

Nonlinear Forecasts of the Gulf of Maine Zooplankton Ecosystem

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Outline

- Modeling Calanus Abundance
- Nonlinear Models and Systems
- Application: Lorenz System
- Forecasts and Intervals
- Future Work

Nonlinear Time Series Analysis of Gulf of Maine Calanus Abundance

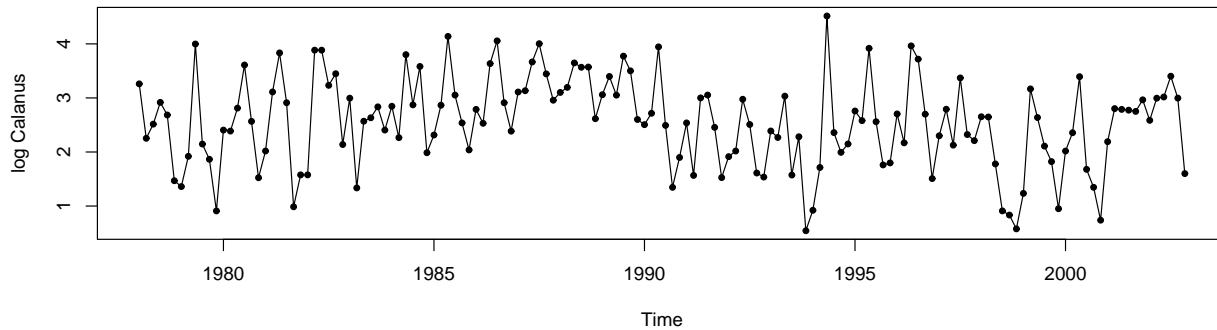
CAFE: Climate-based Assessment of Forecasting for Ecosystems in the Gulf of Maine

Goal: Understand the connection between variability and marine populations and apply this knowledge to develop ecological forecasts.

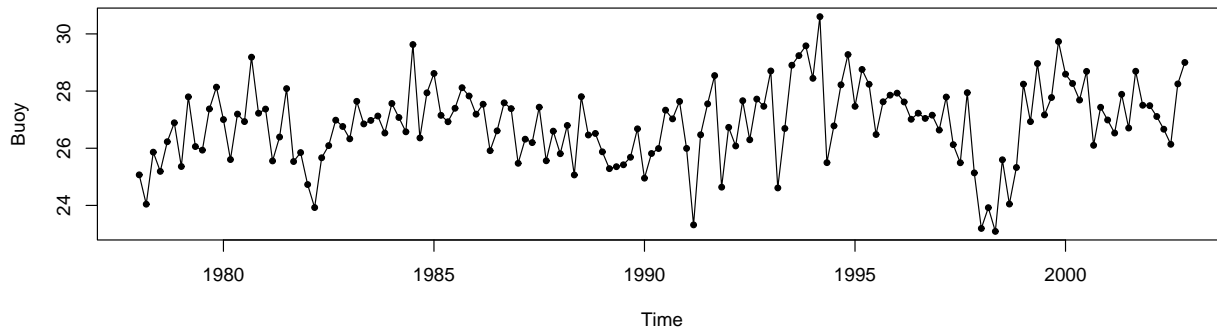
- *Calanus finmarchicus* (C) - log Calanus abundance as measured by the Continuous Plankton Recorder (CPR) surveys by NOAA Fisheries. (The major prey species of the endangered right whale.)
- Coupled Slope Water System (CSWS) - describes the system of currents and water masses between the NW Atlantic Shelf and the Gulf Stream.
- Regional Slope Water Index (RSWT) is a measure of the state of the CSWS. It is the first component of the PCA of the yearly mean temperature in 8 regions of the Gulf of Maine, Scotian Shelf, and adjacent slope waters between 150-200m.
- New Buoy Data: RSWT proxy data from Northeast Channel (near GoMOOS mooring site).
- New Herring Data: mean number per tow

Bimonthly Data: Calanus Abundance and Buoy Data 1978-2002

Calanus Time Series



Buoy (RSWT Proxy) Time Series



Nonlinear Models and Nonlinear Autoregressive Processes

- A nonlinear autoregressive process is a univariate time series:

$$N_t = f(N_{t-1}, N_{t-2}, \dots, N_{t-d}; \boldsymbol{\theta}) + e_t$$

where $\{e_t\}$ is a sequence of i.i.d random variables.

- A more general model that incorporates additional state variables is given by

$$N_t = f(\mathbf{X}_{t-1}, \mathbf{X}_{t-2}, \dots, \mathbf{X}_{t-d}; \boldsymbol{\theta}) + e_t$$

where \mathbf{X} is a vector of predictors.

Example: Lorenz System

- Coupled, nonlinear system of three first order differential equations:

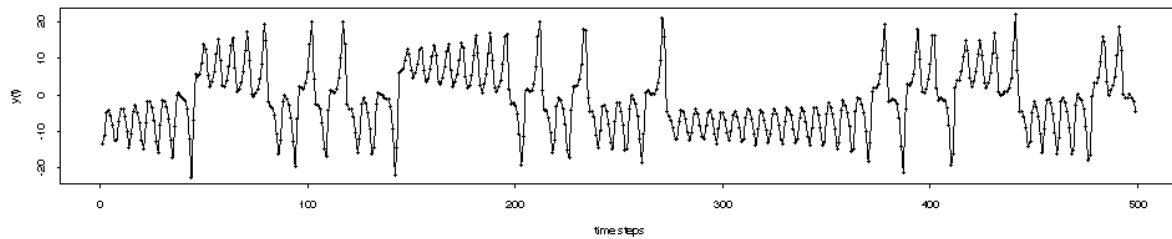
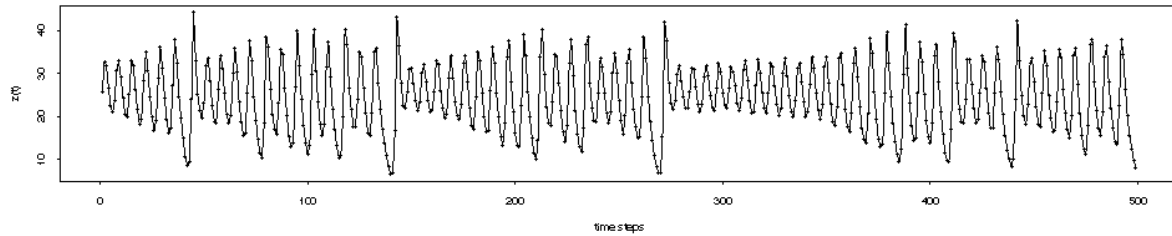
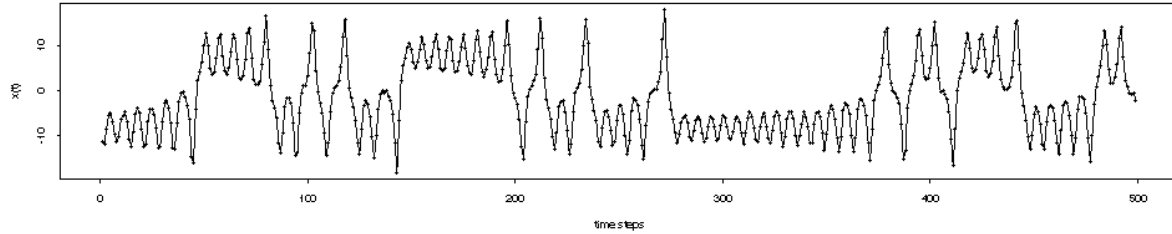
$$\begin{aligned}\frac{dx}{dt} &= -s(x - y) \\ \frac{dy}{dt} &= -xz + rx - y \\ \frac{dz}{dt} &= xy - bz\end{aligned}$$

for $s = 10$, $r = 28$ and $b = 8/3$ get famous “butterfly”

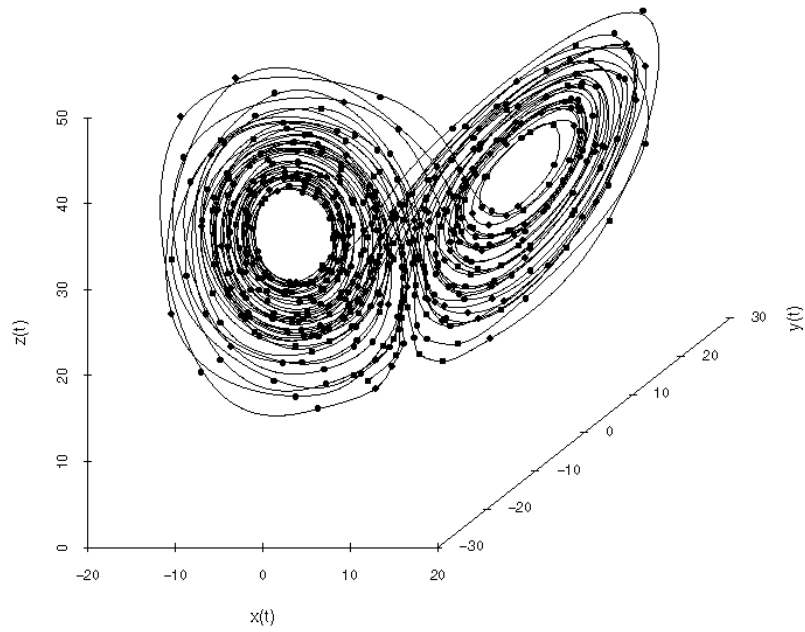
- Data: numerically integrate and add noise at every integration time-step.
- State-Space System:

$$\begin{aligned}X_t &= F_1(X_{t-1}, Y_{t-1}, Z_{t-1}) + e_{1,t} \\ Y_t &= F_2(X_{t-1}, Y_{t-1}, Z_{t-1}) + e_{2,t} \\ Z_t &= F_3(X_{t-1}, Y_{t-1}, Z_{t-1}) + e_{3,t}\end{aligned}$$

Lorenz Time Series, $X(t)$, $Z(t)$, $Y(t)$



Phase Space for Noisy Lorenz System



Neural Networks

- Class of nonlinear models:

$$Y = f(\mathbf{x}; \boldsymbol{\theta}) + e$$

- The form of the model:

$$f(\mathbf{x}) = \beta_0 + \sum_{i=1}^k \beta_i \varphi(\mathbf{x}^T \boldsymbol{\gamma}_i + \mu_i)$$

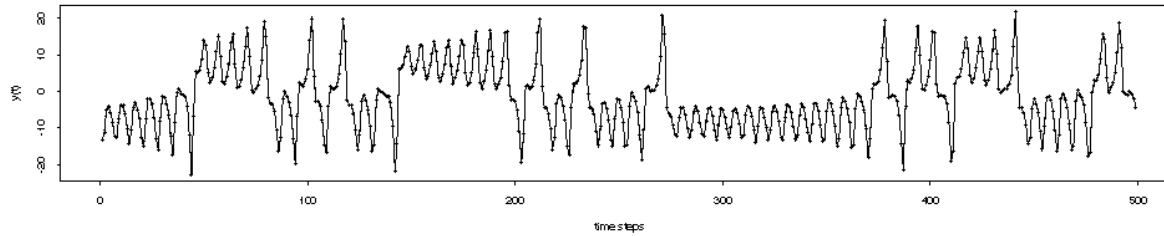
where $\varphi(u) = e^u / (1 + e^u)$

- Net parameters are estimated by nonlinear least squares.
- Total number of parameters is $p = 1 + k(d + 2)$ where d is dimension of \mathbf{x} .
- Complexity of the model chosen by Cross Validation:

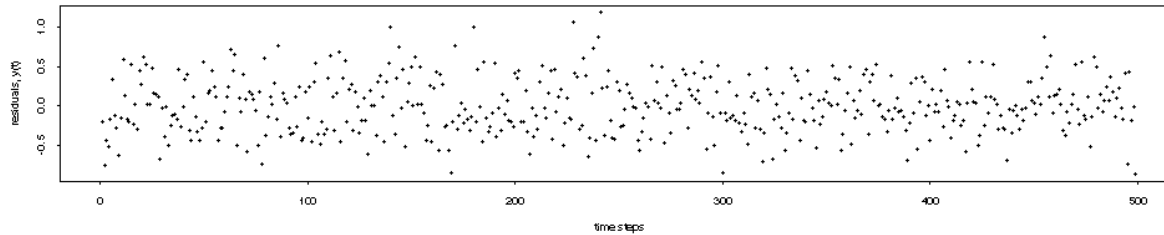
$$V_c = \frac{\frac{1}{n} RSS}{\left(1 - p \frac{c}{n}\right)^2}$$

Neural Network Fits to Noisy Lorenz System, $Y(t)$

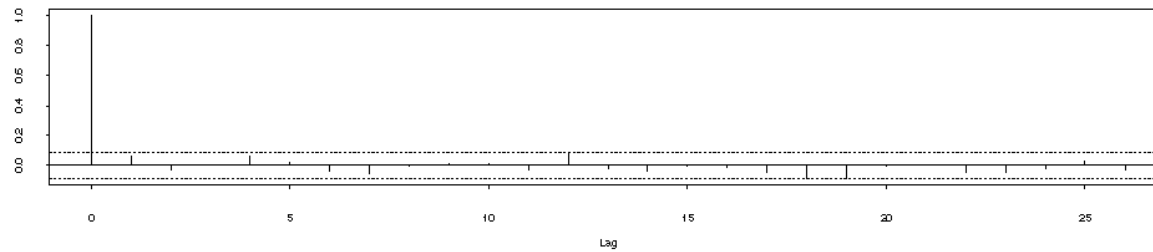
Data and Fitted Values



Residuals



Series : residuals



Nonparametric Bootstrap Prediction Intervals

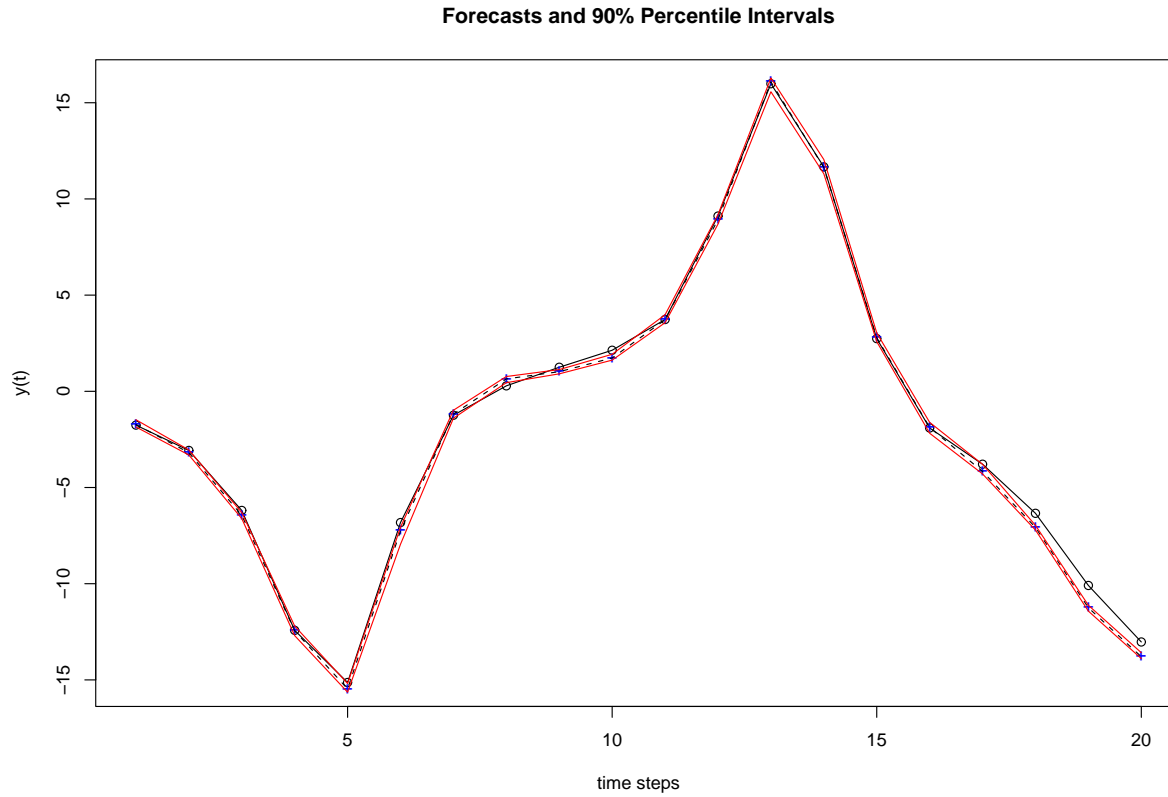
Nonlinear Time Series Model: $Y_t = f(X_{t-1}, Y_{t-1}, Z_{t-1}) + e_t$

Bootstrap “pairs” of the data. Have sample of n observations: (y_t, \mathbf{x}_t)

where $\mathbf{x}_t = (x_{t-1}, y_{t-1}, z_{t-1})$

- Sample the n data pairs with replacement to obtain n bootstrap pairs with which to estimate $f(\cdot)$. Repeat to obtain $B=1000$ bootstrap estimates of f .
- Generate point forecasts as the average of the B bootstrap predicted values.
- Percentile Prediction Intervals: after trimming off the top and bottom 5%, the endpoints of the interval gives the 90th percentile bootstrap prediction interval.

Forecasts for $Y(t)$



Nonlinear Time Series Analysis of Gulf of Maine Calanus Abundance

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- New Buoy Data: RSWT proxy data from Northeast Channel (near GoMOOS mooring site)
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- **Calanus Nonlinear Time Series Model:**

1. Data: bimonthly

$$C_t = f(C_{t-1}, B_{t-12}, \sin(2\pi t/6), \cos(2\pi t/6)) + e_t$$

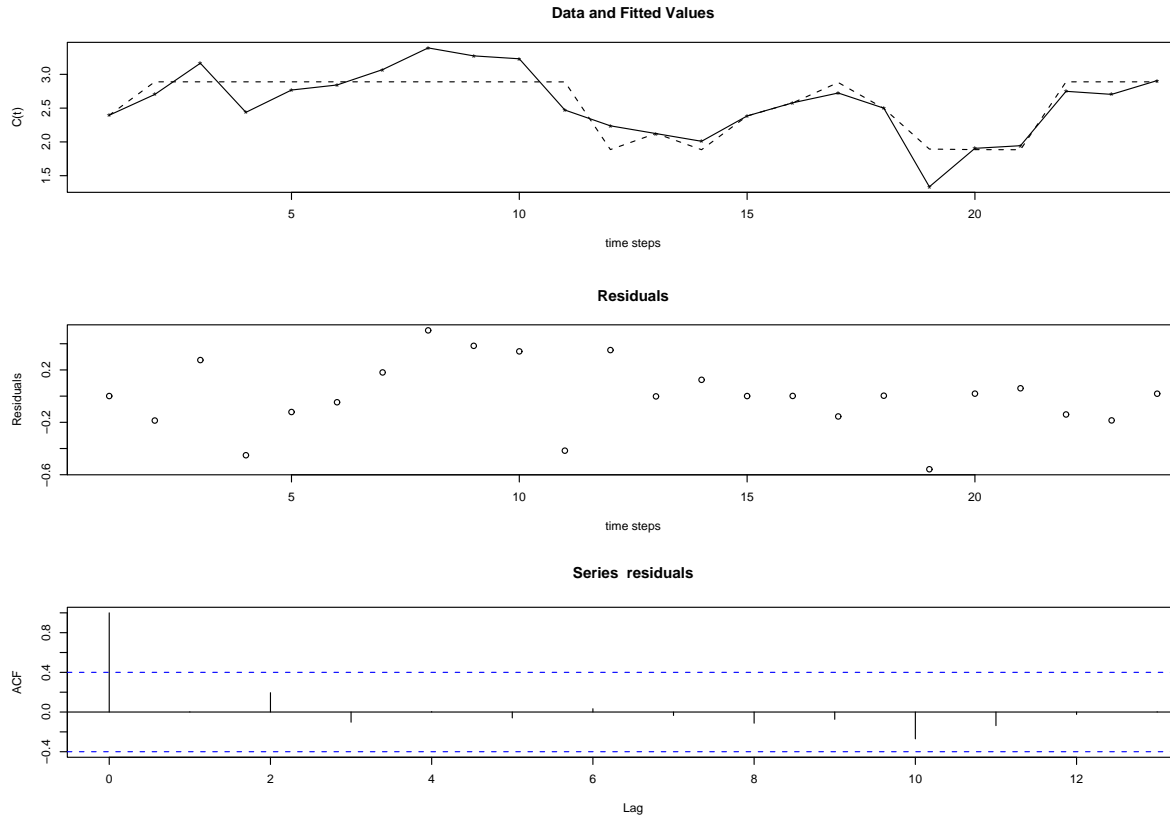
2. Data: yearly

$$C_t = f(C_{t-1}, B_{t-2}, H_{t-1}) + e_t$$

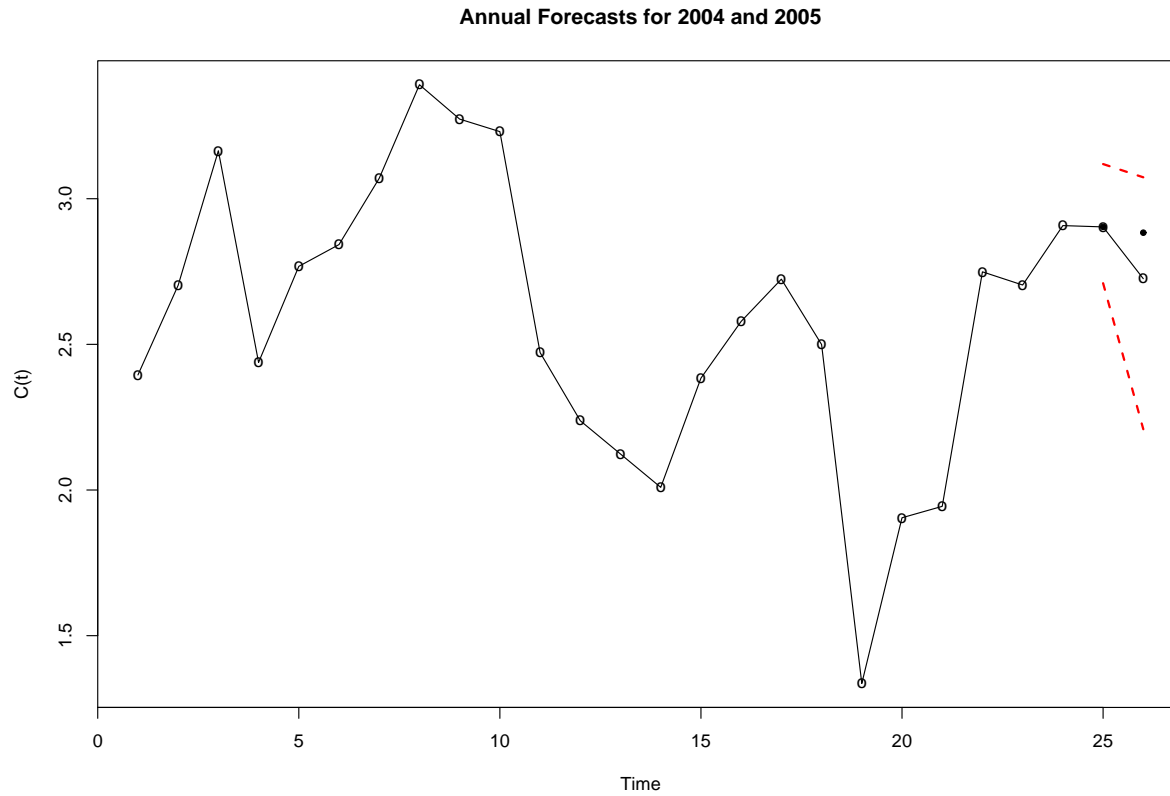
– Fit model to data 1978-2003

– Forecast 2004 and 2005, starting with data at time t and iterating model.

Neural Network Fits to C_t for Annual Data



Forecasts and Bootstrap Percentile Intervals



Future Work

- Look for other important variables to include in the Calanus prediction model.
 - Variables that are suspected to interact with Calanus.
 - Regional: Eastern and Western Gulf of Main Calanus Data