

1. A Brief History of SEM

Overview

1. What is SEM?
2. A History Lesson
3. From ANOVA to SEM

1.1 What is SEM?

1.1 What is SEM?

Structural

Equation

Modeling

1.1 What is SEM?

Structural

There is
hypothesized
underlying
structure to
nature...

Equation

Modeling

1.1 What is SEM?

Structural

There is
hypothesized
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Equation

...that can be
translated to a
series of
mathematical
equations...

Modeling

1.1 What is SEM?

Structural

There is
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Equation

...that can be
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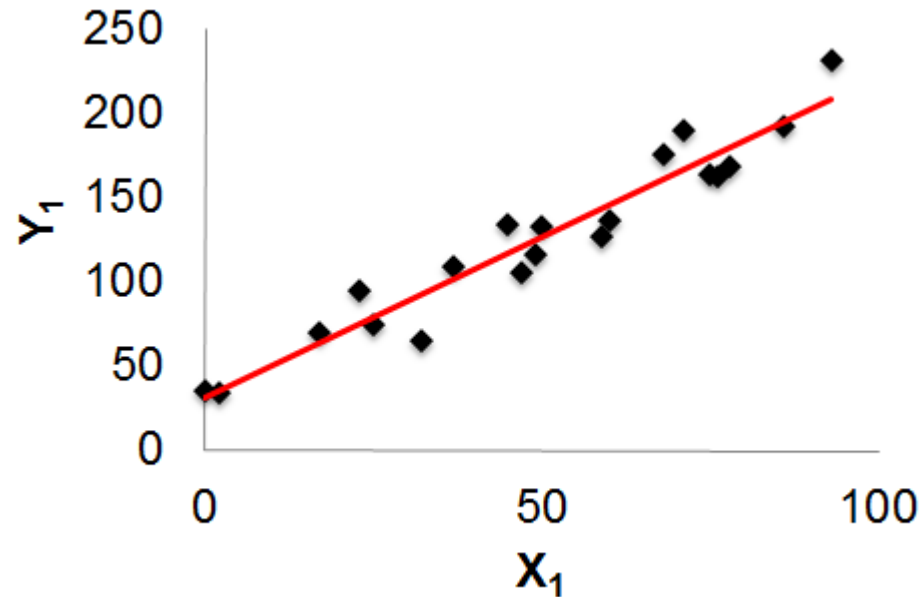
Modeling

...which can be
modeled
against data to
support or
refute the
proposed
structure

1.1 What is SEM? By any other name ...

- Structural equation modelling (SEM)
- (Confirmatory) path analysis
- Directed acyclic graph model

1.1 What is SEM? A graphical approach

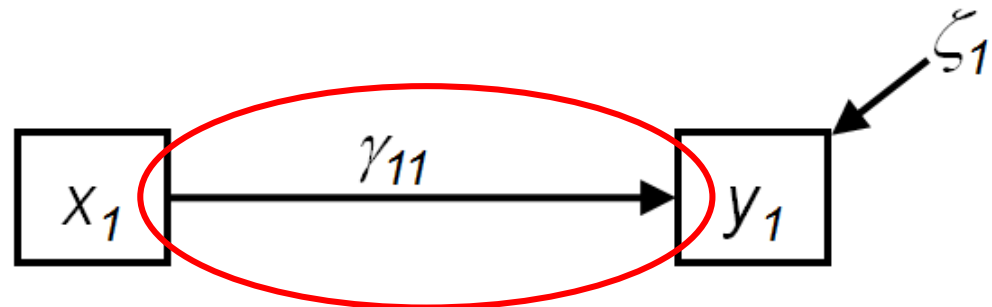


Equation form

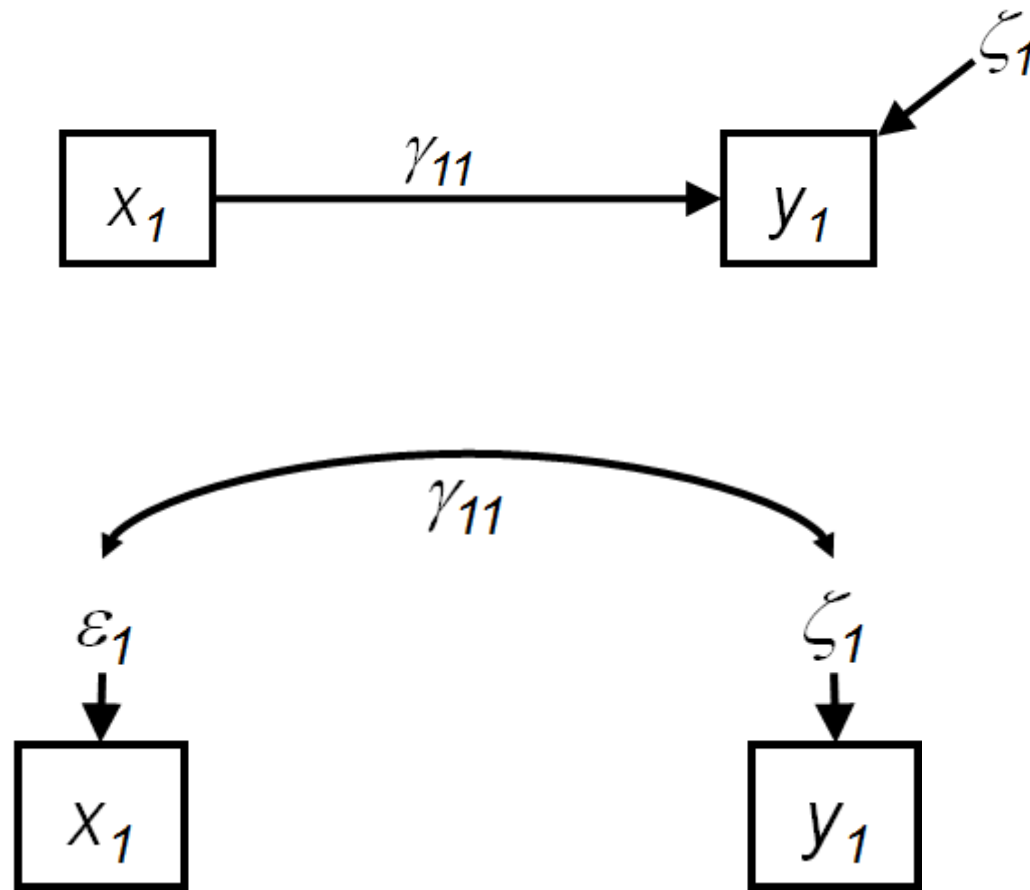
$$y_1 \sim \alpha_1 + \beta_1 x_1 + \varepsilon$$

$$y_{11} \sim \alpha_1 + \gamma_{11} x_1 + \zeta_1$$

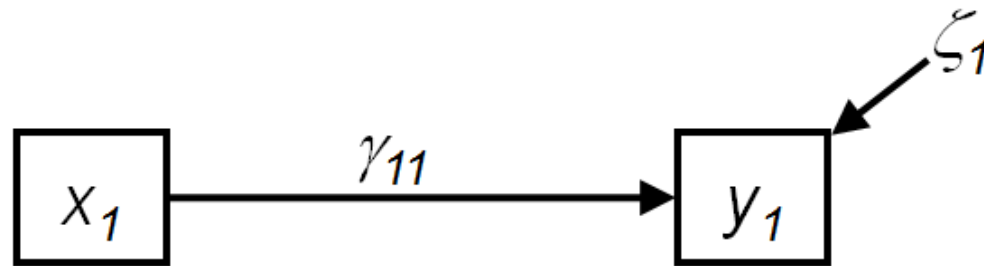
Graphical form



1.1 What is SEM? Directionality

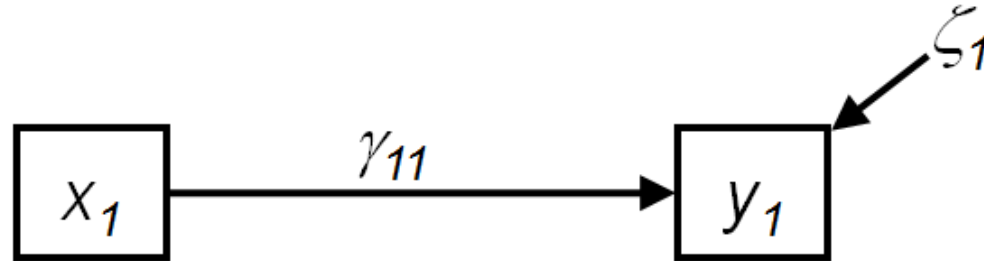


1.1 What is SEM? The elephant in the room



“An equation...can be said to be *structural* if there exists sufficient evidence from all available sources to support the interpretation that x_1 has a causal effect on y_1 .” (Grace, 2006)

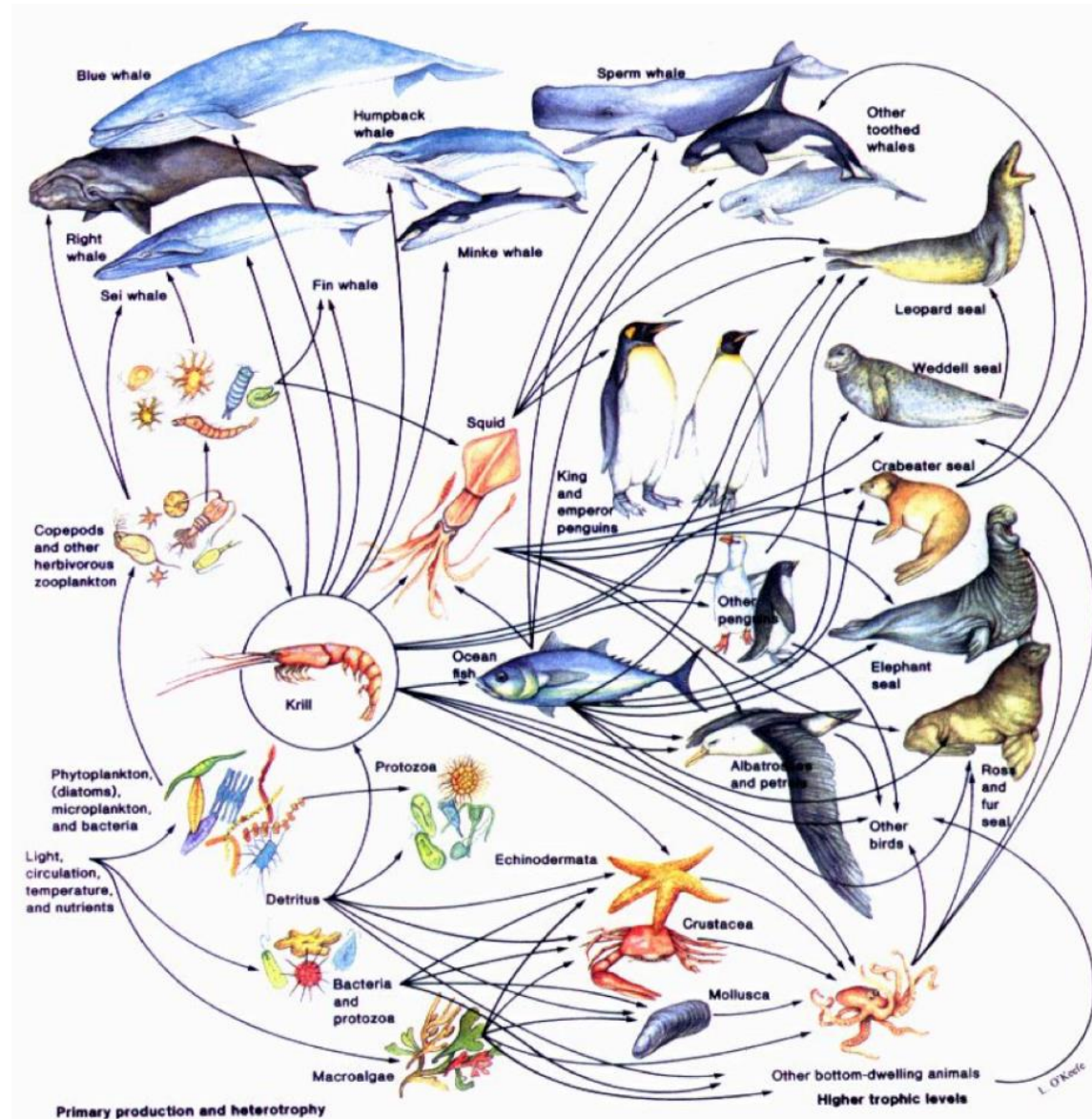
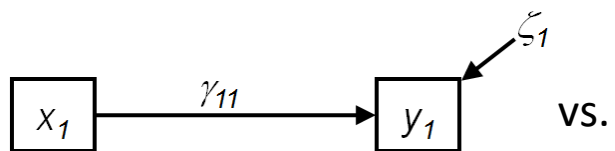
1.1 What is SEM? What is *causation*?



- **Key Point #1:** SEM assumes that x causes y
 - Prior observation (a hunch)
 - Prior statistical models
 - Prior experimentation
 - Some or all of the above
- Does not assume ultimate causation

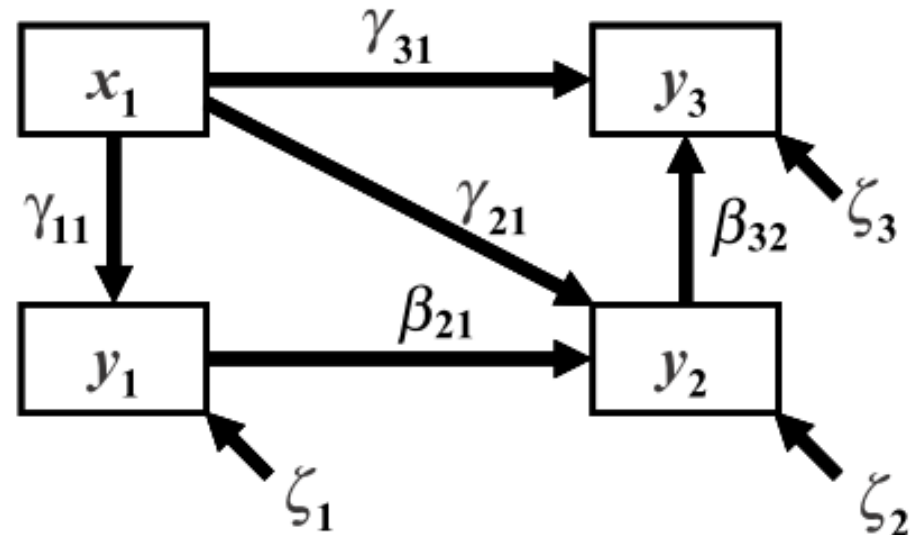


1.1 What is SEM? Nature is complex



1.1 What is SEM? Incorporating complexity

Graphical model



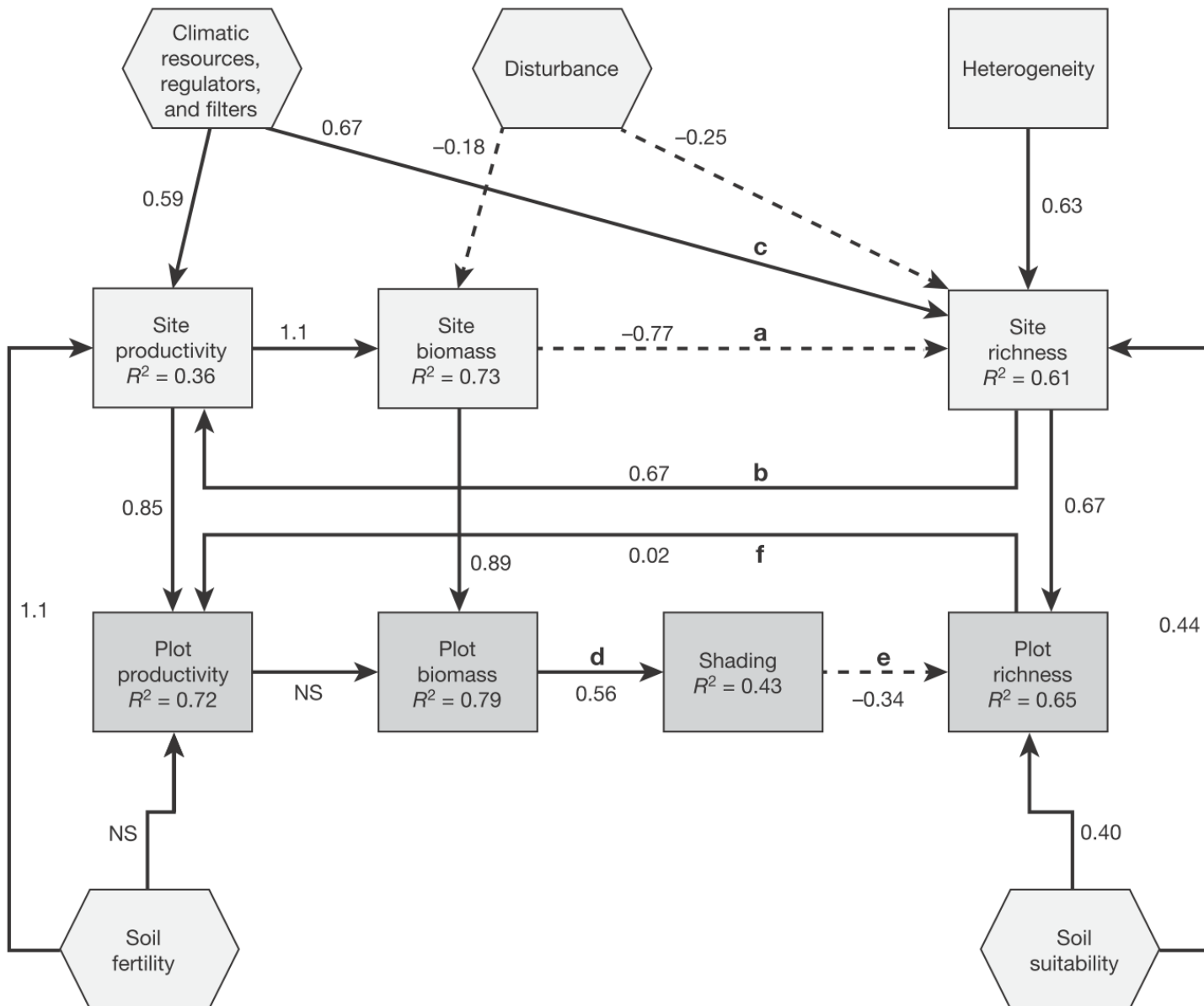
List of equations

$$y_1 = \alpha_1 + \gamma_{11}x_1 + \zeta_1$$

$$y_2 = \alpha_2 + \beta_{21}y_1 + \gamma_{21}x_1 + \zeta_2$$

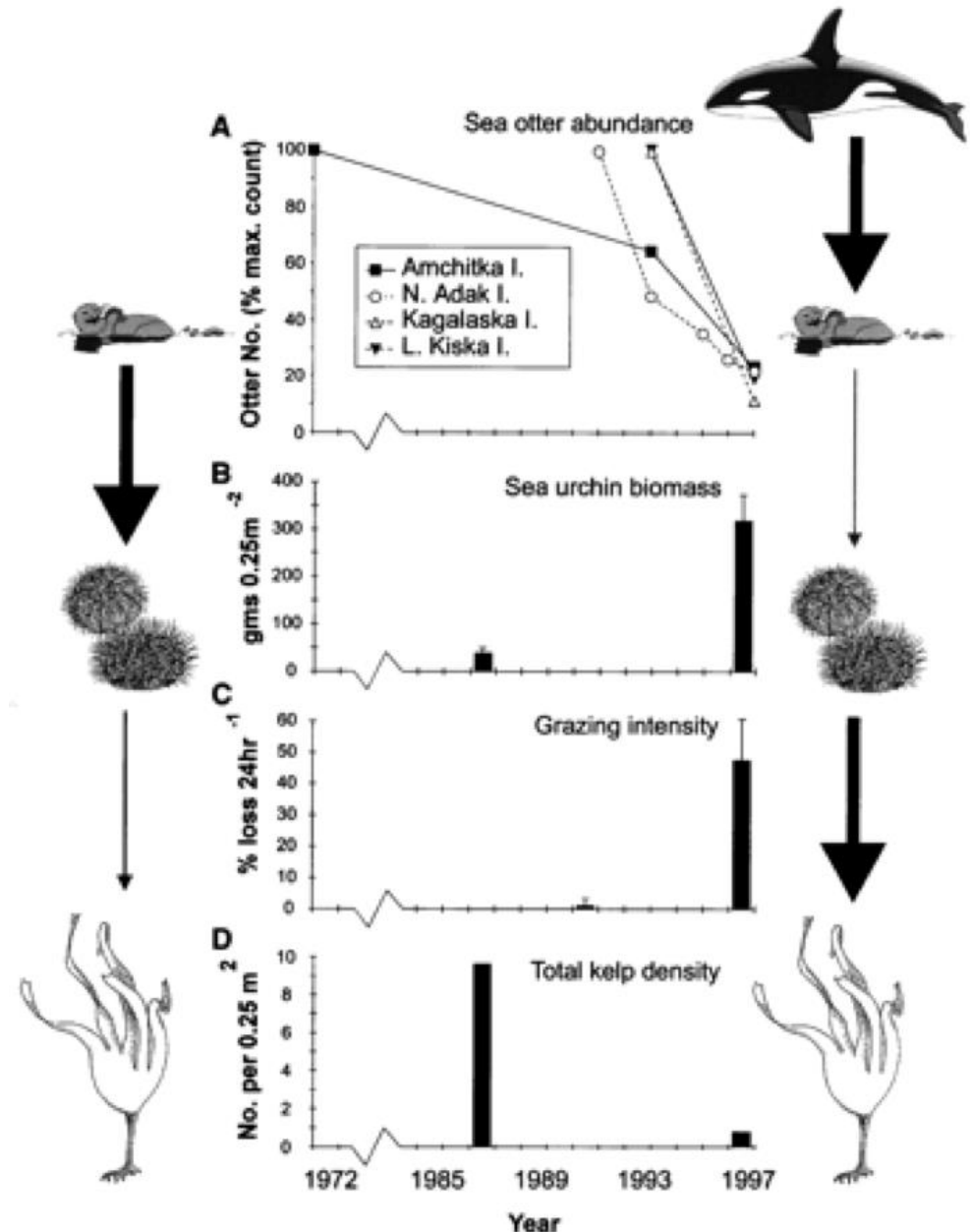
$$y_3 = \alpha_3 + \beta_{32}y_2 + \gamma_{31}x_1 + \zeta_3$$

1.1 What is SEM? Building up

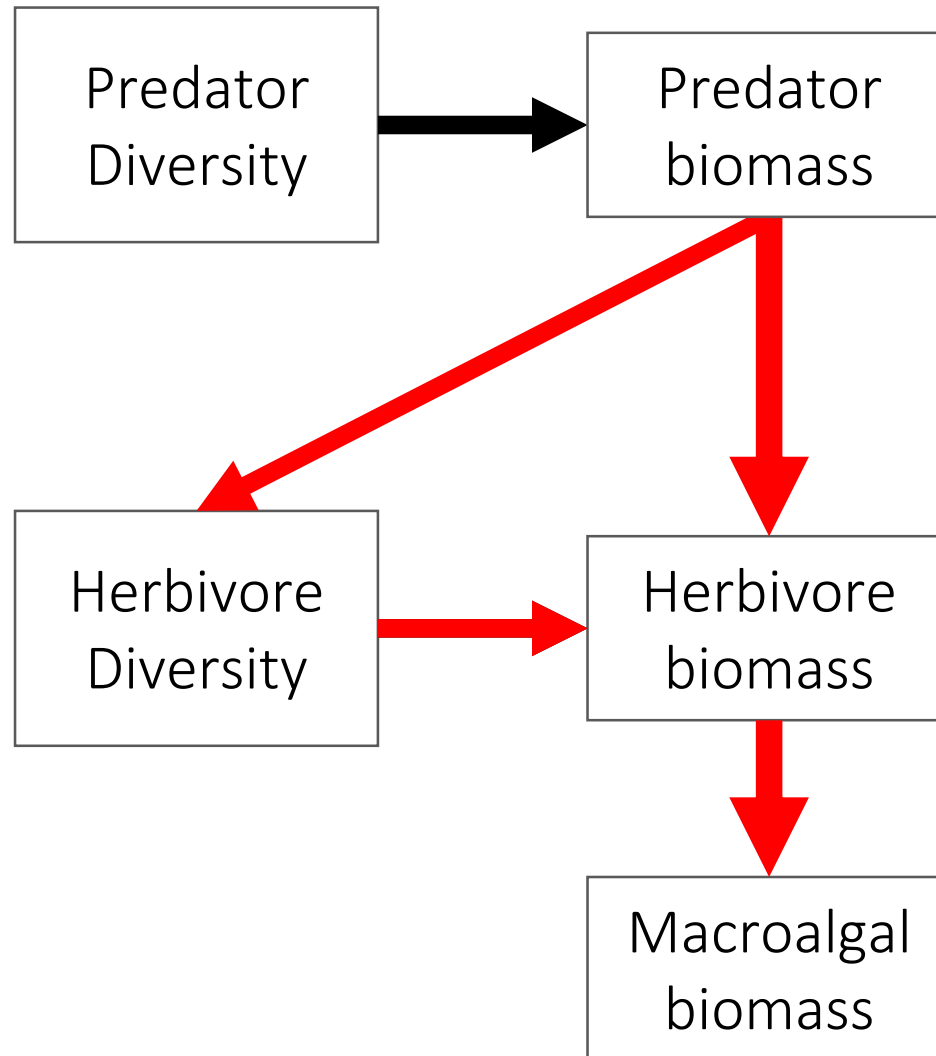


1.1 What is SEM? A complicated network

- **Key Point #2:** By combining inferences across multiple equations, SEM addresses both direct and indirect effects in a system



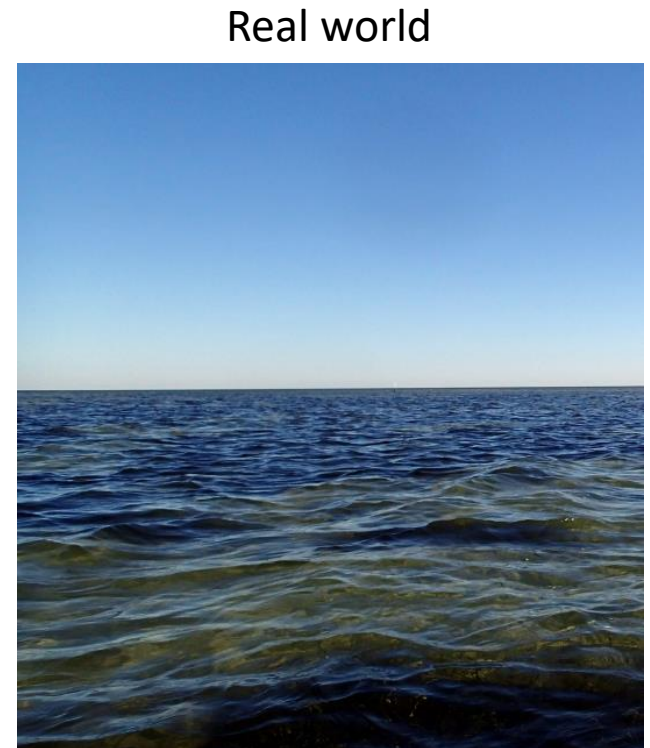
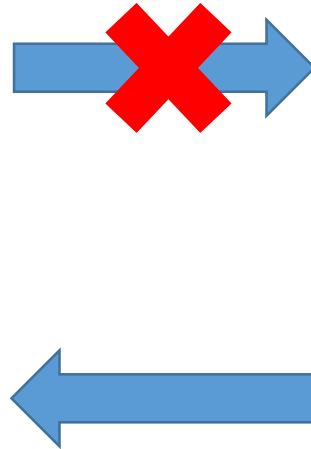
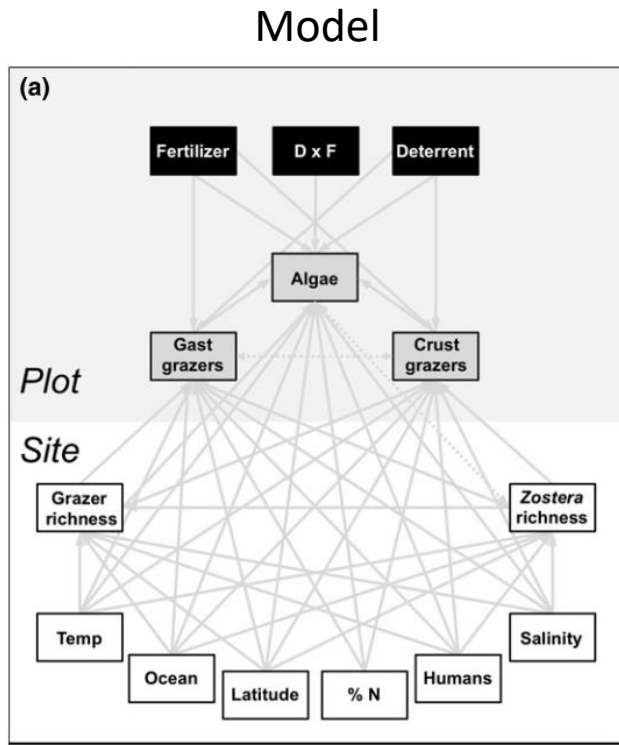
1.1 What is SEM? Indirect effects



1.1 What is SEM? Putting it all together

- **Key Point #1:** SEM assumes that x causes y
- **Key Point #2:** By combining inferences across multiple equations, SEM addresses both direct (proximate) and indirect (ultimate) effects in a system

1.1 What is SEM? Reality vs. model



Real world informs the model, not the other way around!

1.2 A History Lesson

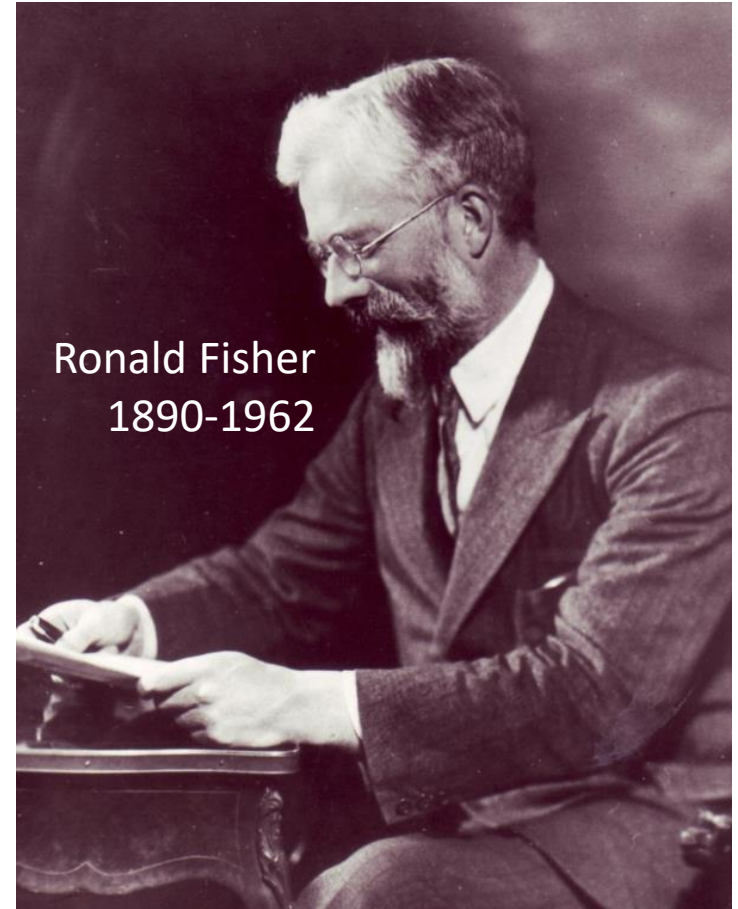
1.2 History. Time warp



Karl Pearson
1857-1936

Francis Galton
1822-1911

Pearson product moment correlation, r
Chi-squared goodness of fit, χ^2



Ronald Fisher
1890-1962

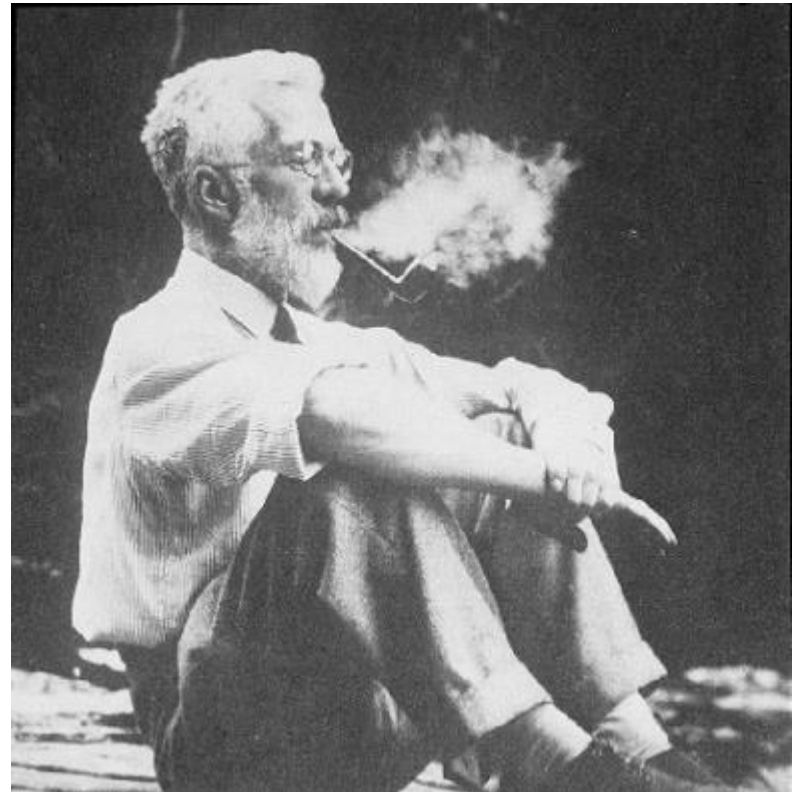
ANOVA
Maximum likelihood estimation

1.2 History. Causation vs. correlation

Smoking \rightarrow Cancer

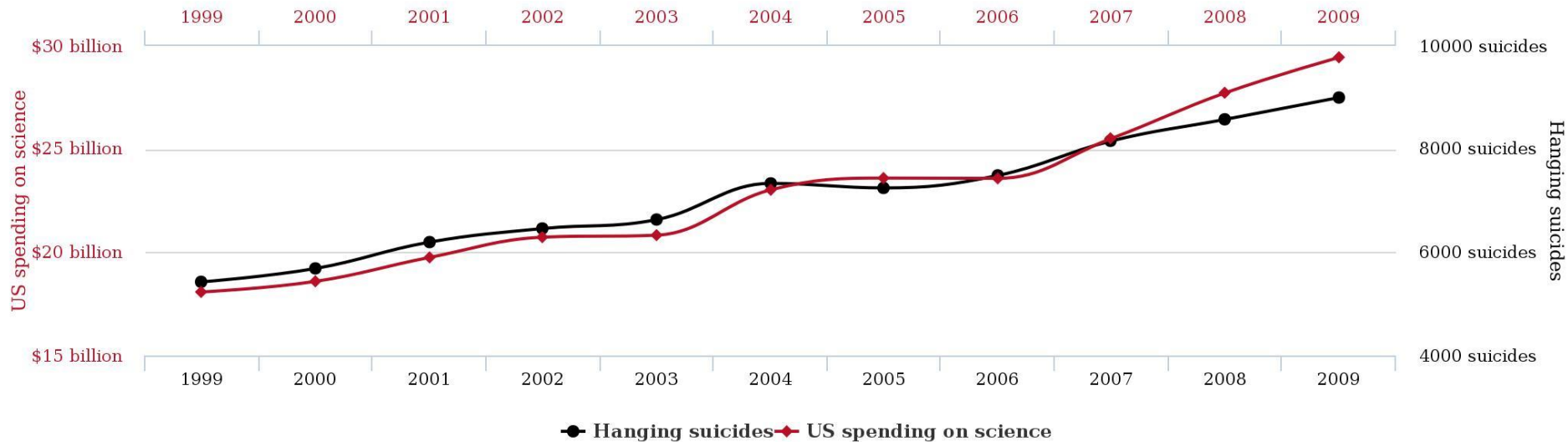
Cancer \rightarrow Smoking

Smoking \leftarrow Gene \rightarrow Cancer



1.2 History. Causation vs. correlation

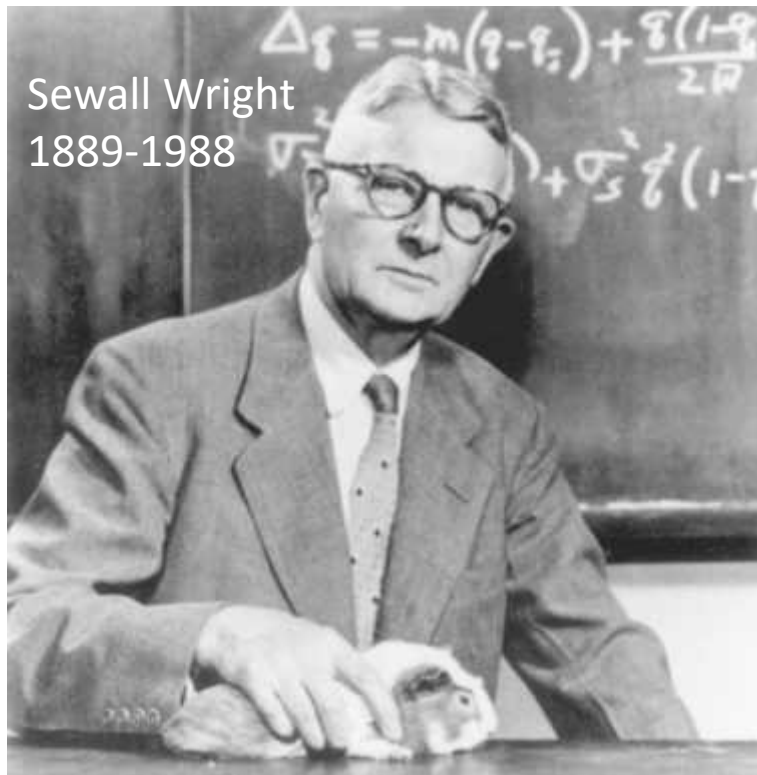
US spending on science, space, and technology correlates with Suicides by hanging, strangulation and suffocation



tylervigen.com

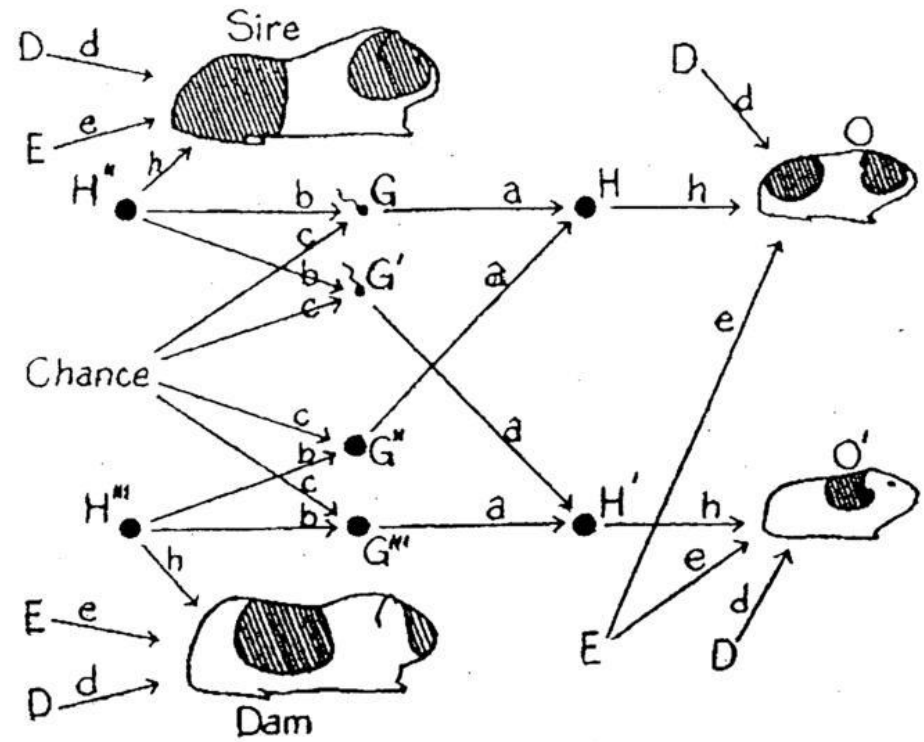
<http://www.tylervigen.com/>

1.2 History. Causation vs. correlation



Sewall Wright
1889-1988

Path analysis



1.2 History. Causation vs. correlation

“The basic fallacy of the method appears to be the assumption that it is possible to set up *a priori* a comparatively simple graphic system which will truly represent the lines of action of several variables upon each other, and upon a common result. . . . The pure mathematics by which this is shown is apparently faultless in the sense of algebraic manipulation, but it is based upon assumptions which are wholly without warrant from the standpoint of concrete, phenomenal actuality.” (Niles, 1922)

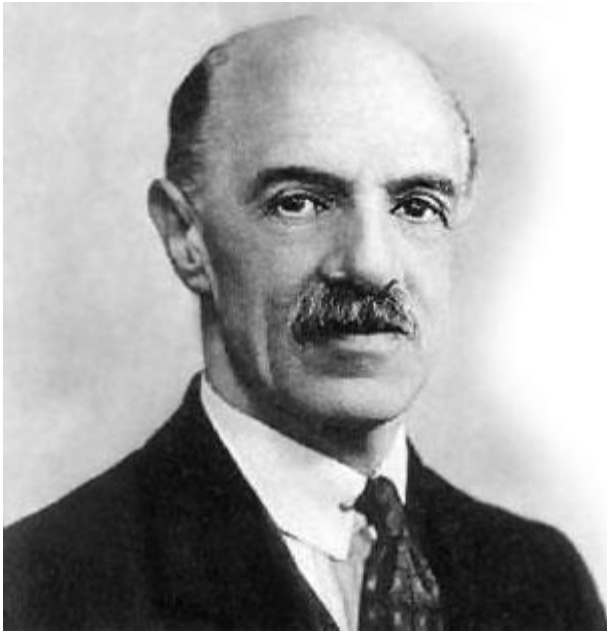
“The writer has never made the preposterous claim that the theory of path coefficients provides a general formula for the deduction of causal relations. He wishes to submit that the *combination* of knowledge of correlations with knowledge of causal relations, to obtain certain results, is a different thing from the *deduction* of causal relations from correlations implied by Niles’s statement. Prior knowledge of the causal relations is assumed as a prerequisite in the former case. Whether such knowledge is ever possible seems to be the subject of Niles’s philosophical discussion of the nature of causation.”
(Wright, 1923)

1.2 History. Causation vs. correlation

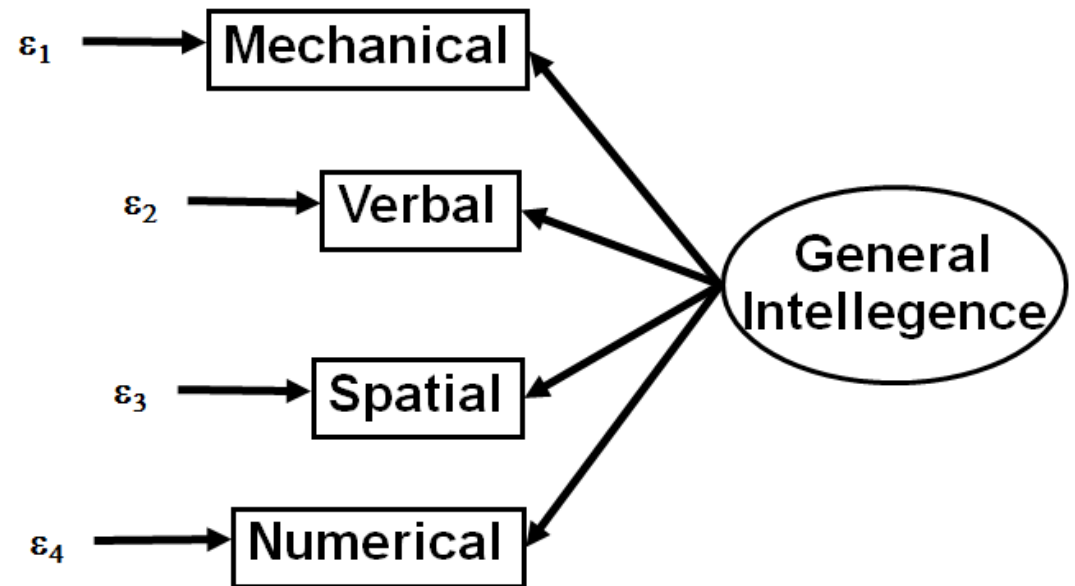
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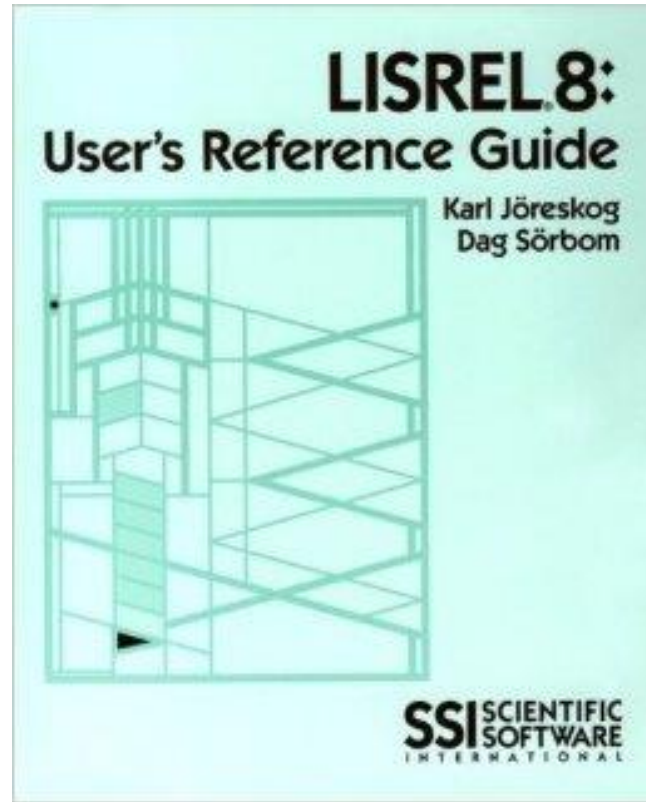
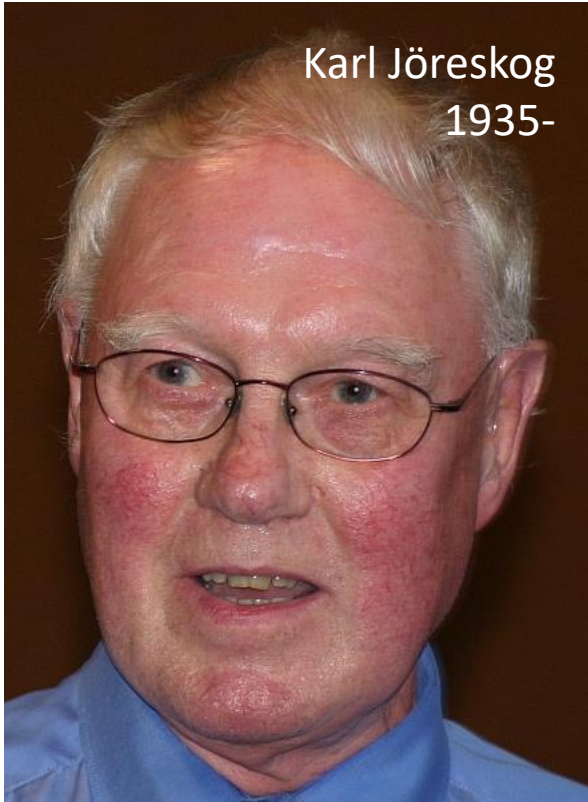
1.2 History. Factor Analysis



Charles Spearman
1863-1945

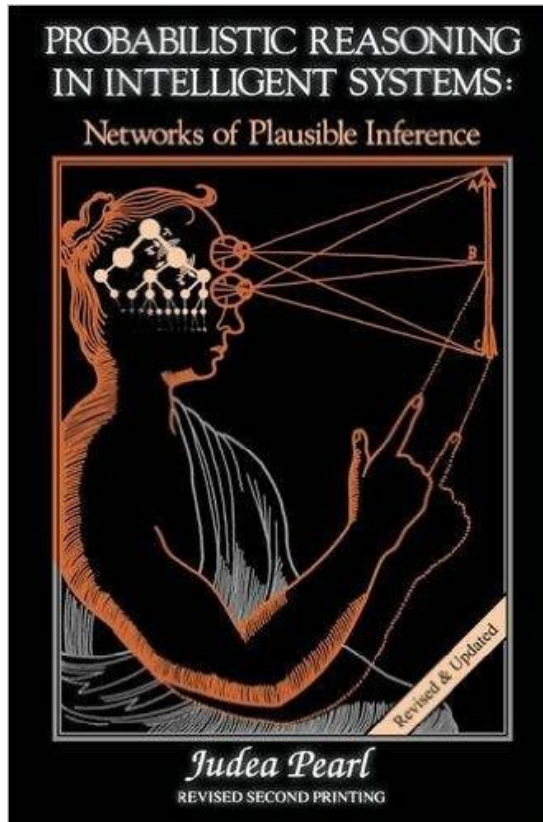


1.2 History. 2nd Generation SEM



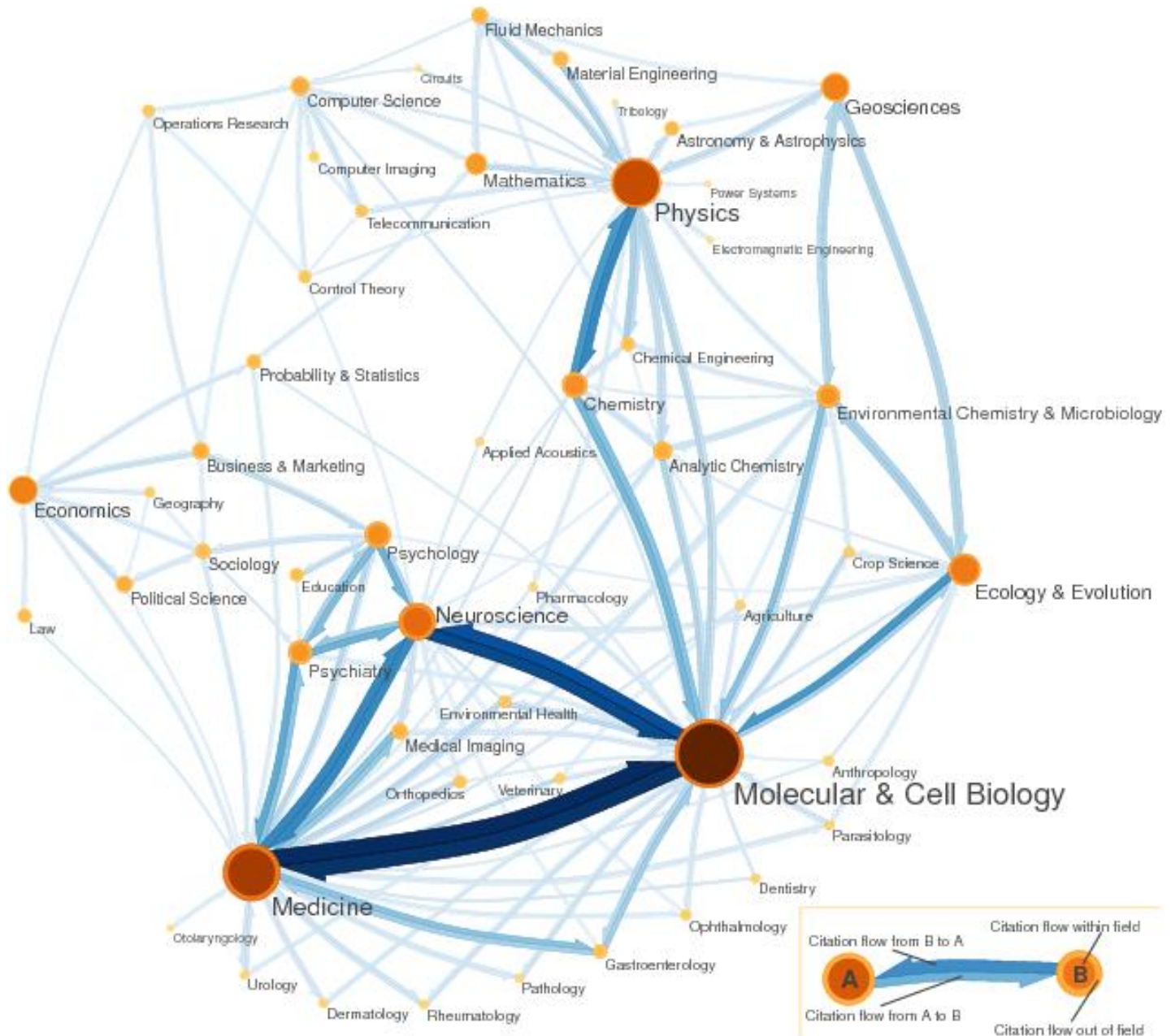
- LISREL = combine path and factor analysis
- Model fit using covariance and ML estimation
- Assess and compare fit of multivariate model

1.2 History. 3rd Generation SEM

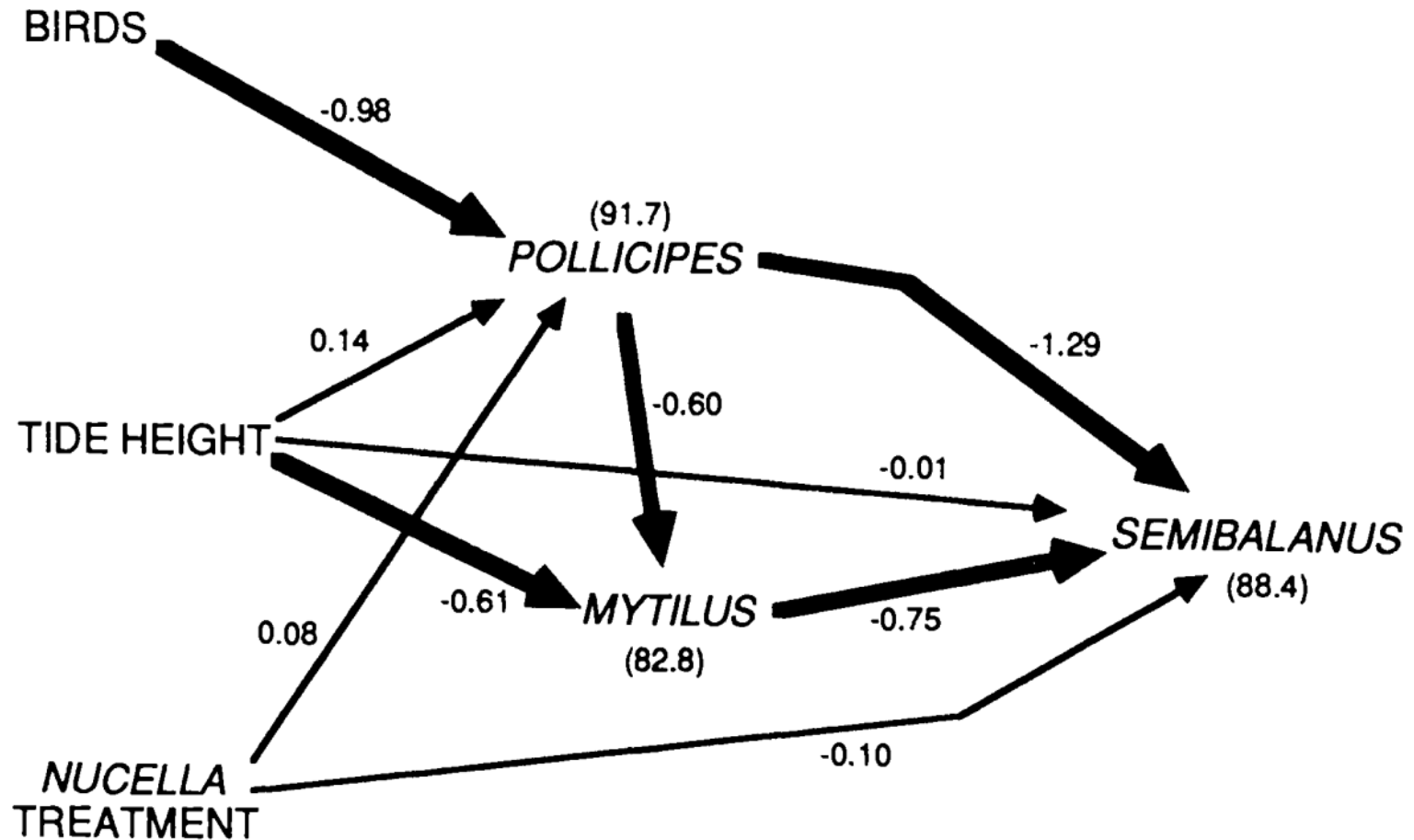


- Unite SEM with graph theory
- Causality is central
- Flexible methods with piecewise approach

1.2 History. SEM and Ecology



1.2 History. SEM and Ecology

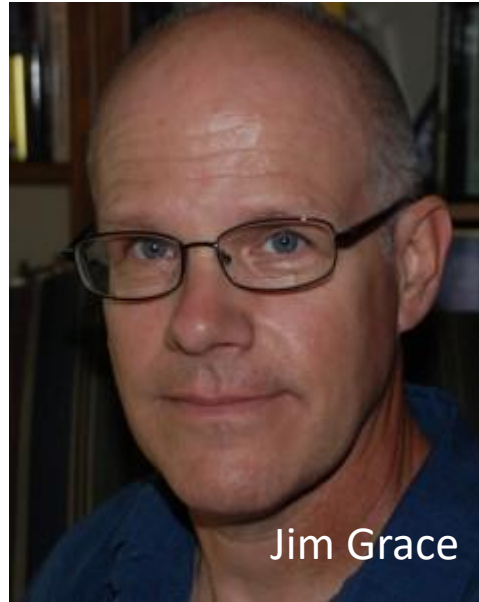


Wootton (1994) *Ecology*

1.2 History. SEM and Ecology



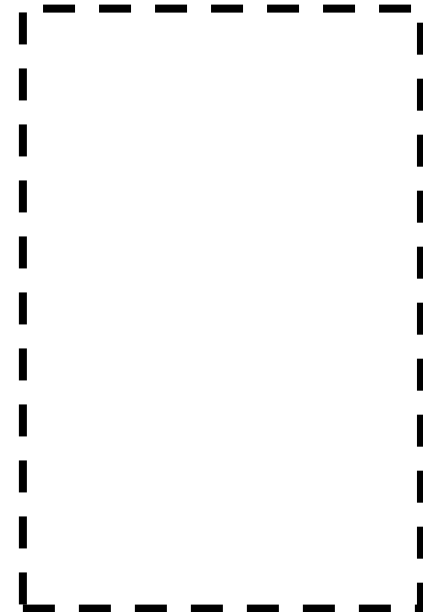
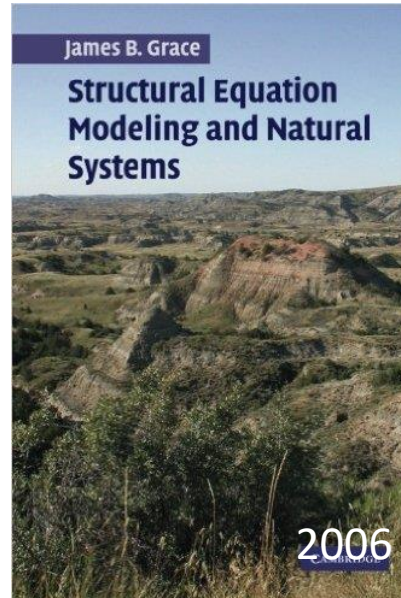
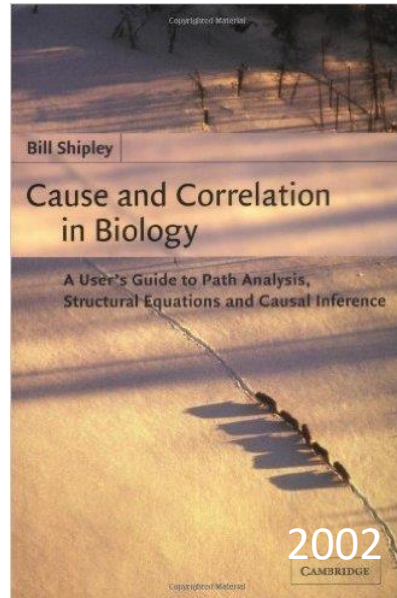
Bill Shipley



Jim Grace



Jarrett Byrnes



1.3 From ANOVA to SEM

1.3 From ANOVA to SEM. Whalen et al. 2013

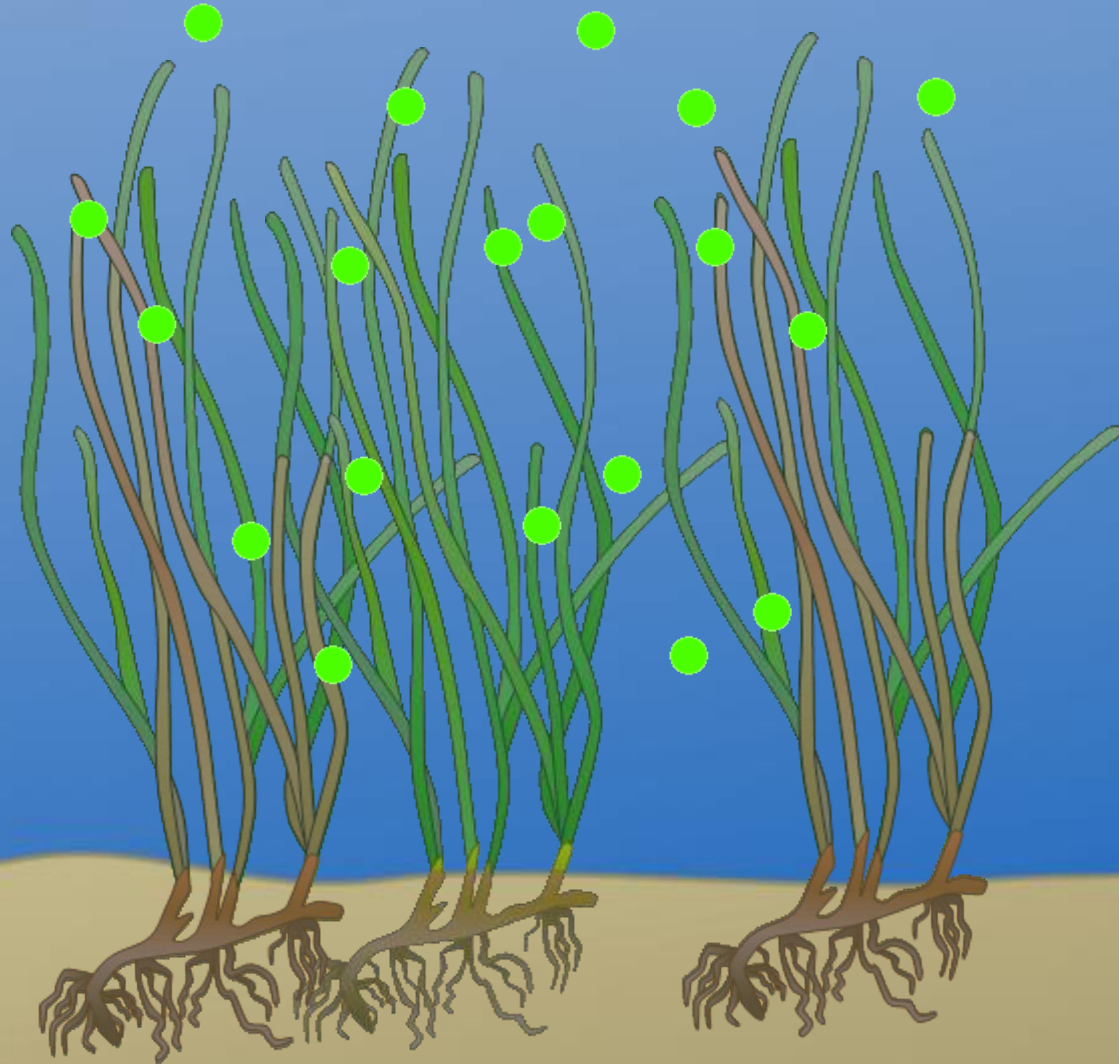
Ecology, 94(2), 2013, pp. 510–520
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Temporal shifts in top-down vs. bottom-up control of epiphytic algae in a seagrass ecosystem

MATTHEW A. WHALEN,^{1,3} J. EMMETT DUFFY,¹ AND JAMES B. GRACE²



1.3 From ANOVA to SEM. Seagrass systems

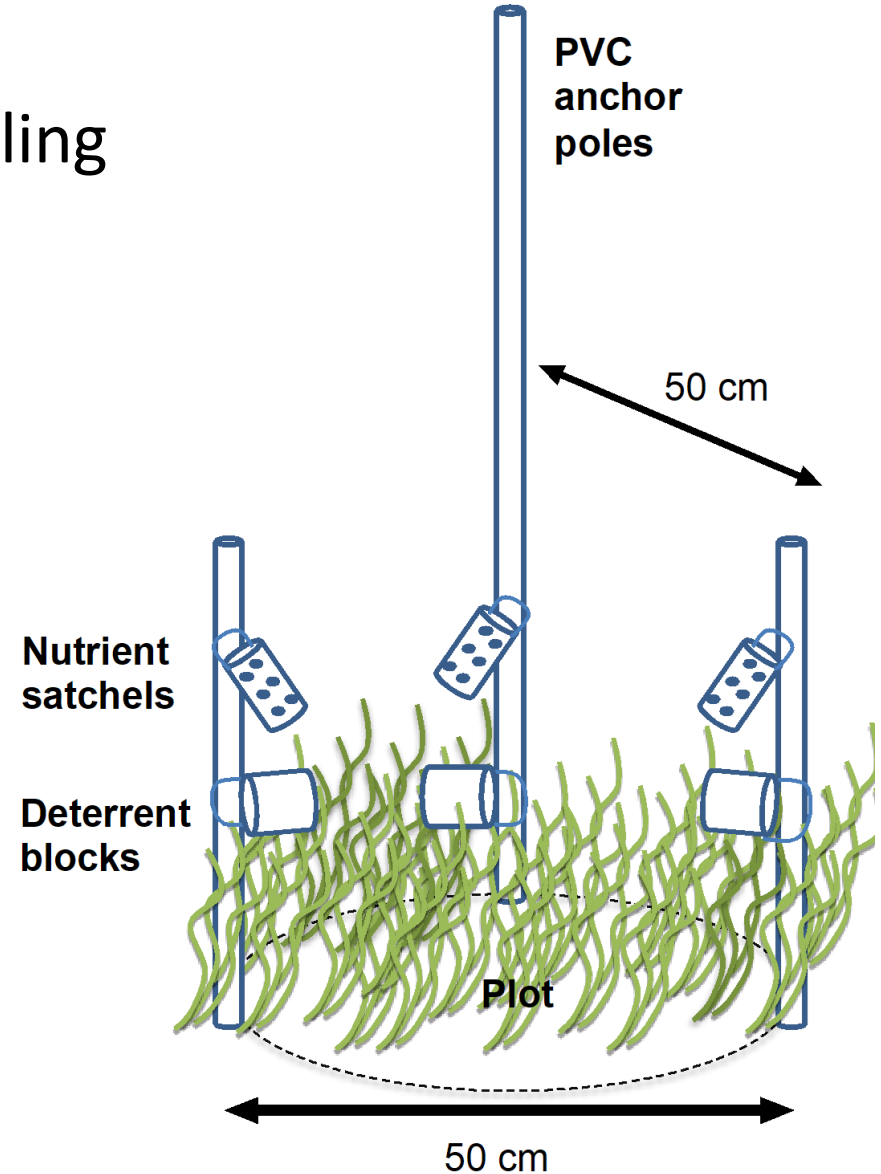
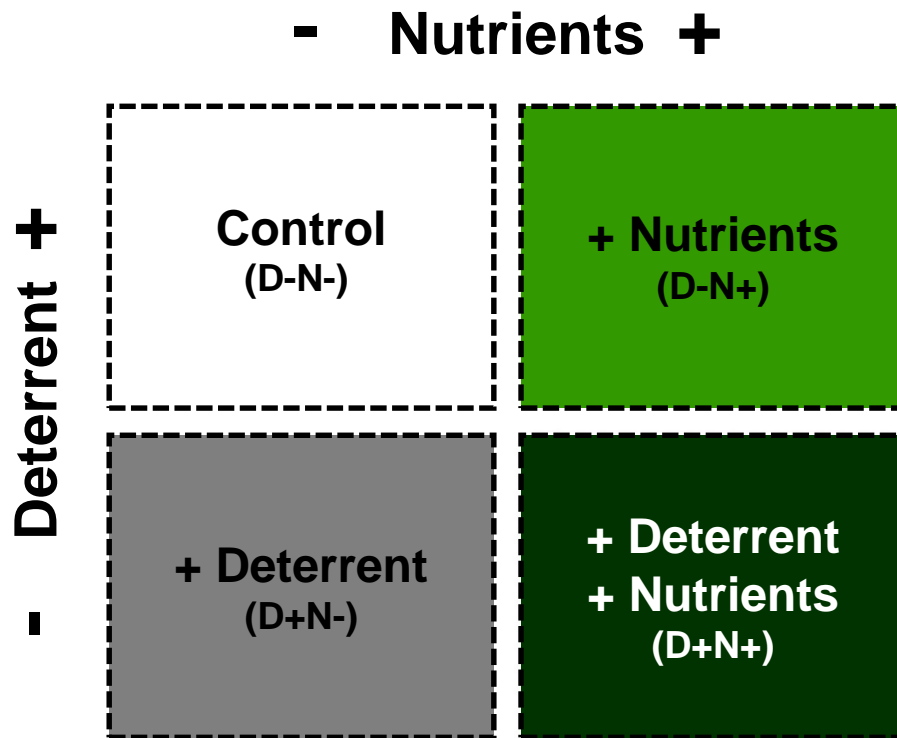






1.3 From ANOVA to SEM. Experimental Design

What are the relative influence
of top-down vs.
bottom-up control in controlling
seagrass ecosystems?



1.3 From ANOVA to SEM. Experimental Design

ZEN



1.3 From ANOVA to SEM. Whalen et al. 2013



1.3 From ANOVA to SEM. Whalen et al. 2013

TABLE 1. Univariate analyses of mesograzer densities and epiphyte biomass from (A) fall and (B) summer experiments in an eelgrass (*Zostera marina*) bed in the York River, Virginia, USA.

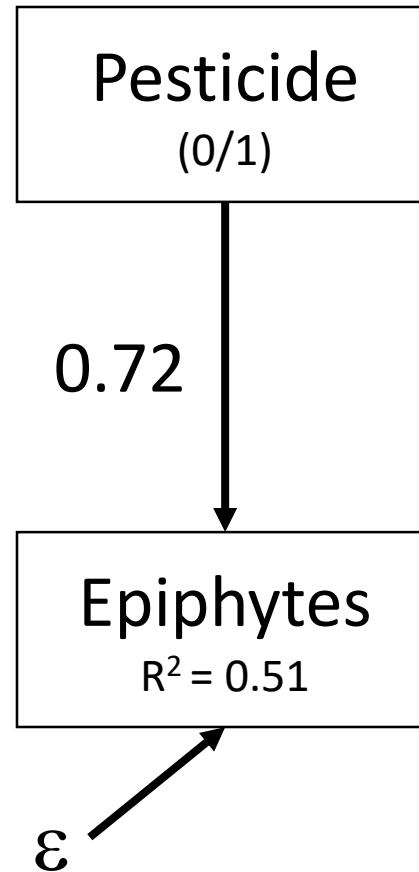
Experiment and response source	Crustaceans			Gastropods			Epiphytes		
	df	<i>F</i>	<i>P</i>	df	<i>F</i>	<i>P</i>	df	<i>F</i>	<i>P</i>
A) Fall									
Deterrent	1	42.84	<0.001	1	0.33	0.574	1	3.97	0.052
Fertilization							1	3.10	0.084
Sampling date	2	13.77	<0.001	2	0.12	0.887	1	78.24	<0.001
Det. × fert.							1	0.86	0.358
Det. × date	2	2.48	0.108	2	1.27	0.301	1	3.72	0.059
Fert. × date							1	7.00	0.011
Det. × fert. × date							1	0.81	0.371
Residual	21			21			51		
B) Summer									
Deterrent	1	129.24	<0.001	1	1.07	0.306	1	66.22	<0.001
Fertilization	1	0.00	0.958	1	0.01	0.920	1	2.19	0.145
Sampling date	1	0.89	0.349	1	11.00	0.002	1	0.83	0.367
Det. × fert.	1	0.10	0.756	1	2.00	0.163	1	1.00	0.322
Det. × date	1	0.58	0.448	1	2.96	0.091	1	6.21	0.016
Fert. × date	1	2.90	0.094	1	0.71	0.403	1	0.53	0.468
Det. × fert. × date	1	1.57	0.216	1	0.27	0.606	1	1.14	0.290
Residual	56			56			56		

DEATH BY F-TABLES!

Notes: ANOVA tables for linear models describe the effects of chemical deterrent, nutrient fertilization, and sampling date on crustacean mesograzer density, gastropod mesograzer density, and epiphyte biomass. All data were natural-log-transformed except summer gastropods (square-root transformed). Model terms were tested using *F* tests and type III sums of squares. Note that the analyses presented for the summer experiment are balanced. *P* values <0.05 are shown in boldface.

1.3 From ANOVA to SEM. ANOVA

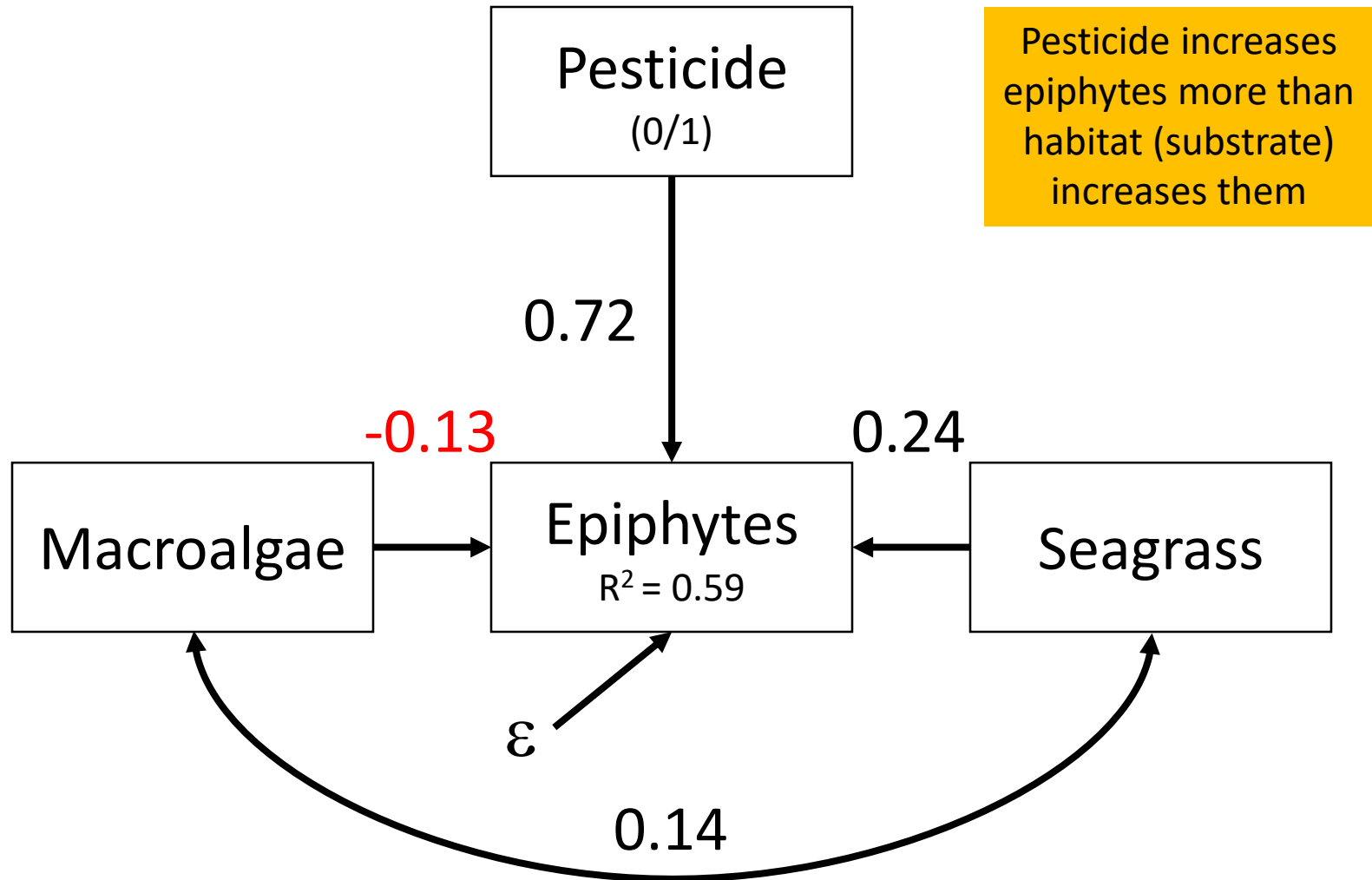
Epiphytes \sim Pesticide



Pesticide
increases
epiphytes

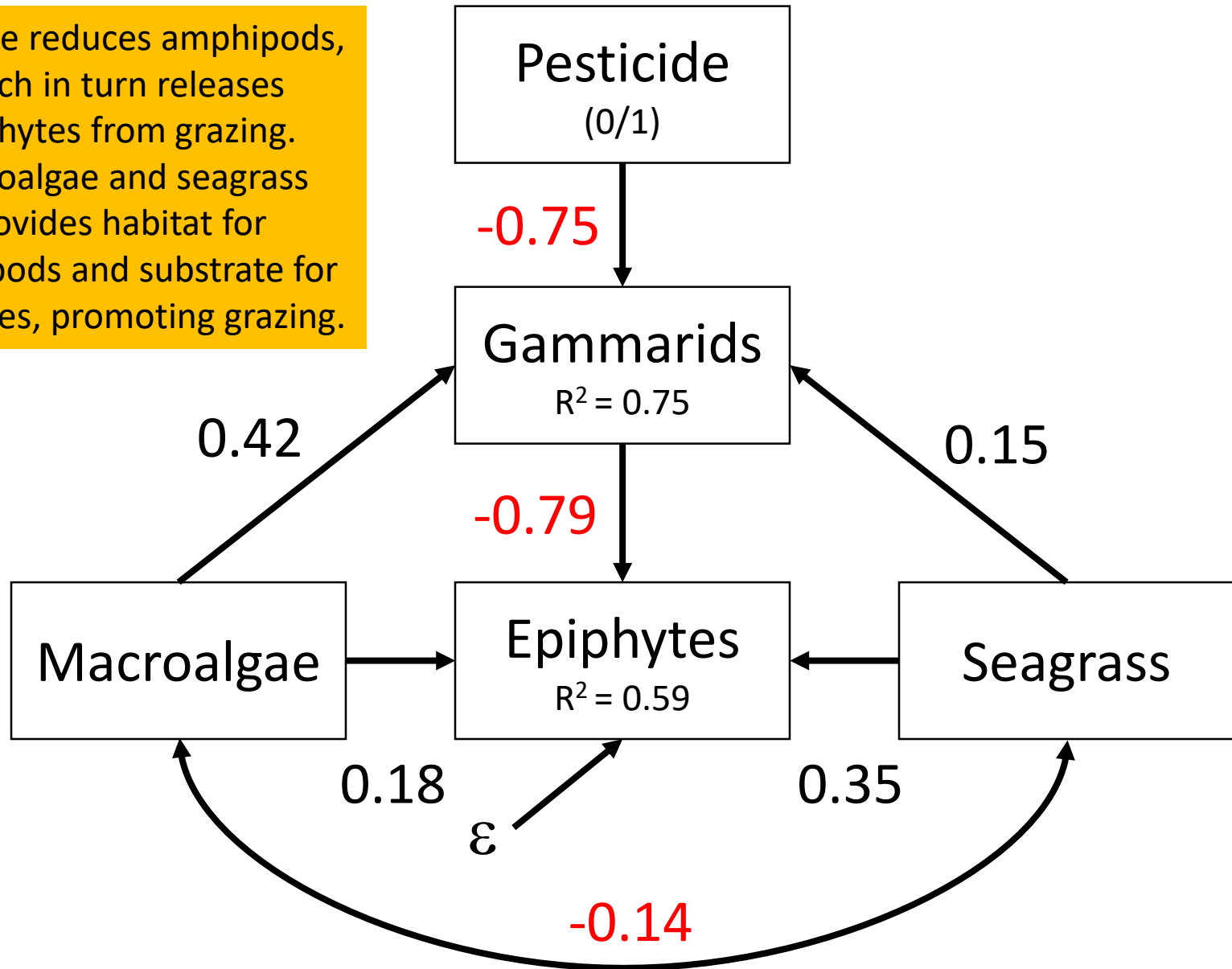
1.3 From ANOVA to SEM. ANCOVA

Epiphytes \sim Pesticide + Macroalgae + Seagrass

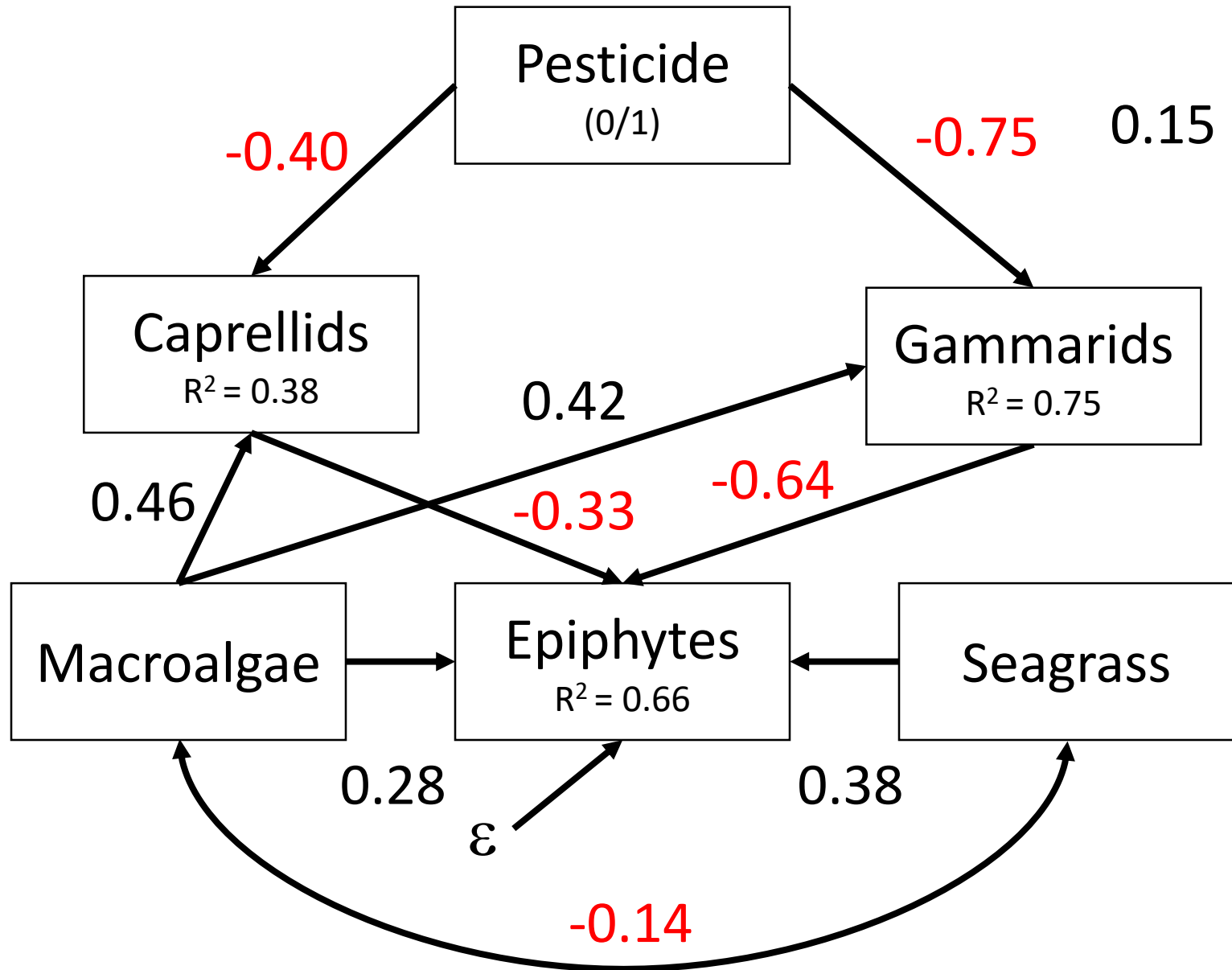


1.3 From ANOVA to SEM. Mediation

Pesticide reduces amphipods, which in turn releases epiphytes from grazing. Macroalgae and seagrass provides habitat for amphipods and substrate for epiphytes, promoting grazing.



1.3 From ANOVA to SEM. Mediation x2



1.3 From ANOVA to SEM. Increasing inference

Pesticide
reduces
epiphytes

ANOVA

Pesticide increases
epiphytes more than
habitat (substrate)
increases them

ANCOVA

Pesticide reduces
amphipods, which in
turn releases epiphytes
from grazing.
Macroalgae and
seagrass provides
habitat for amphipods
and substrate for
epiphytes, promoting
grazing.

SEM

Pesticide reduces
gammarid and caprellid
amphipods, which in
turn releases epiphytes
from grazing.
Gammarids appear to
control epiphytes in this
system. Macroalgae
provides habitat for
amphipods, promoting
grazing, while eelgrass
provides substrate for
epiphytes.

Full SEM

1.3 From ANOVA to SEM. Increasing inference

- Teases out *complex relationships*
- Identification and comparison of *direct vs. indirect effects* & potential mediators
- Precise *mechanistic* explanations

