

# Fitting Models to Data in Vector-Borne Disease Systems

Samraat Pawar



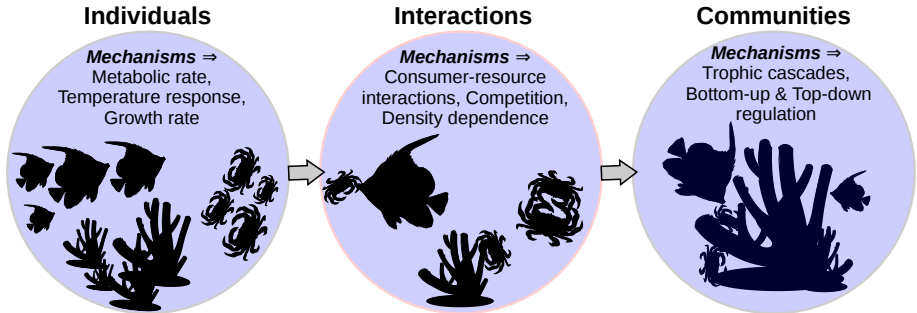
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# MECHANISTIC VS. PHENOMENOLOGICAL MODELS

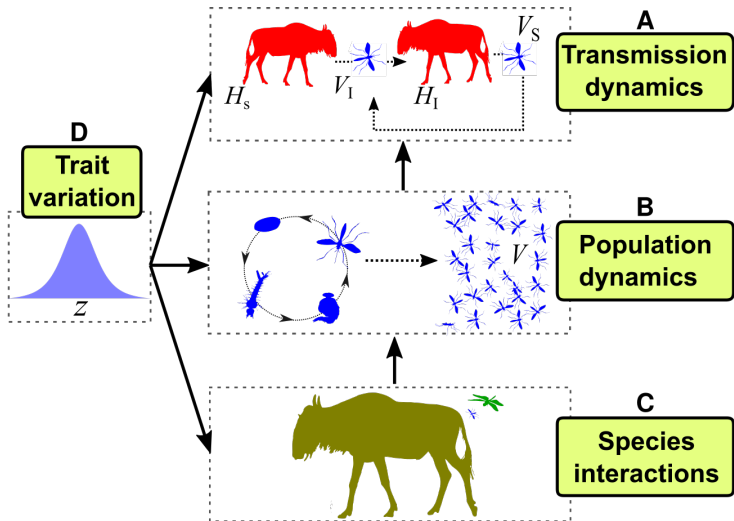
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- For example, disease invasions, outbreaks and spread: Why cycles? Why traveling waves? — What mechanisms operate? (Vector-parasite interaction? Vector-environment interaction?)

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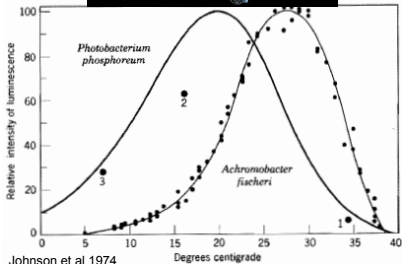
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- But is this REALLY mechanistic? What are  $r$  and  $k$  really?
- Many (including yours truly!) now argue that we have not progressed far enough because the first level has been ignored!

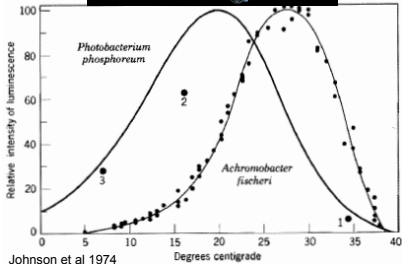
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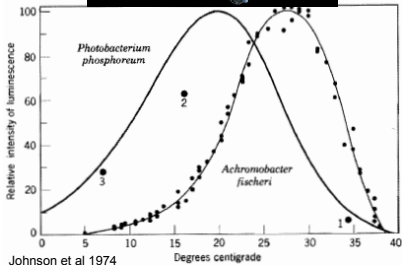
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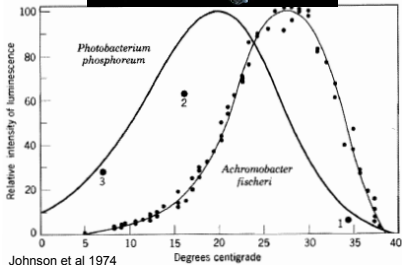
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- *What about alternative models?*

# MODELLING, AND FITTING MODELS TO DATA: WHAT'S THE BIG IDEA?

- If possible, use biological knowledge to construct models
- See if the models “agree well” with data
- Whichever model “agrees best” is most likely to have the right mechanisms
- That's the one that's best for predictions (e.g. population cycles), estimating rates (e.g. population or individual growth rates), etc
- Don't use models you already know have the wrong mechanisms just because they are popular!
- Phenomenological models often perform better than mechanistic ones

# MODELS: HOW TO BUILD THEM?

- It's an art, takes practice
- Build models one mechanism at a time — in biology, it means start at the right level of organization!
- Always consider an alternative that is more parsimonious, even if it is phenomenological (the thermal performance curves example: Sharpe-Schoolfield, Briere, or Polynomial?)!
- For example, the Boltzmann-Arrhenius model is a good first try describe and uncover mechanisms underlying individual level rates (e.g., vector fecundity or development rate)
- The next step would be to include species interactions with temperature dependence of individuals (or go in an evolutionary direction)

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- This is an advance over the traditional “null hypothesis” approach in Biology
- Necessary for developing the advancement of Biology from from an observational and axiomatic discipline to one with general theories.
- Necessary for understanding the mechanisms underlying Biological (read, VBD) patterns and dynamics

# FITTING MODELS TO DATA

Two main ways to do it:

- One-step forecasting and machine learning (appropriate for discrete models) and time series data — focus in on maximizing ability to predict at the cost of mechanistic insights
- Ensemble fitting (appropriate for full time series or responses)
  - Least Squares methods
    - Linear
    - Non-linear
  - Likelihood based methods
    - Maximum Likelihood Estimation (MLE)
    - Bayesian

# ENSEMBLE FITTING

- These include MLE, Bayesian methods, and least squares optimization or fitting.
- Non-linear least squares (NLLS) fitting is a particularly versatile and powerful approach, because many mechanisms in biology and inherently non-linear.
- MLE/Bayesian methods are more robust if you are able to calculate the likelihood function.

# COMPARING AND SELECTING MODELS

- It's all about the “Likelihood” of a model:  
the set of parameter values of the model ( $\theta$ ) given outcomes ( $x$ ), equals the probability of those observed outcomes given those parameter values, that is,

$$\mathcal{L}(\theta|x) = P(x|\theta)$$

- The easiest thing to do for you is to use information theory (including AIC and BIC) to compare models.
- Both AIC and BIC use the *estimated likelihoods of a model*:  
AIC:  $-2 \ln[\mathcal{L}(\theta|x)] + 2p$   
Small sample AIC (AICc):  $-2 \ln[\mathcal{L}(\theta|x)] + 2p$   
BIC (Schwartz criterion):  $-2 \ln[\mathcal{L}(\theta|x)] + p \ln(n)$   
(where  $n$  = sample size,  $p$  number of free parameters)
- The lower the AIC or BIC, the better.

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(note  $n$  and  $p$ !)
- And BIC is  $n + n * \log(2 * \pi) + n * \log(rss / n) + (\log(n)) * (p + 1)$
- That is,  $\mathcal{L}(\theta|x) = -\frac{n}{2/\ln(RSS/n)}$
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Also note that:

- $R^2 = 1 - (rss/tss)$ , where  $tss$  is total sum of squares:  
`tss = sum((Observations - mean(Predictions)) ** 2)`  
(a useful measure of goodness of fit – you should report it)



# COMPARING AND SELECTING MODELS: MORE STUFF

- You can also calculate Akaike Weights, which is very useful/important when comparing  $> 2$  models. These weights can then be used to perform *model averaging*
- Model selection using the Likelihood-Ratio test (LRT) is another option when you are comparing 2 models
- Adjusted  $R^2$  can be used to get rigorous “idea” about how alternative models are performing.
- Very often, you will end up doing model simplification, especially in *for linear least squares model fitting* — starting with a complex model and then dropping terms till you have found a the most parsimonious version of the original model. There are functions in R to do this (of course!).

# READINGS

- Levins, R. (1966) The strategy of model building in population biology. *Am. Sci.* 54, 421–431.
- Johnson, J. B. & Omland, K. S. (2004) Model selection in ecology and evolution. *Trends Ecol. Evol.* 19, 101–108.
- Bolker, B. M. et al. (2013) Strategies for fitting nonlinear ecological models in R, AD Model Builder, and BUGS. *Methods Ecol. Evol.* 4, 501–512 .
- Some illustrative examples of (non-linear) model fitting to ecological/evolutionary data <https://groups.nceas.ucsb.edu/non-linear-modeling/projects>
- Additional readings at the end of Miniproject Chapter of your CMEE Notes