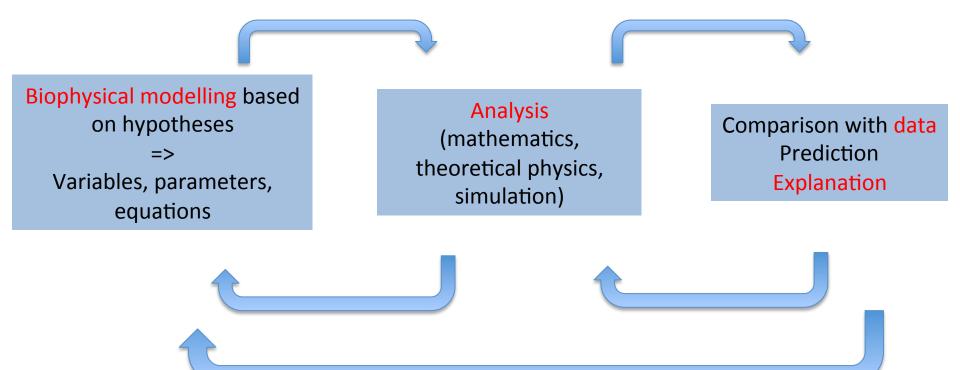
Uwe Heinemann 1944-2016

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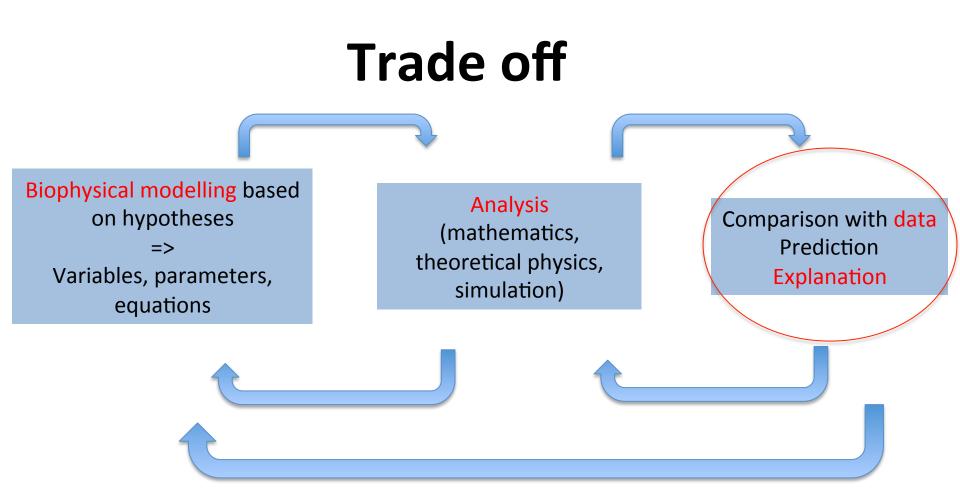
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A perceived feeling of individuality, possibly linked to robustness of biological systems



To predict is not to explain (R. Thom)

Generic mechanisms, organizing principles



To predict is not to explain (R. Thom)

Generic mechanisms, organizing principles

Trade off

Parameters and initial values of variables

Experimentalist

Modeler

Explanations and predictions (comparison with experimental data)

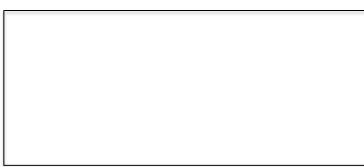
Modeling

Top down



Phenomenological prediction.

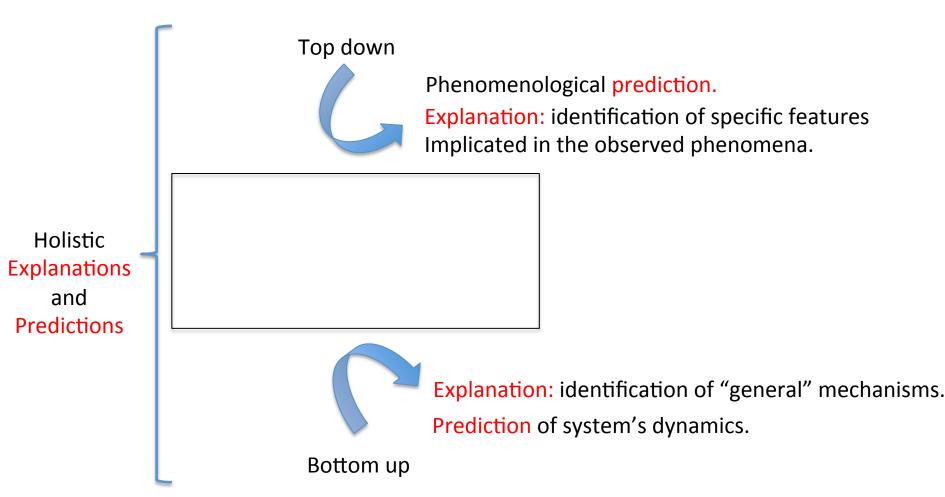
Explanation: identification of specific features implicated in the observed phenomena.



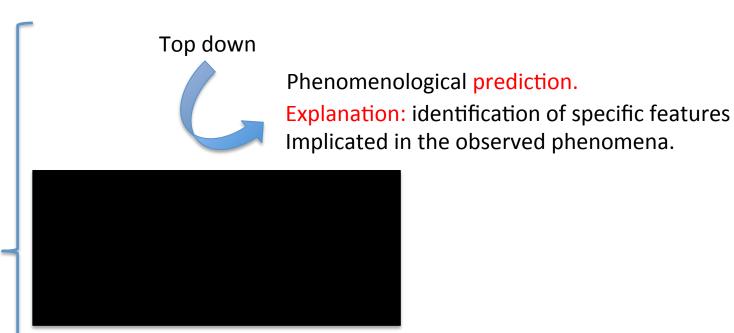
Explanation: identification of "general" mechanisms.
Prediction of system's dynamics.

Bottom up

Modeling



Modeling



Explanation: identification of "general" mechanisms. Prediction of system's dynamics.

Bottom up

?



Quantity of information (measurements); how to implement it



function

Impossible to measure everything (robustness is a limit, among others) Impossible to measure everything (robustness is a limit, among others)

Very difficult to implement models of complex systems and fit them to biological data (at every single level, even more with interconnected levels)

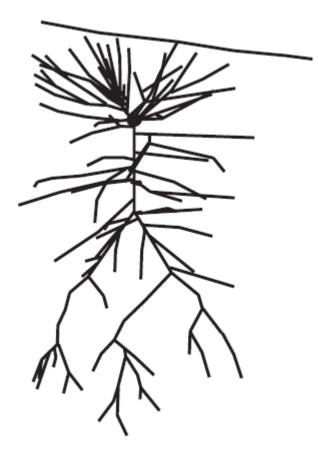
A personal experience

A personal experience (of a poor neurobiologist)

Functional effect of human pathogenic mutations of voltage-gated sodium channels

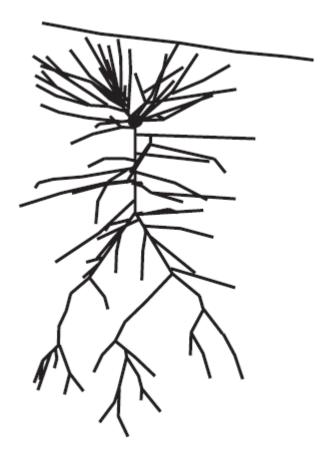
Functional effect of human pathogenic mutations of voltage-gated sodium channels

Realistic computer model of a neuron (*morphology, type and subcellular distribution of conductances*).



Functional effect of human pathogenic mutations of voltage-gated sodium channels

Realistic computer model of a neuron (*morphology, type and subcellular distribution of conductances*).



Slow, cumbersome to implement modifications, sometime difficult interpretation of results

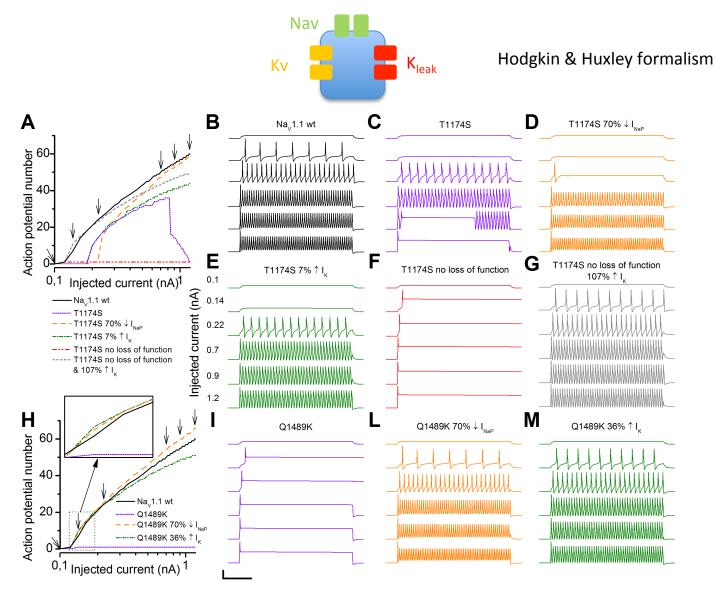
Extremely simplified computer model of a neuronal axon initial segment



Hodgkin & Huxley formalism

$$\begin{cases} C\frac{dV}{dt} = -g_L(V - V_L) - g_{Na}^M m^3 h (V - V_{Na}) - g_K^M n^4 (V - V_K) + i_{ext} \\ \frac{dn}{dt} = -\beta_n n + \alpha_n (1 - n) = \frac{n_\infty(V) - n}{\tau_n(V)} \\ \frac{dm}{dt} = -\beta_m m + \alpha_m (1 - m) = \frac{m_\infty(V) - m}{\tau_m(V)} \\ \frac{dh}{dt} = -\beta_h h + \alpha_h (1 - h) = \frac{h_\infty(V) - h}{\tau_h(V)}. \end{cases}$$

Extremely simplified computer model of a neuronal axon initial segment



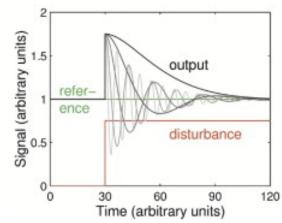
Cestele et al 2013 PNAS Cestele et al 2013 Epilepsia Cestele et al 2015 Neurobiology Disease

The engineering definition

the persistence of a system's characteristic behavior under perturbations or conditions of uncertainty.

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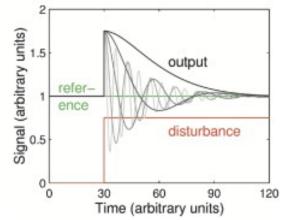
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Redundancy, Modularity and Feedback controls

The engineering definition

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Redundancy, Modularity and Feedback controls

Biological robustness

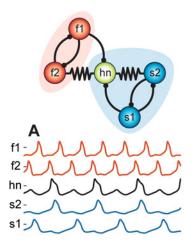
Degeneracy

Structurally different entities perform overlapping functions.

They can act in different combinations of type of entity and "strength" of each entity.



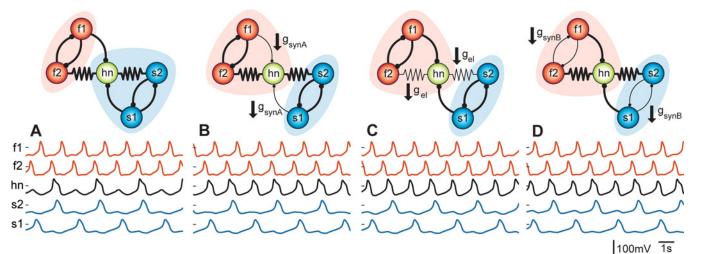
G. Edelman, J. Gally Degeneracy and complexity in biological systems PNAS 2001





Marder et al. 2016 Curr Op Neurobiol

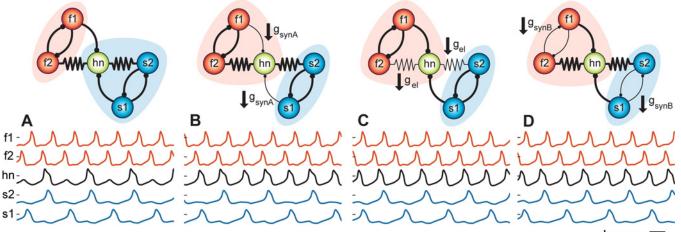
Eve Marder





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Marder et al. 2016 Curr Op Neurobiol

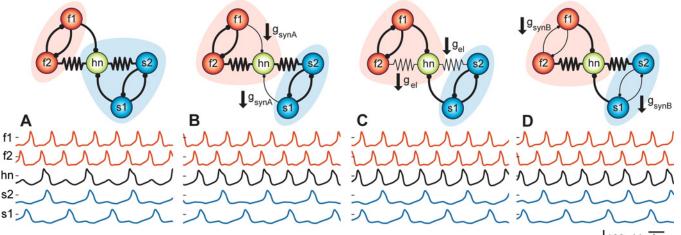


Marder et al. 2016 Curr Op Neurobiol

Brameter A

100mV 1s

Eve Marder

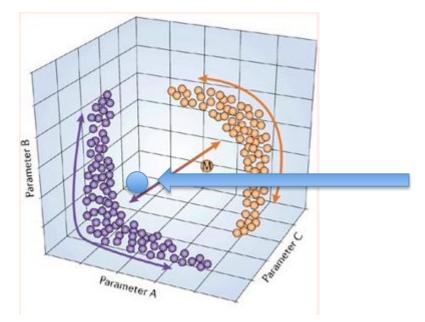




100mV 1s

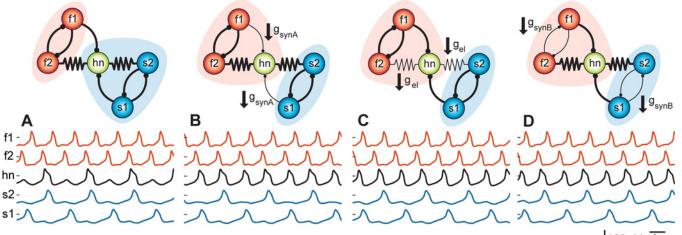
Eve Marder

Marder et al. 2015 Curr Op Neurobiol



The commonly used averages identify a combination of values that does not exist!

Marder & Goaillard 2006 Nature Neurosci

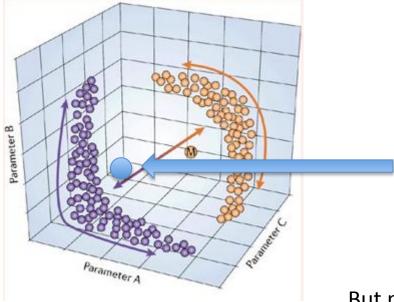




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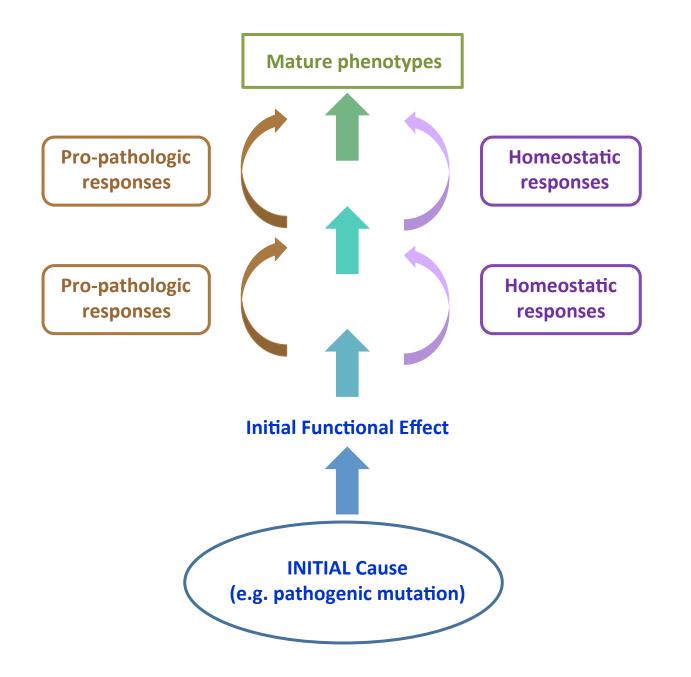
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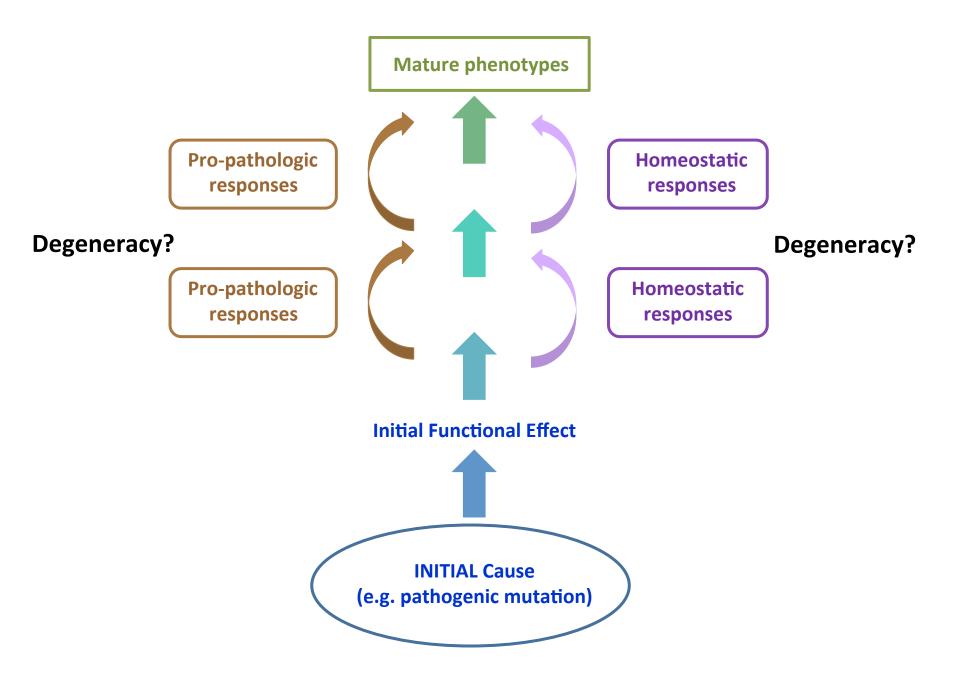
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But normally we obtain and provide mean values....







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