Data Assimilation and Genetic Algorithms for the Parameter Estimation Problem in Simple Climate Models

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 Start with some observation data for some phenomenon

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- Start with some observation data for some phenomenon
- Start with a well-principled model for that phenomenon

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- Start with some observation data for some phenomenon
- Start with a well-principled model for that phenomenon
- Can we find model parameters to reproduce the observation data?



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Systems of Interest

Lorenz '63

$$\frac{dx}{dt} = \sigma(y - x)$$

$$\frac{dy}{dt} = \rho x - y - xz$$

$$\frac{dz}{dt} = xy - \beta z.$$

Lorenz '96

$$\frac{dx_{i}}{dt} = x_{i-1}(x_{i+1} - x_{i-2}) - x_{i} + F - \frac{hc}{b} \sum_{j=1}^{J} y_{(j,i)}$$

$$\frac{dy_{(j,i)}}{dt} = cby_{(j+1,i)}(y_{(j-1,i)} - y_{(j+2,i)}) - cy_{(j,i)} + \frac{hc}{b}x_{i}$$

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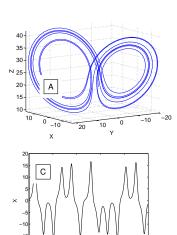
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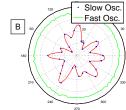
Lorenz '63 (L63)

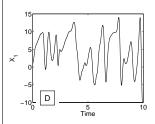


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Lorenz '96 (L96)







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Evolution is a pretty sweet problem solver...just need to fit the paradigm

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- Evolution is a pretty sweet problem solver...just need to fit the paradigm
- Population: real-valued vectors (called individuals or genes), each index represents a parameter choice (or alleles)

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- Population: real-valued vectors (called individuals or genes), each index represents a parameter choice (or alleles)
- Genetic Mutation: with some probability, add or subtract a small amount from an index in a vector

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- ➤ Selection Pressure: integrate model using parameters from an individual in the population. The point-wise root mean square error of this time series compared to observation data (or fitness)

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- Genetic Mutation: with some probability, add or subtract a small amount from an index in a vector
- ➤ Selection Pressure: integrate model using parameters from an individual in the population. The point-wise root mean square error of this time series compared to observation data (or fitness)
- Feproduction/Crossover: select two individuals from the population, called parents, and select a random index. Create to children vectors. The first child copies the entries from Parent 1 up to the selected index, and fills the remaining entries from Parent 2. Swap the roles of the parents to create the second child. The children replace the parents in the

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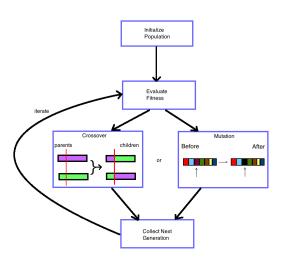
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▶ For the L63: $\sigma = 10 \& \beta = 8/3$

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- ▶ For the L63: $\sigma = 10 \& \beta = 8/3$
- ▶ Test the effects of system dynamics on GA by tuning $\rho \in [22, 28, 35]$

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- ▶ For the L63: $\sigma = 10 \& \beta = 8/3$
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- ▶ For the L96: h = 1, b = c = 10, F = 14 & J = 4

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- ▶ For the L63: $\sigma = 10 \& \beta = 8/3$
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- ► For the L96: h = 1, b = c = 10, F = 14 & J = 4
- Test the effects of system dimensionality on GA by tuning I ∈ [4, 8, 10, 15]

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- Test the effects of subsampling

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- ▶ Test the effects of system dynamics on GA by tuning $\rho \in [22, 28, 35]$
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- ► Test the effects of system dimensionality on GA by tuning I ∈ [4, 8, 10, 15]
- Test the effects of subsampling
- Test the effects of normal and uniformly distributed noise in the observation data

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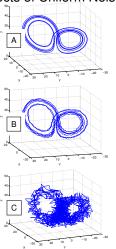
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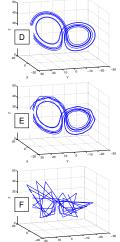


The Effects of Subsampling and Artificial Noise in Observation Data

Effects of Uniform Noise:



Effects of Subsampling:



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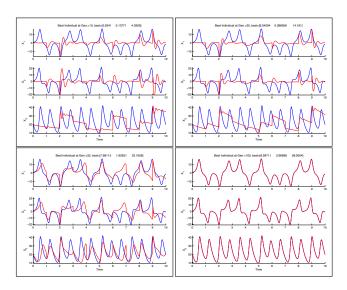
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Running the GA



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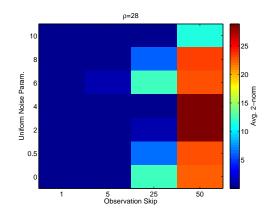
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System Dynamics in L63



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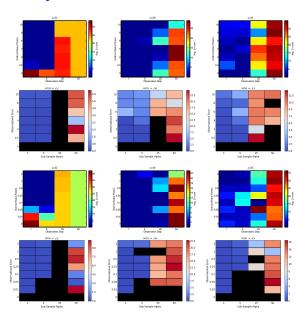
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System Dynamics in L63



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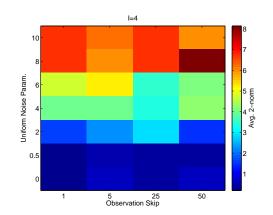
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Dimensionality in L96



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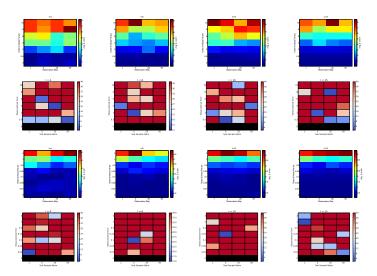
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Dimensionality in L96



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► GA performs comparably to Data Assimilation (DA) for Parameter Estimation as "out-of-the-box" tools

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- GA performs comparably to Data Assimilation (DA) for Parameter Estimation as "out-of-the-box" tools
- DA diverged for some experiments. Steps can be taken to prevent this. GA performs without special considerations.

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- GA performs comparably to Data Assimilation (DA) for Parameter Estimation as "out-of-the-box" tools
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- Initial results show GA FAILS the one-variable problem, while DA got it right for many cases.

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- need to try with other simple models. Need to see how far up the model hierarchy before GA is impractical or ineffective.

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- DA diverged for some experiments. Steps can be taken to prevent this. GA performs without special considerations.
- Initial results show GA FAILS the one-variable problem, while DA got it right for many cases.
- need to try with other simple models. Need to see how far up the model hierarchy before GA is impractical or ineffective.
- Attempt a real-world problem

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