

# **Expanding the Dynamic Shift Detector Algorithm to** Monitor Insect Populations Through Time



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Results: Simulations

#### Introduction

- Insect populations are dynamic through time & regulated by biotic & abiotic constrains
- Many ways to model pop. dynamics
- Environmentally responsive & density dependent parameters, r (intrinsic growth rate) & k (carrying capacity), control pop. dynamics
- Understanding population fluctuations & identifying drivers of fluctuations key to insect conservation
- Dynamic Shift Detector (DSD) is an algorithmic tool designed to model shifts in parameters in timeseries abundance data
- Original DSD uses one population model, the Ricker Model, which may not be optimal for all situations
- Here we test different pop models in the DSD algorithm & evaluate their performance using simulations & case studies

### Methods

- Simulations: Simulate data with 3 different population models & underlying conditions
- Vary noise, starting r, change r, change k, & series length

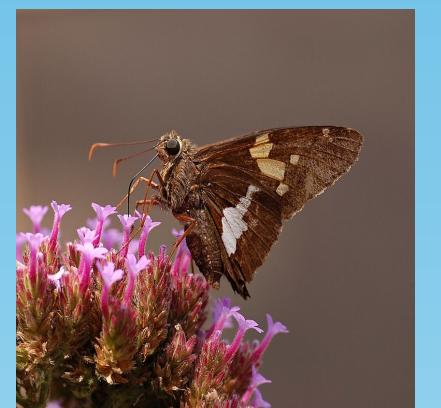
 $N_{t+1} = N_t \exp(r(1 - (N_t/k))) + \varepsilon_t$ 

 $N_{t+1} = N_t \exp(r)/(1 + \exp(r) ((1/k) N_t) + \varepsilon_t$ **Beverton-Holt Model** 

 $N_{t+1} = (N_t(r))(1 - (N_t/k)) + \varepsilon_t$ Logistic Model

r: intrinsic growth rate k: carrying capacity

- Case Study: Long-term butterfly monitoring program in Cuyahoga Valley National Park
- Applied DSD with different pop. models to 4 commonly observed species in the dataset



Ricker Model



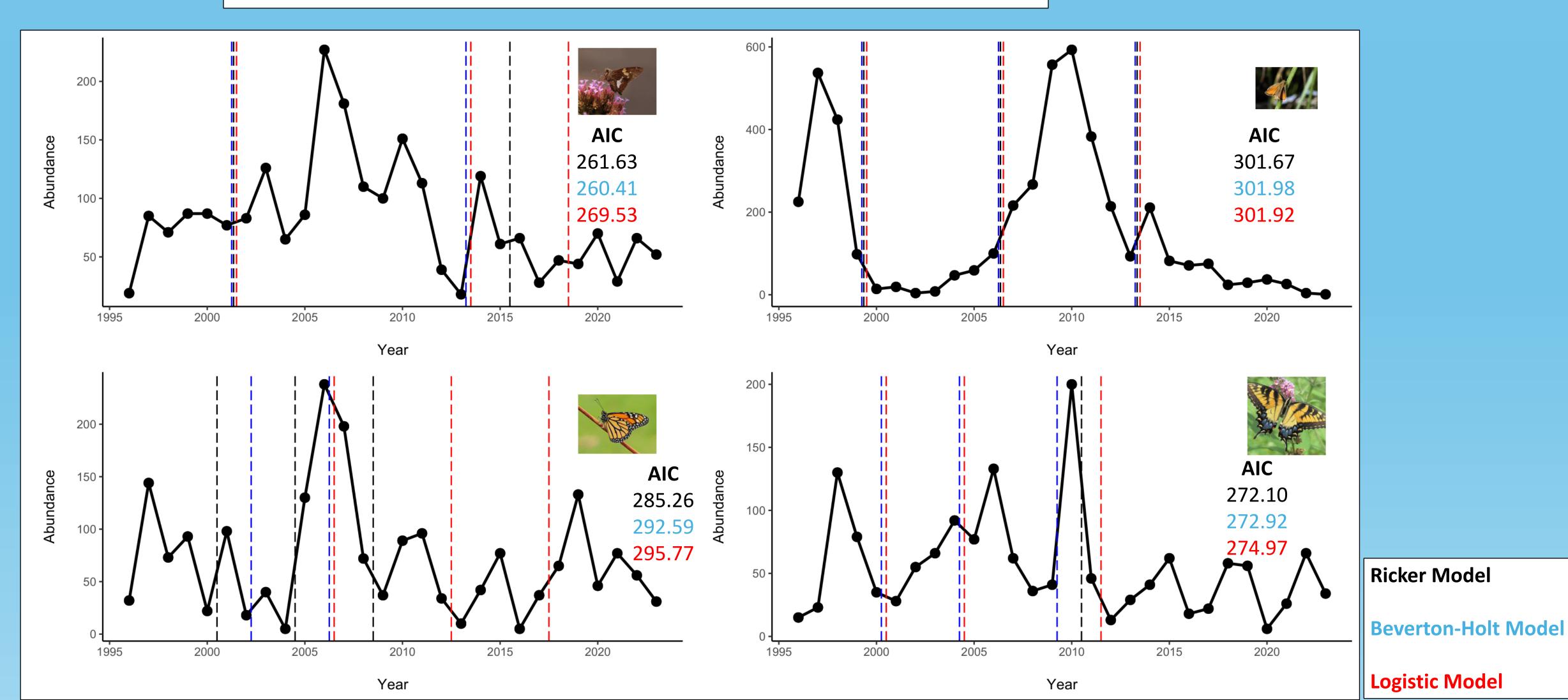


# Logistic breaks **Beverton-Holt**

## Results: Case Study

% change in k

% change in r



#### • Discussion

- Simulations & case studies indicate logistic model performs worst out of tested models
- Simplified model of population growth tends to crash to 0, which is likely impacting DSD performance
- Simulations & Case studies indicate Ricker & Beverton-Holt models produce comparable results
- Case study suggests environmental drivers of population change
  - Break points align approximately with years with high degree days & low daily precipitation

#### **Future Directions**

Test more population models

**Number of** 

- Integrate all population models into single Dynamic Shift Detector Function
- Incorporate environmental variables to see what may be driving observed shifts
- May need to be a different algorithm
- Investigate how DSD recovers underlying population via AIC

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**Monarch Butterfly** 

Tiger Swallowtail