

Thesis proposal

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The proposal for my thesis, *Simulation models for predictive ecology*

1 Within the last several hundred years, human activity has rapidly reshaped both Earth's surface. These
2 changes can roughly be divided into two categories: (1) Land-use change, which has rapidly reshaped our
3 planet's surface and restructured habitats for every species, and (2) climate change, words here, as a result
4 of greenhouse gas emissions. As a result *ecological forecasting* (**dietze?**), or modeling how ecological
5 systems will change over time, has as an imperative to mitigate the effect of these changes on Earth's
6 ecosystems, their functioning, and the services they provide to humans.

7 An oft applied definition of the origin of ecology is "the application of the scientific method to natural
8 history." Since its origin ecology has been a descriptive science. This is a natural biproduct of the immense
9 variability of Earth's biosphere.

10 In recent years, there has been an interest in an epistemological shift in ecology. emerged to explain
11 particular phenomena at particular scales.

12 To shift ecology into a predictive science. The justification for this shift is twofold: (1) bogged down
13 philosophy of science, by further rooting our understanding of ecosystem function and dynamics in an
14 ability to predict their structure (**PredictiveEcology?**). and (2) the practical need for models for *ecological*
15 *forecasting*.

16 This term implicitly creates an analogy between predicting how ecosystems will change in the future and
17 weather forecasting. Use of computational methods in NWP. Much as one would not aim to forecast the
18 weather to Quebec by applying Navier-Stokes. NWP has worked because it incorporates information
19 about data and meteorological processes collected at difference scales into models that.

20
21 Historically the term "theory," as applied in the physical sciences, refers to mathematical models. Many
22 (although not all) of these models refers to a equation describing how the value of an observable state of

23 the system, $\begin{bmatrix} x_1 \\ x_2 \\ \vdots \end{bmatrix} = \vec{x}$, changes as a function of time.

24 differential equations (in continuous time, difference equations for discrete time)

25 Ecological processes vary across more variables than the tools of analytic models are suited for. As the
26 number of variables in an analytic model increases, so does the ability of the scientist to discern clear
27 relationships between them. Chaotic dynamics emerge from simple analytic models, and .

28 Whether ecosystems actually exhibit chaotic behavior is a different question.

29 Until the 20th century, no theory of the gravitational dynamics of more than 2 bodies. Understanding the
30 gravitational dynamics of more than two planets with any reliability proved difficult. Using the same
31 models (diffeqs), how could we adequately predict ecosystems?

32

33 Transition to theme of optimization given unknown information. A forecast gives us a range of future
34 values with uncertainty around them. Further a convenient property that a forecasting model's
35 uncertainty goes up over time (if we assume the underlying process is Markov—this is a strong assumption
36 but oft true of the models we fit to temporal data)

37 In face of uncertainty, decision making is an optimization problem. We have some goal state for the
38 future, and some estimate of what the state of the world will be given a set of actions. Frame optimization
39 problem mathematically an introduce concept of solution-space and constraint.

40 Indeed Marx's most well known quote that "philosophers have hitherto only interpreted the world in
41 various ways; the point is to change it."

42 and a necessary step toward establishing a just and sustainable world.

43 Transition to specifics of this thesis.

44 [Figure 1 about here.]

45 **CH1 Forecasting the phenological uncoupling of a plant-pollinator** 46 **network**

47 This chapter uses several years of data on bee-flower phenology and interactions, combined with spatial
48 records of species occurrence via GBIF, to predict the probability of each realized interaction network as a
49 function of location and time.

50 **Data**

51 **Methods**

- 52 • simulate species distribution and efficacy of detection given a set of observation points where the
53 dist from observation site decays.
- 54 • optimize set of repeated sampling locations L for a *known* distribution D .
- 55 • address SDM not being the territory

56 **Preliminary Results**

57 Transition to next chapter by discussing uncertainty in interaction prediction across space.

58 **CH2 optimizing sampling of interactions**

59 This chapter discusses the effect of species relative abundance on samples of interaction data, and
60 proposes a method for optimizing spatial sampling of a possible interaction between species as a function
61 of the estimated distribution of both species.

62 **Methods**

- 63 • the missing link paper, turn this into optimizing with two different SDMs

64 **Results**

65 **CH3 optimizing corridor placement**

66 This chapter proposes an algorithm for optimizing (corridorplacement/restoration effort) given a raster
67 where each cell indicates land-cover. The optimization method uses the result of a simulated process
68 (specifically occupancy dynamics in the landscape) and uses simulated annealing to estimate the global
69 optimum of the targetstate (specifically mean-time-to-extinction for the occupancy dynamics example).

70 **Methods**

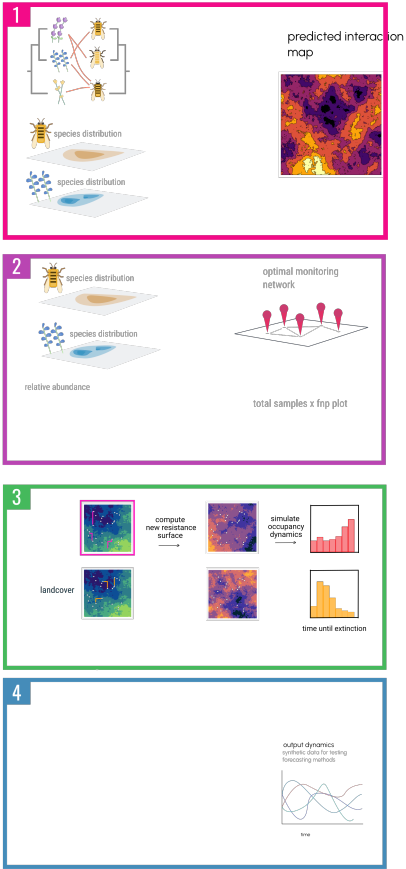
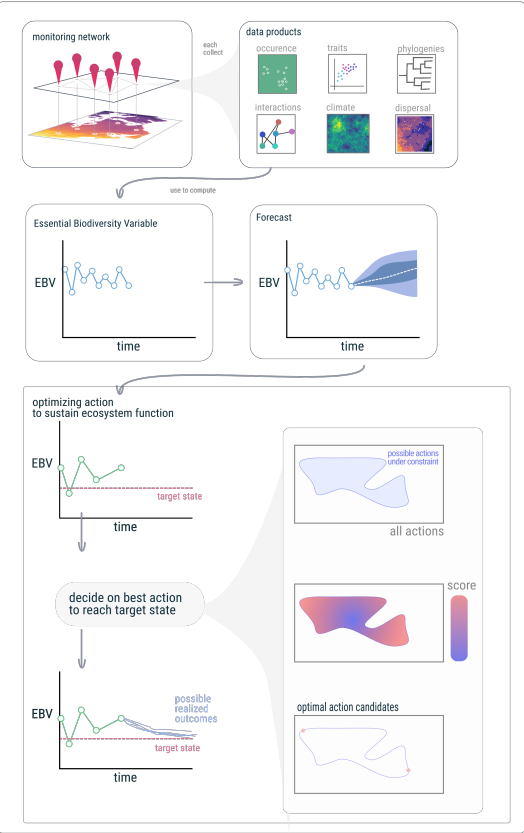
- 71 • land cover -> resistance -> extinction time
- 72 • simulated annealing to optimize landscape optimization

73 **CH4 a software note on the resulting packages.**

74 (MetacommunityDynamics.jl: a virtual laboratory for community ecology): a collection of modules in the
75 Julia language for different aspects of metacommunity ecology, including most of the code used for the
76 preceding chapters.

- 77 • TK conceptual figure with interfaces between what I'm writing / have contributed to and linked
78 with other libraries
- 79 • Observatories.jl, Corridors.jl, MCD.jl

The flow of data to monitor ecosystems and mitigate anthropogenic change, and maintain ecosystem function



chapter one
forecasting the spatiotemporal uncoupling of a plant-pollinator network

chapter two
optimizing sampling of interactions

chapter three
optimizing corridor placement against ecological dynamics

chapter four
MetacommunityDynamics.jl: a virtual laboratory for community ecology

Figure 1: thesis concept