

Types of Error

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x}) + \boldsymbol{\zeta} \quad (\text{Type I})$$

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x} + \boldsymbol{\xi}) \quad (\text{Type II})$$

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x} + \boldsymbol{\xi}) + \boldsymbol{\zeta} \quad (\text{Type III})$$

For simplicity, we assume that $\boldsymbol{\zeta}$ and $\boldsymbol{\xi}$ are both constant in time. when the error is small, they exhibit behaviors similar to those of the system.

$$\zeta_i = A \sin\left(2\pi \frac{i-1}{N}\right)$$

$$\xi_i = B \sin\left(2\pi \frac{i-1}{N}\right), \quad i = 1, \dots, N$$

where A and B are constants.

Model 1

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x}) + \boldsymbol{\zeta} \quad (\text{Type I})$$

$$\mathbf{b}_n = m(\mathbf{x}_{n-1}) - \hat{m}(\mathbf{x}_{n-1})$$

We therefore incorporate \mathbf{b}_n to the augmented system:

$$\mathbf{x}_n^f = \hat{m}(\mathbf{x}_{n-1}^a) + \mathbf{b}_n^f$$

$$\mathbf{b}_n^f = f_b(\mathbf{x}_{n-1}^a, \mathbf{b}_n^a)$$

In the case where $\boldsymbol{\zeta}$ is constant, we assume \mathbf{b}_n is constant too, and then

$$\mathbf{b}_n^f = f_b(\mathbf{x}_{n-1}^a, \mathbf{b}_{n-1}^a) = \mathbf{b}_{n-1}^a$$

Model 2

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x} + \boldsymbol{\xi}) \quad (\text{Type II})$$

$$\mathbf{c}_n = m(\mathbf{x}_{n-1}^t) - \hat{m}(\mathbf{x}_{n-1}^m) = m(\mathbf{x}_{n-1}^t) - \hat{m}(\mathbf{x}_{n-1}^t - \mathbf{c}_{n-1})$$

The augmented system then becomes

$$\mathbf{x}_n^f = \hat{m}(\mathbf{x}_{n-1}^a)$$

$$\mathbf{c}_n^f = f_c(\mathbf{x}_{n-1}^a, \mathbf{c}_n^a)$$

For simplicity, we assume we know $\boldsymbol{\xi}$ is constant, which means \mathbf{c}_n should be constant, and then

$$\mathbf{c}_n^f = f_c(\mathbf{x}_{n-1}^a, \mathbf{c}_{n-1}^a) = \mathbf{c}_{n-1}^a$$

The analysis will model the model trajectory $\hat{\mathbf{x}}_n$ instead of the true trajectory \mathbf{x}_n , and therefore the observation should be:

$$\mathbf{y}_n = H(\hat{\mathbf{x}}_n + \mathbf{c}_n)$$

Model 3

$$\dot{\mathbf{x}} = \mathbf{L}(\mathbf{x} + \boldsymbol{\xi}) + \boldsymbol{\zeta} \quad (\text{Type III})$$

Combine the previous two models together, we have the following scheme:

$$\begin{aligned}\mathbf{x}_n^f &= \hat{m}(\mathbf{x}_{n-1}^a) + \mathbf{b}_n^f \\ \mathbf{b}_n^f &= f_b(\mathbf{x}_{n-1}^a, \mathbf{b}_{n-1}^a, \mathbf{c}_{n-1}^a) \\ \mathbf{c}_n^f &= f_c(\mathbf{x}_{n-1}^a, \mathbf{b}_{n-1}^a, \mathbf{c}_{n-1}^a)\end{aligned}$$

Again, for simplicity, we assume \mathbf{c}_n and \mathbf{b}_n are constant, i.e.,

$$\begin{aligned}\mathbf{b}_n^f &= f_b(\mathbf{x}_{n-1}^a, \mathbf{b}_{n-1}^a) = \mathbf{b}_{n-1}^a \\ \mathbf{c}_n^f &= f_c(\mathbf{x}_{n-1}^a, \mathbf{c}_{n-1}^a) = \mathbf{c}_{n-1}^a\end{aligned}$$

Observation operator:

$$\mathbf{y}_n = H(\hat{\mathbf{x}}_n + \mathbf{c}_n)$$

Experiment Setup

To improve the analyses in our experiments, we employ variance inflation

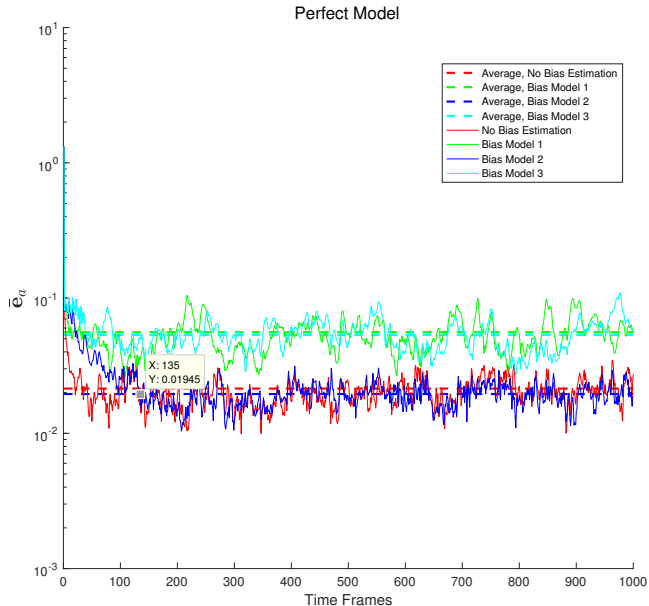
$$\mathbf{P}_n^a \rightarrow \mathbf{P}_n^a + \frac{\mu\Lambda}{K} \mathbf{I}_K$$

Time Step: $\Delta t = 0.05$

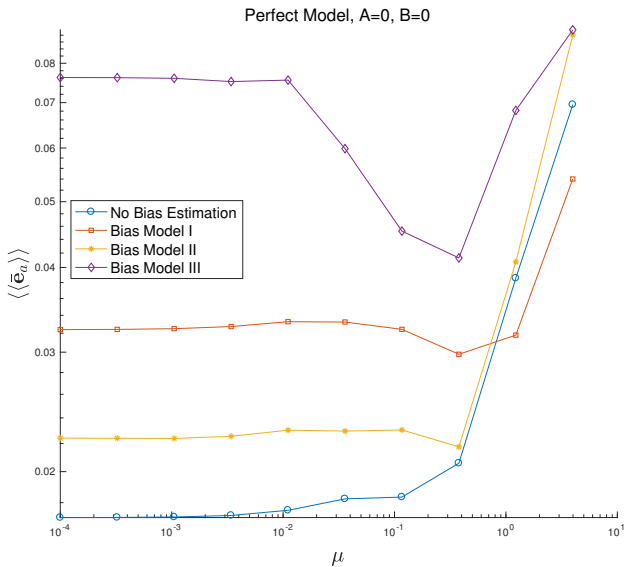
Initial Spread: 1.3

Ensemble Size: 40

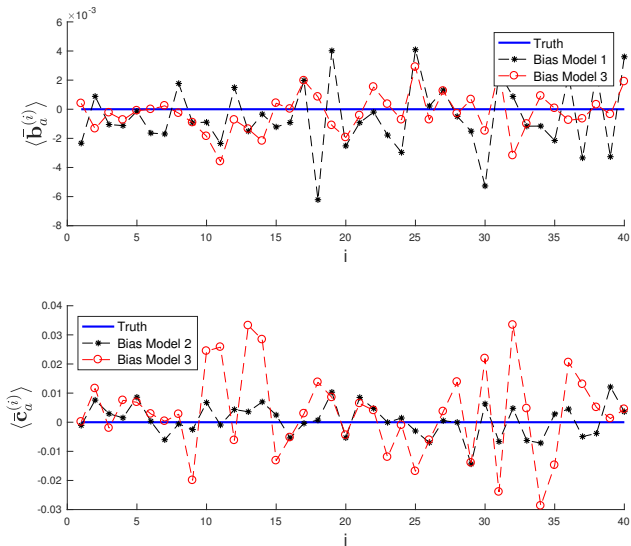
Perfect Situation: Settling Time for Error



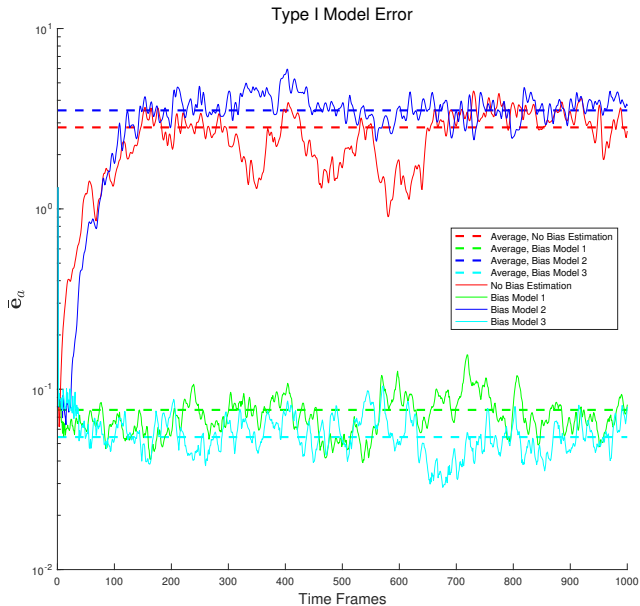
Perfect Situation: Performance and Inflation



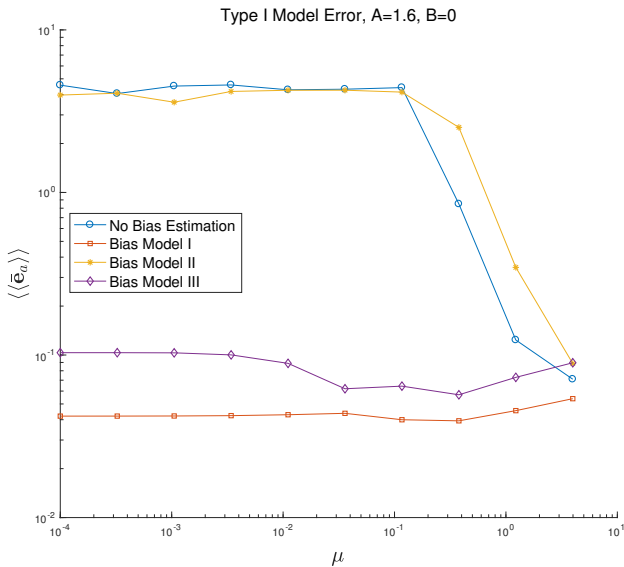
Perfect Situation: Bias Estimation



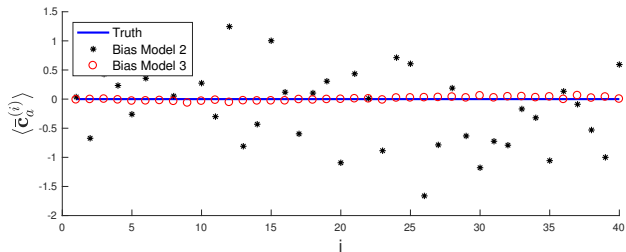
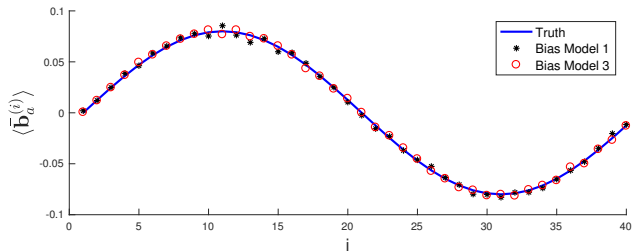
Type I Model Error: Settling Time for Error



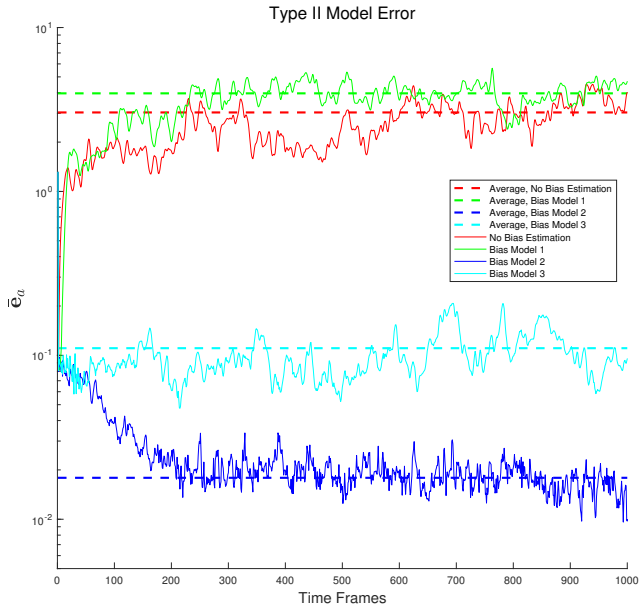
Type I Model Error: Performance and Inflation



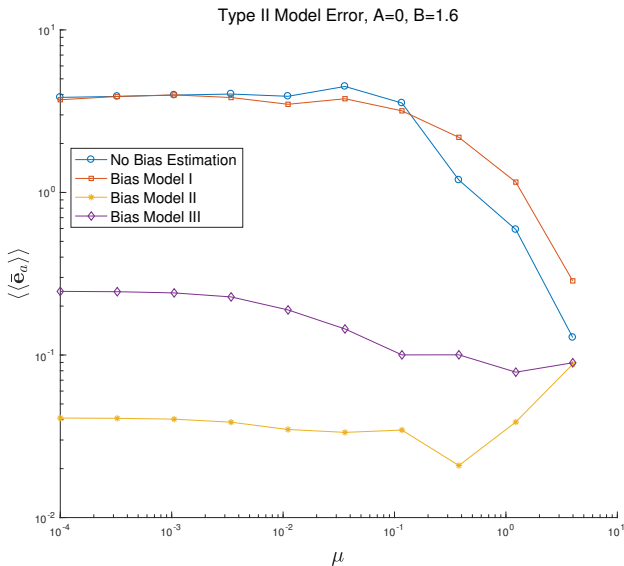
Type I Model Error: Bias Estimation



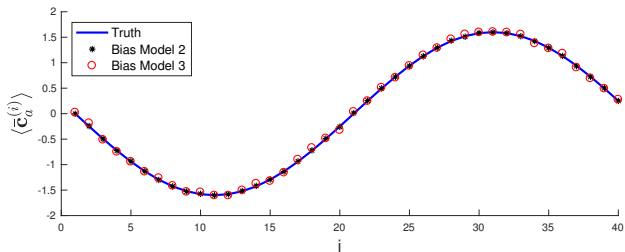
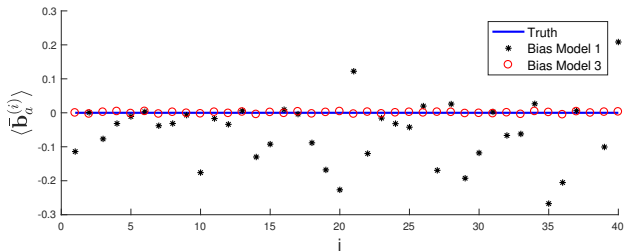
Type II Model Error: Settling Time for Error



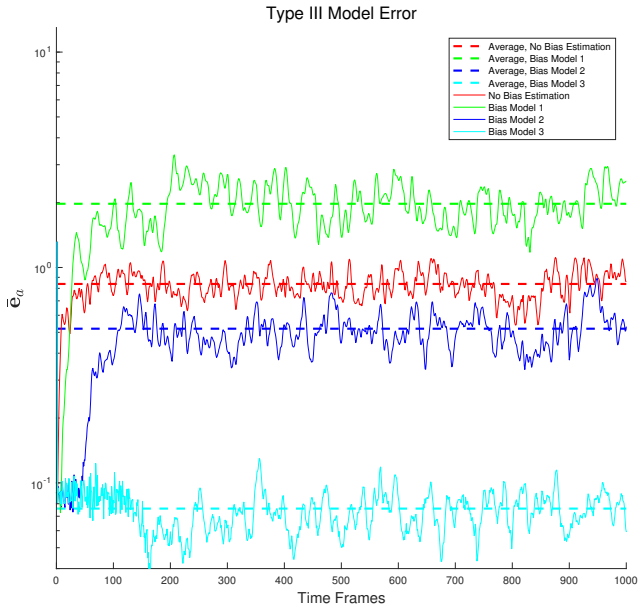
Type II Model Error: Performance and Inflation



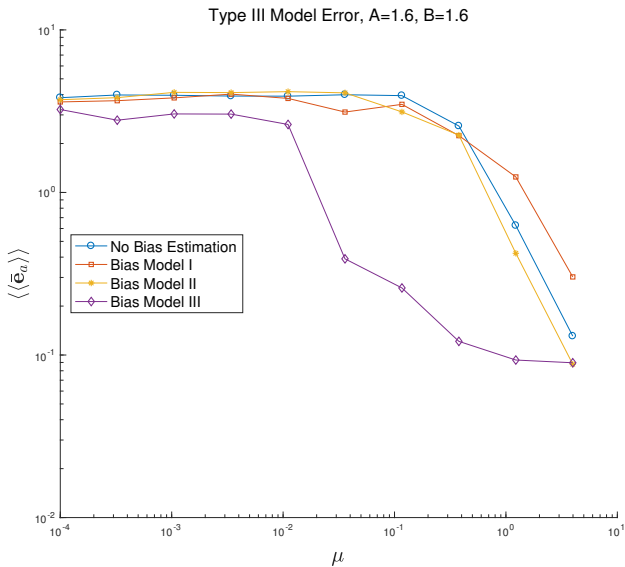
Type II Model Error: Bias Estimation



Type III Model Error: Settling Time for Error



Type III Model Error: Performance and Inflation



Type III Model Error: Bias Estimation

