

# Nonlinear Forces and Omitted Masses: Mass-Spring-Damper Models and Their Model-Form Errors

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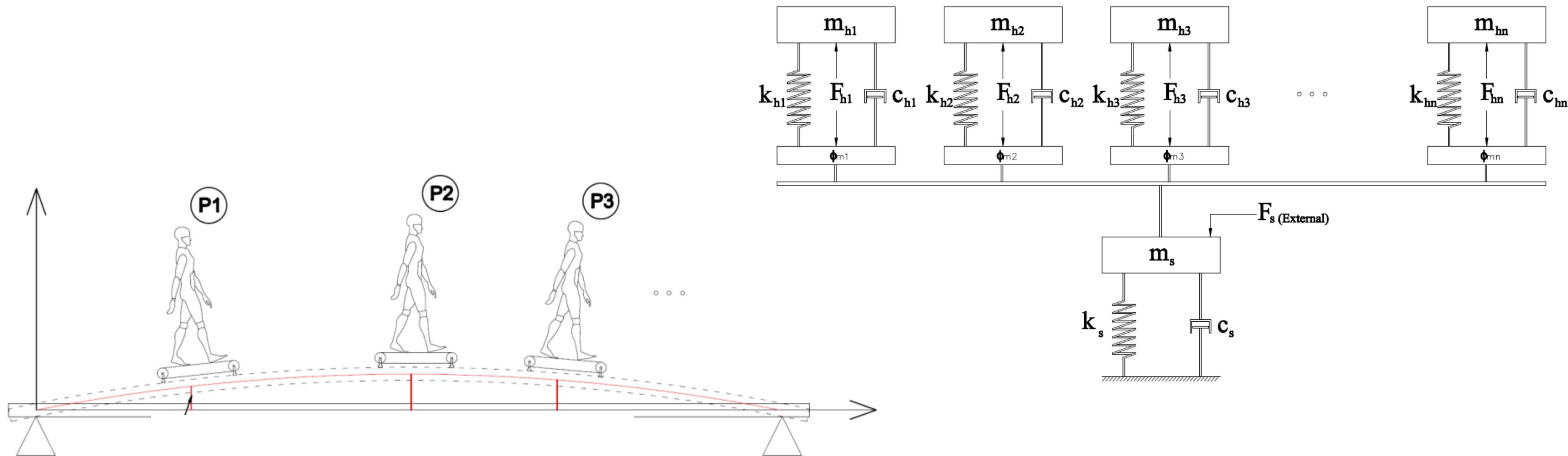
February 29, 2024



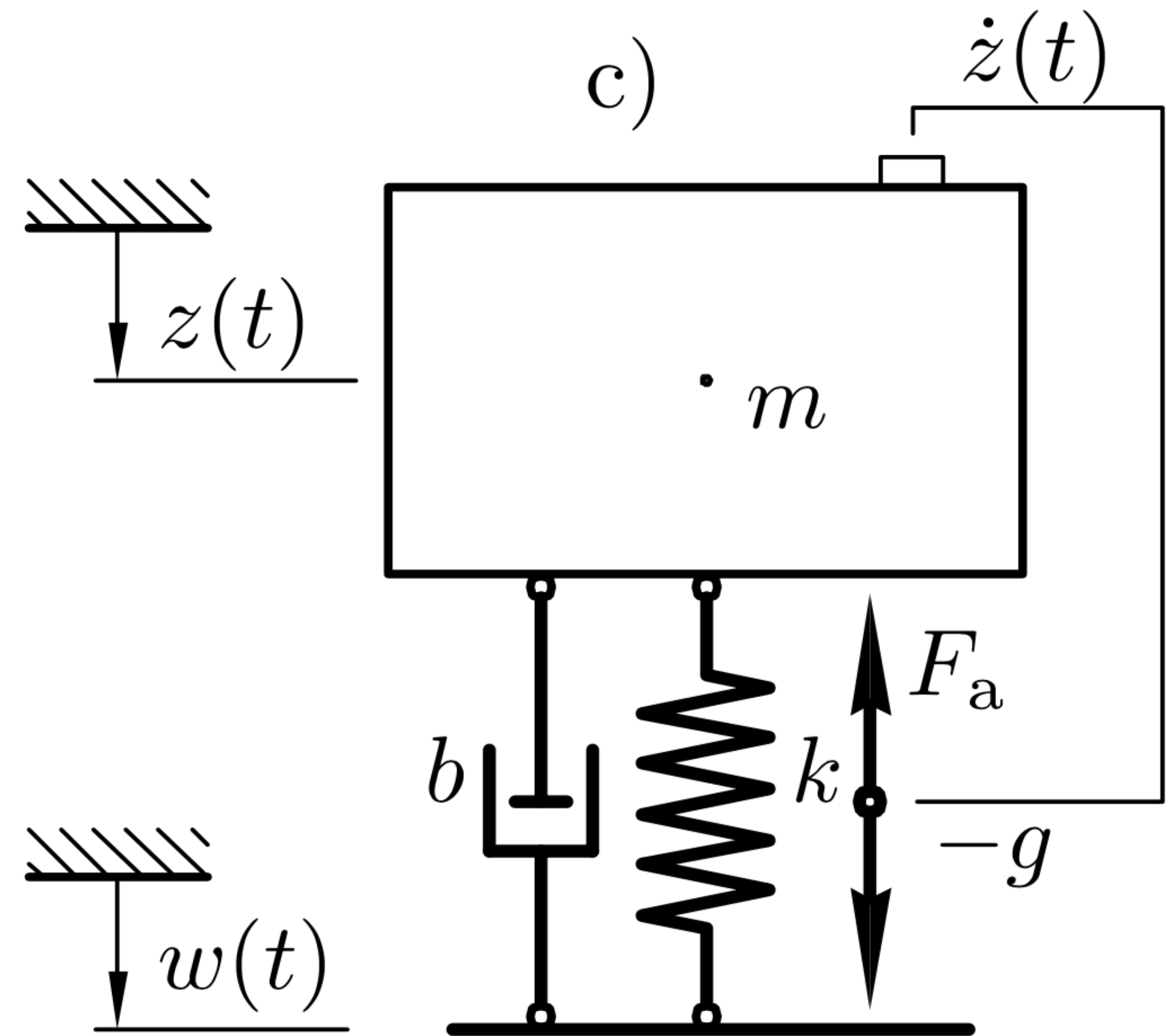
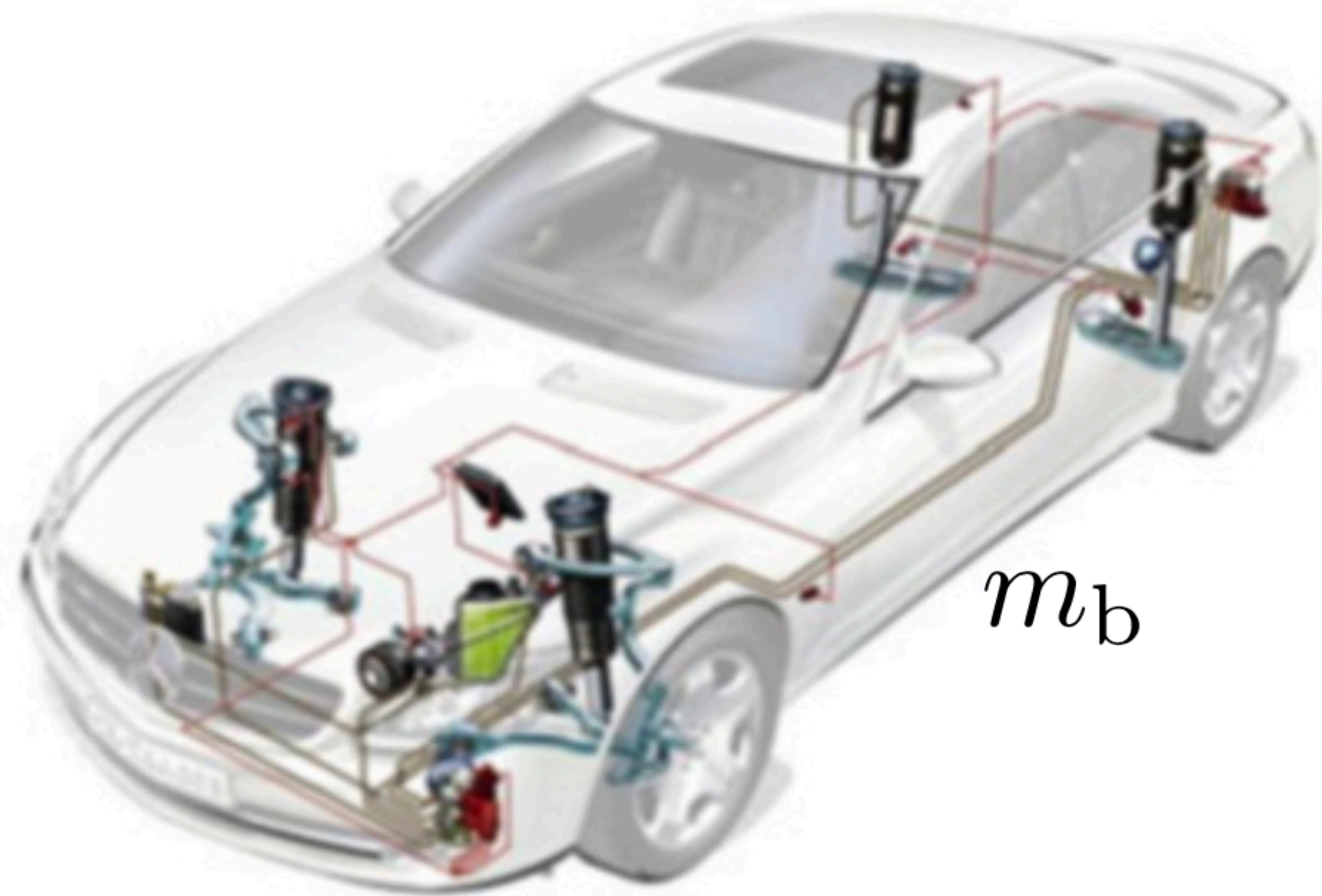
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SAND:TODO



# Mass-spring-damper models approximate the impact load on bridges

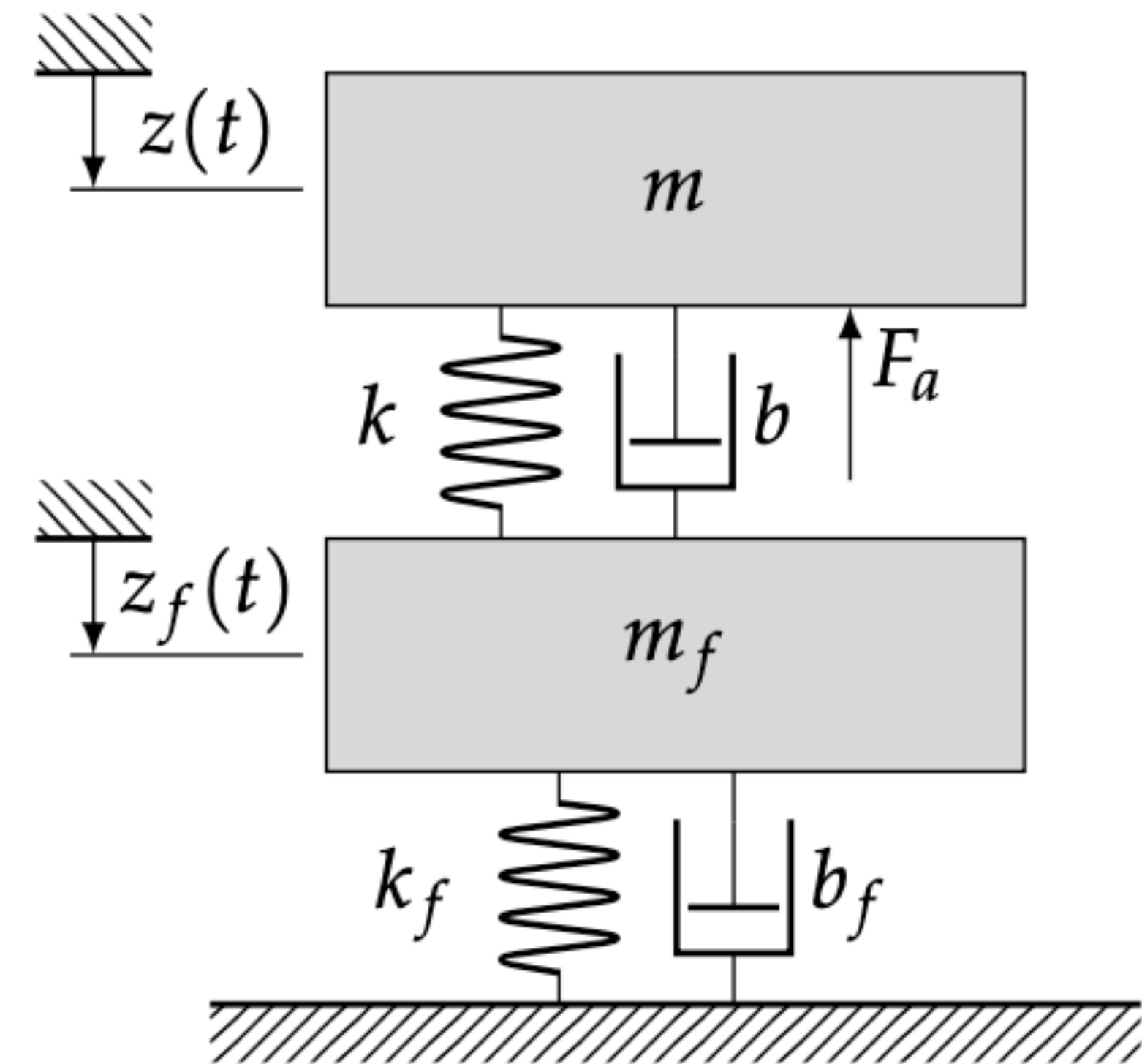
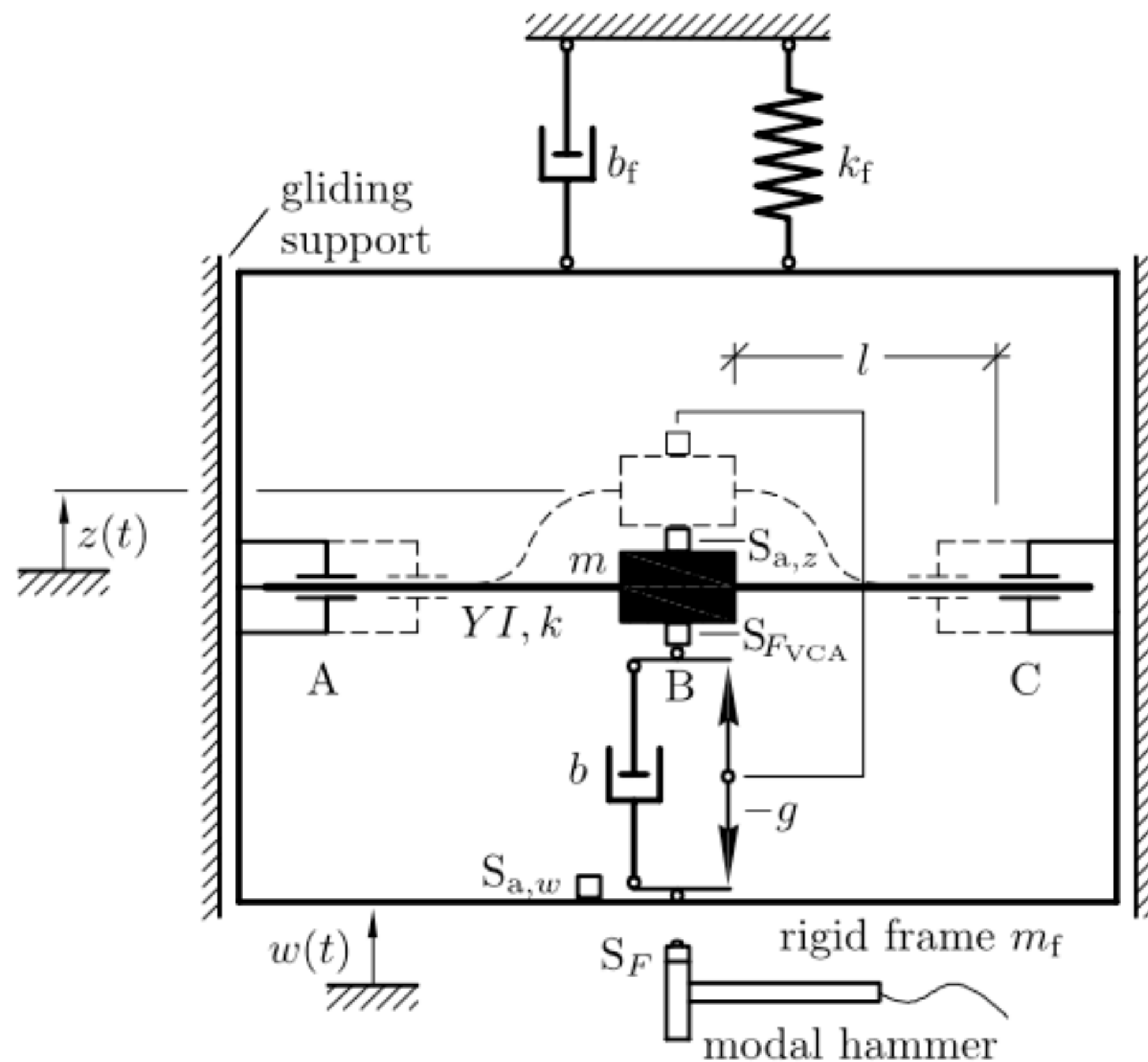


# A one-mass oscillator estimates a driving car's vertical dynamic behavior



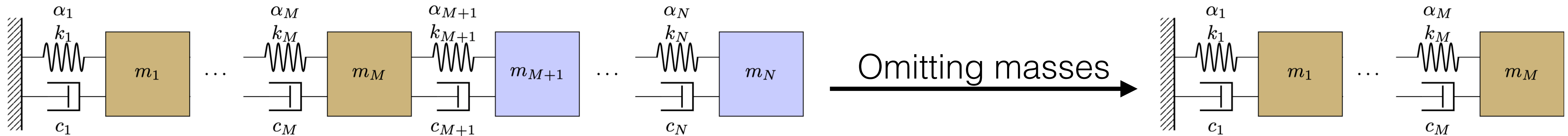


# A two mass oscillator approximates the vibration isolation experiment

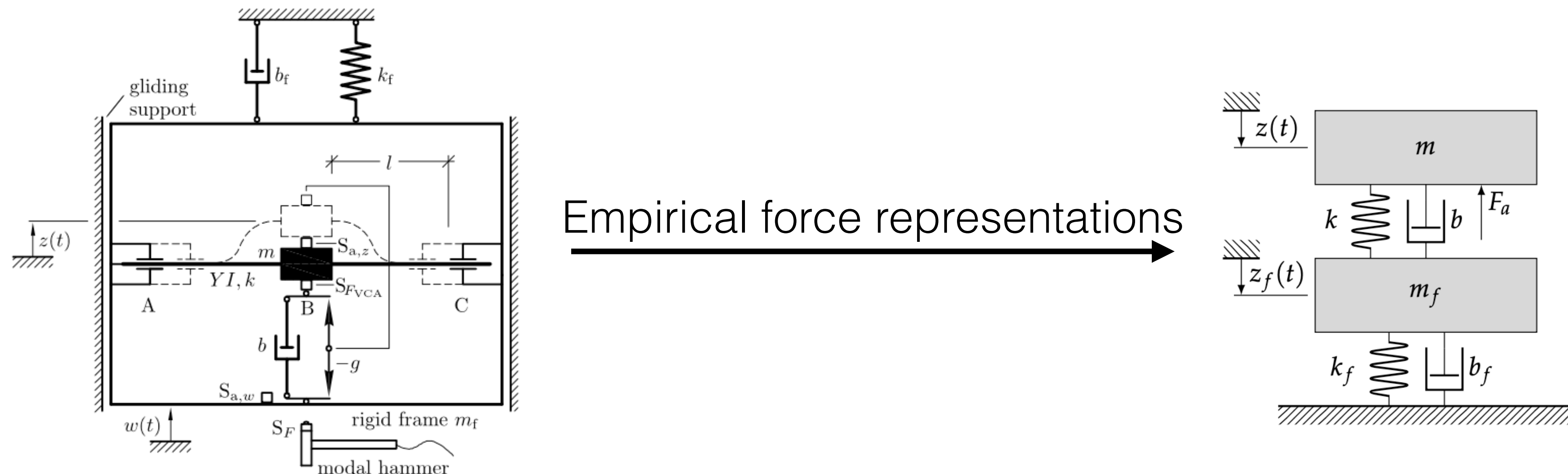


# How do model-form errors arise?

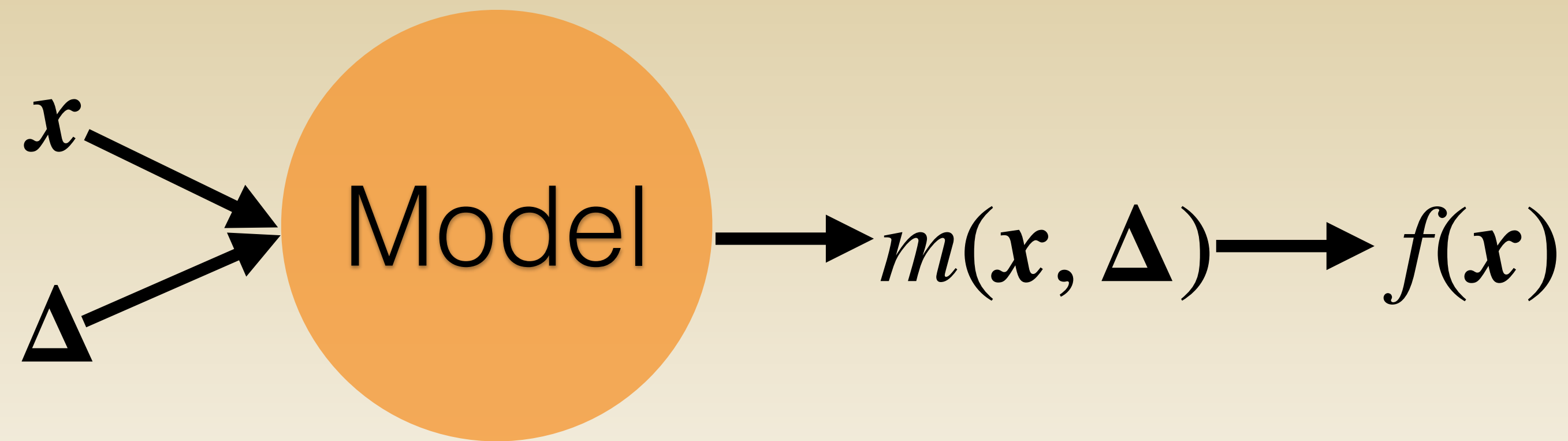
## Case 1



## Case 2

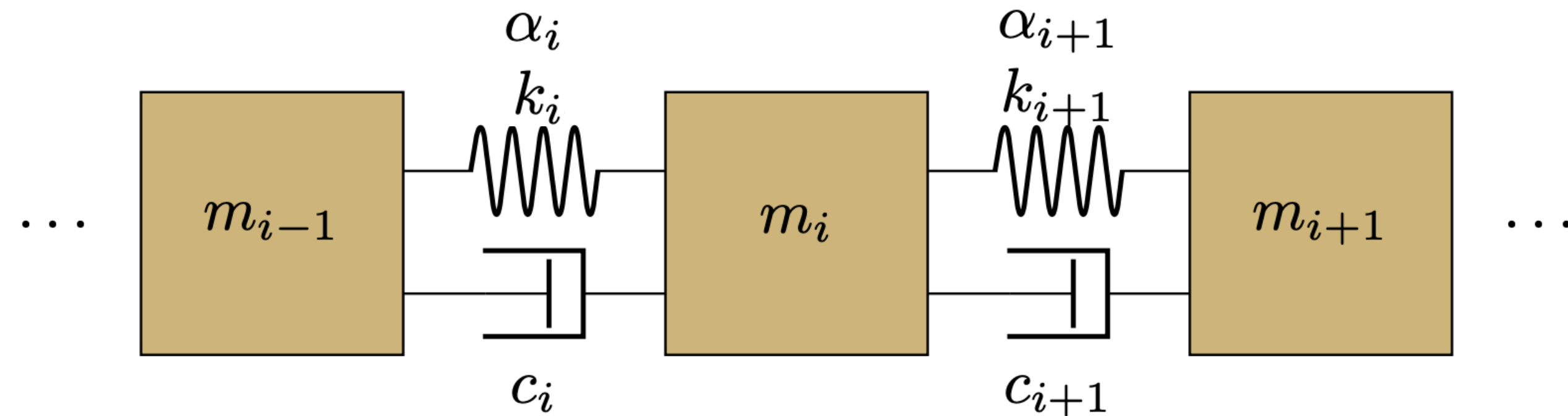


Embedded model enrichments reduce the model-form error and enable extrapolative predictions



- $\Delta$  is physics-informed  $\rightarrow$  extrapolative
- Quantified uncertainties informed by Bayesian calibration
- Computational cost:  
low-fidelity model  $\approx$  enriched model  $\ll$  high-fidelity model

# Case 1: forming mass-spring-damper models



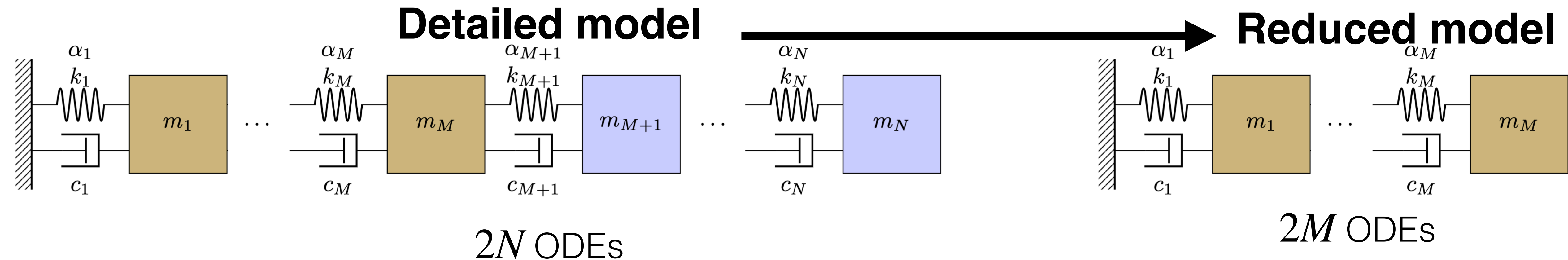
Newton's Second Law of Motion:

$$\text{diag}(\mathbf{m})\ddot{\mathbf{x}}(t) = \sum \mathbf{f} \\ = \mathbf{f}_d(t) + \mathbf{f}_s(t)$$

**Damping**  $\longrightarrow f_{d,i}(t) = -c_i \dot{x}_i(t)$

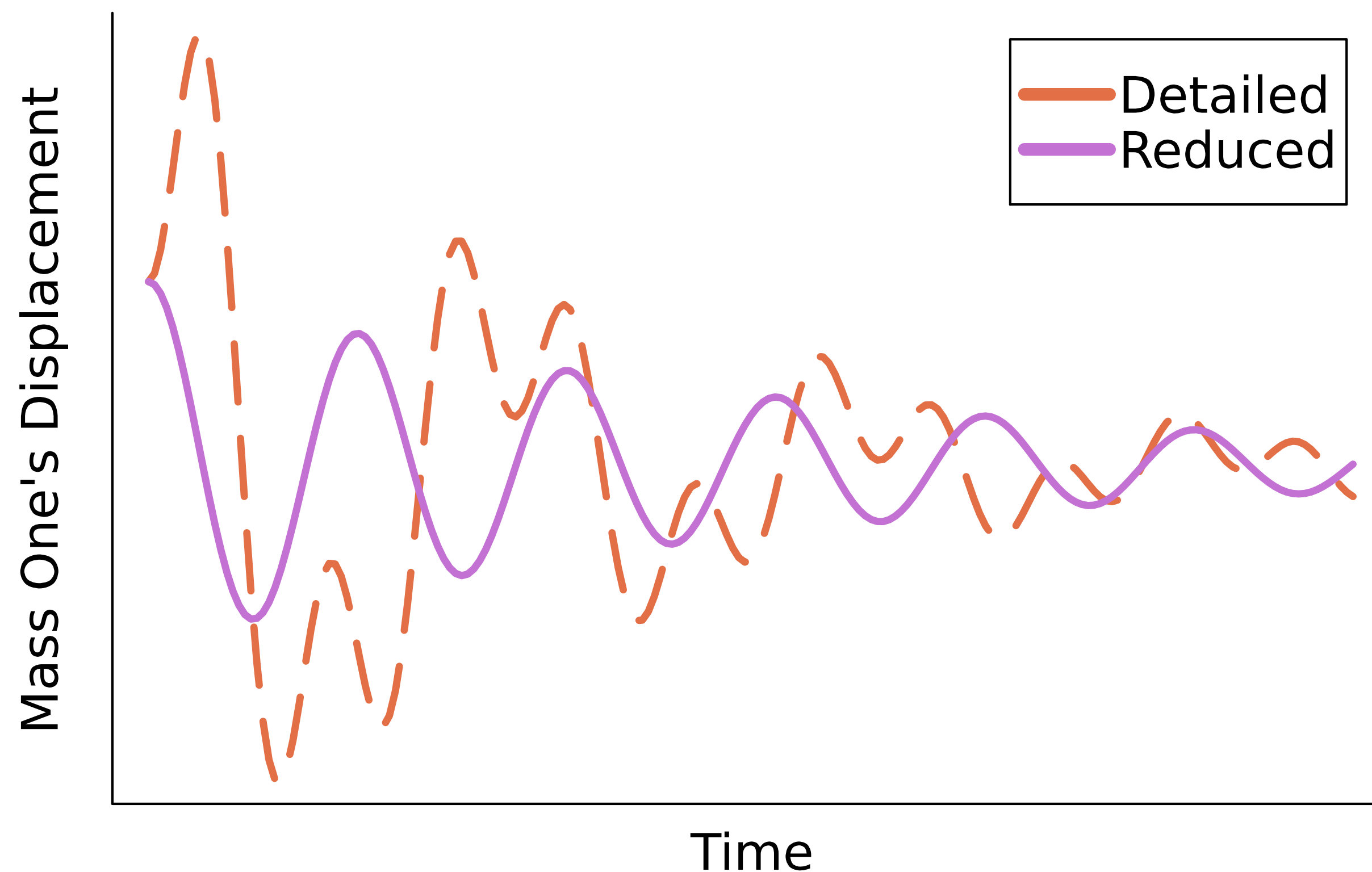
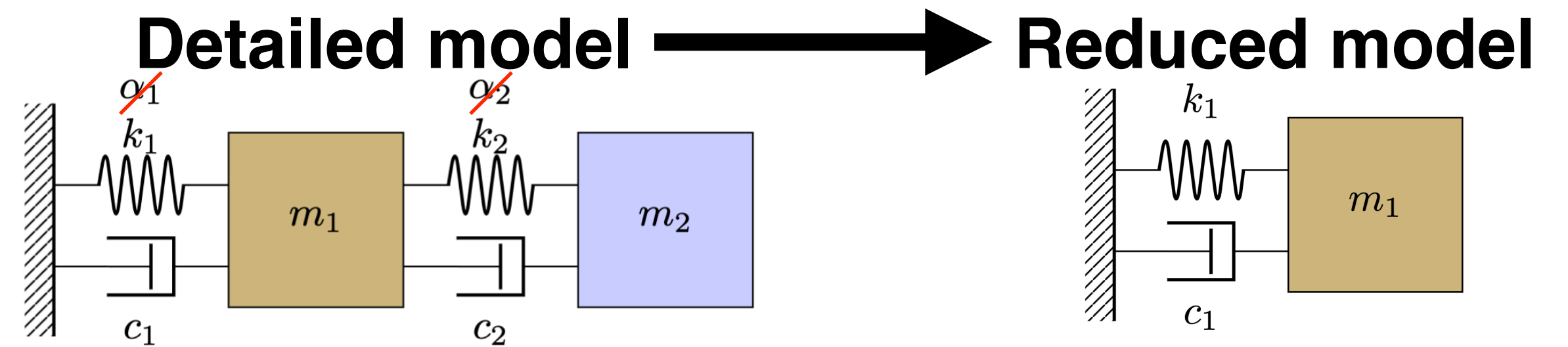
**Spring**  $\longrightarrow f_{s,i}(t) = \left[ \underbrace{-k_i(x_i(t) - x_{i-1}(t)) + k_{i+1}(x_{i+1}(t) - x_i(t))}_{\text{Hooke's law}} \right] \left( 1 + \underbrace{\alpha_i(x_{i+1}(t) - x_{i-1}(t))}_{\text{nonlinear term}} \right)$

The reduced model is formed  
by subsampling the detailed model

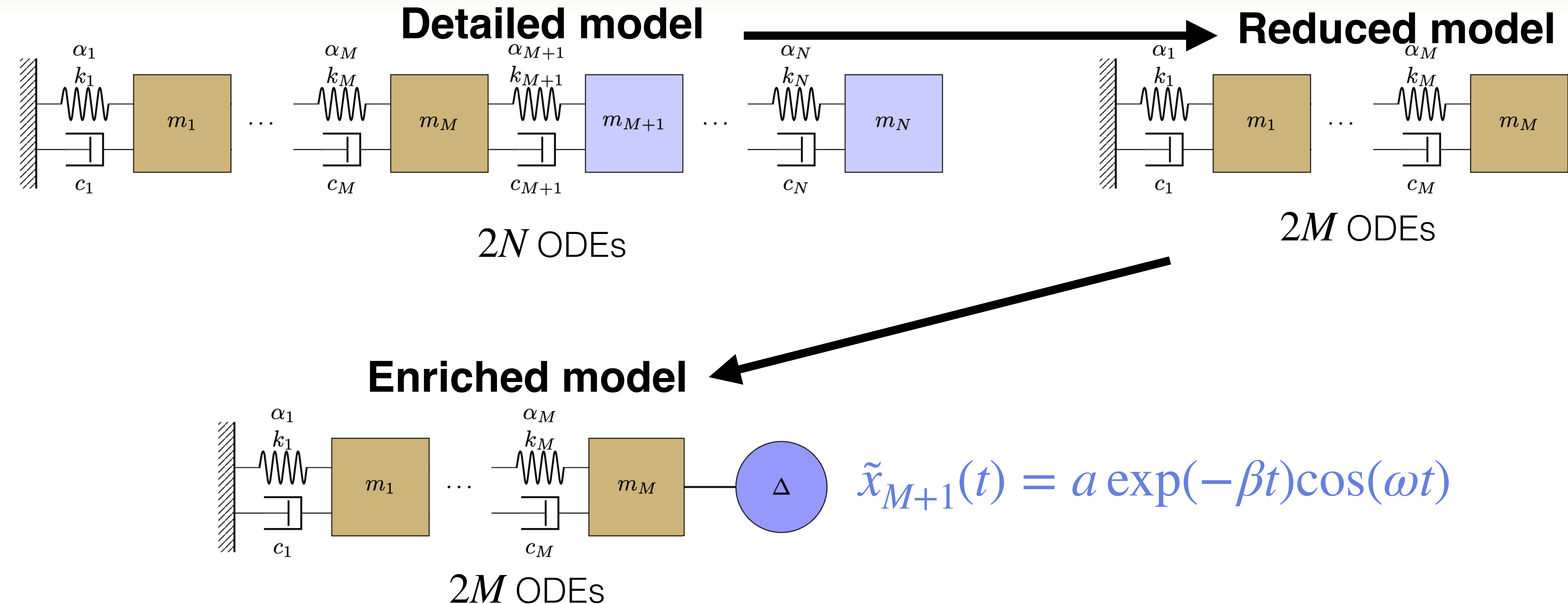




# Omitting masses causes error in the reduced model

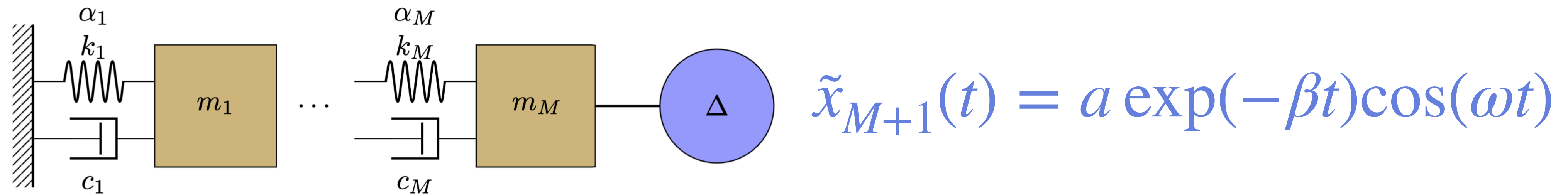


An enrichment operator is added to the reduced model to form the enriched models



The enrichment operator approximates the movement of mass  $M + 1$  with a simple oscillator

### Enriched model



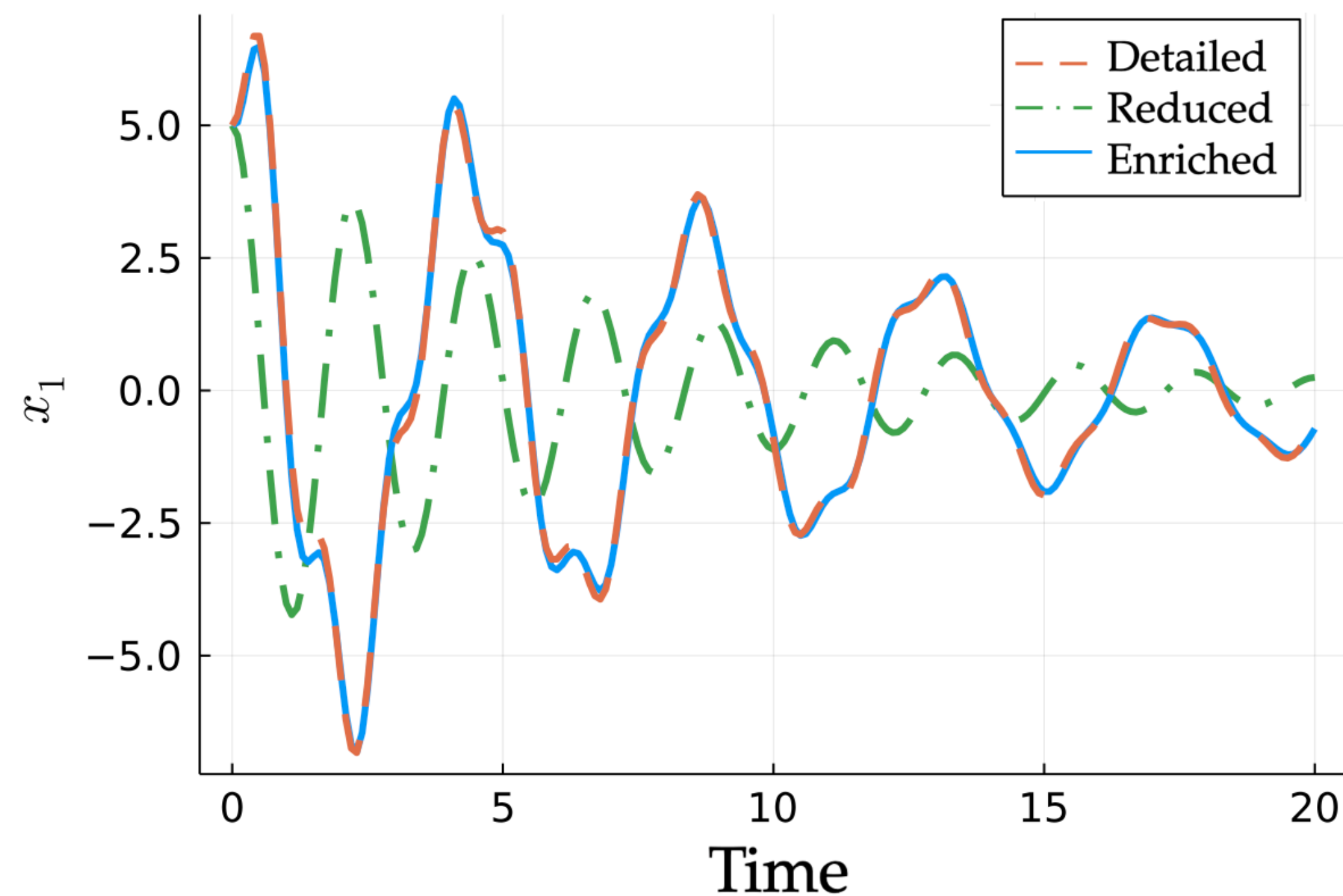
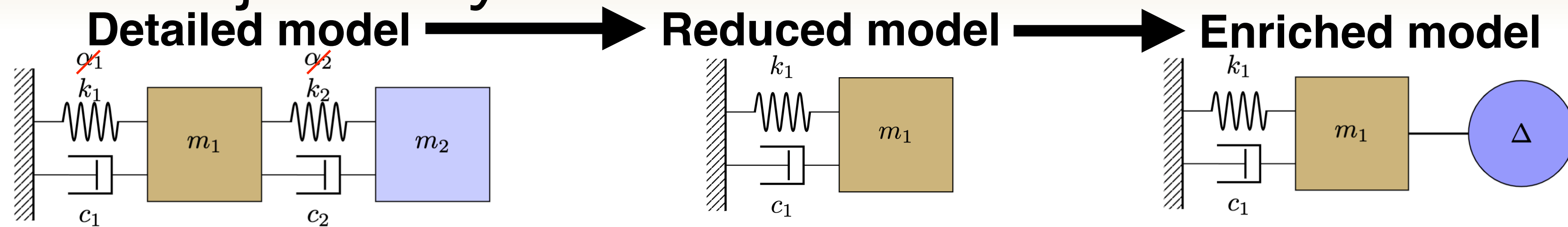
$$\ddot{x}_M(t) = \frac{1}{m_M} \left\{ -c_M \dot{x}(t) + \left[ -k_M (x_M(t) - x_{M-1}(t)) + \delta(\tilde{x}_{M+1}(t) - x_M(t)) \right] (1 + \alpha_i(\tilde{x}_{M+1}(t) - x_{M-1}(t))) \right\}$$

- Model parameters:  $\boldsymbol{\theta} = (\delta, a, \beta, \omega)$

$$\theta_i \sim \mathcal{N}(\mu_i, \sigma_i^2), \text{ where } \mu_i \in \mathbb{R}, \sigma_i \in \mathbb{R}_{\geq 0}, \text{ and } i = \{1, 2, 3, 4\}$$

- Hyperparameters:  $\boldsymbol{\phi} = (\boldsymbol{\phi}_1, \boldsymbol{\phi}_2, \boldsymbol{\phi}_3, \boldsymbol{\phi}_4)$ , where  $\boldsymbol{\phi}_i = (\mu_i, \sigma_i)$
- We use hierarchical Bayesian calibration to sample a posterior on  $\boldsymbol{\phi}$

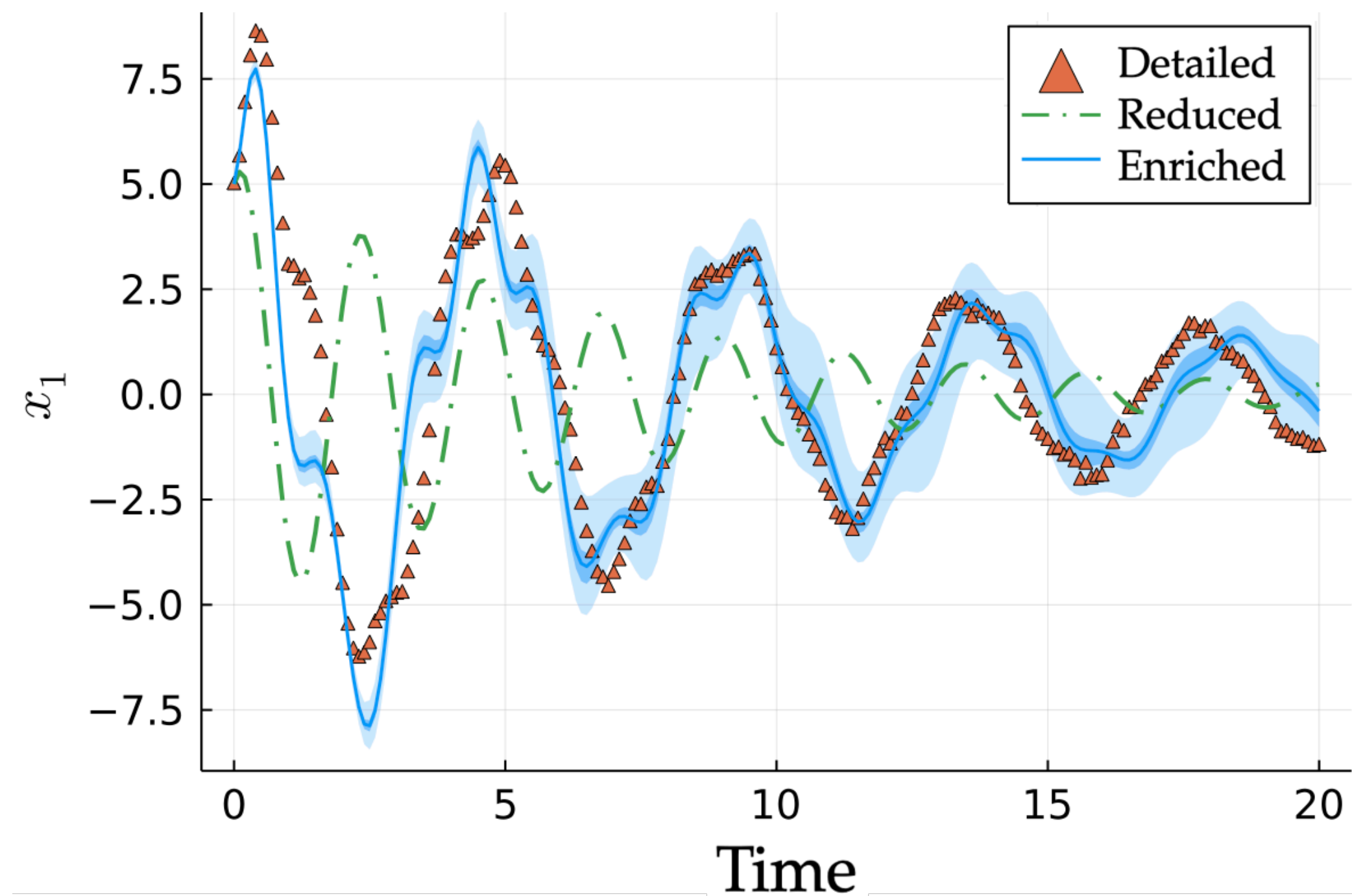
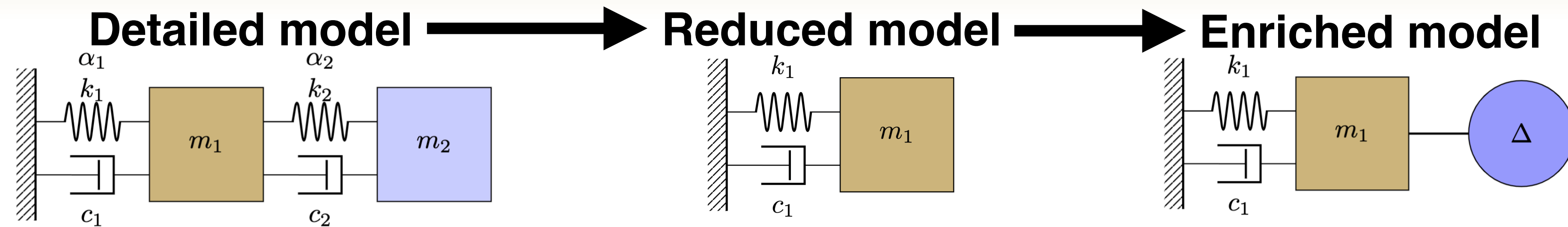
# The enriched model almost perfectly matches the trajectory of the **linear** detailed model



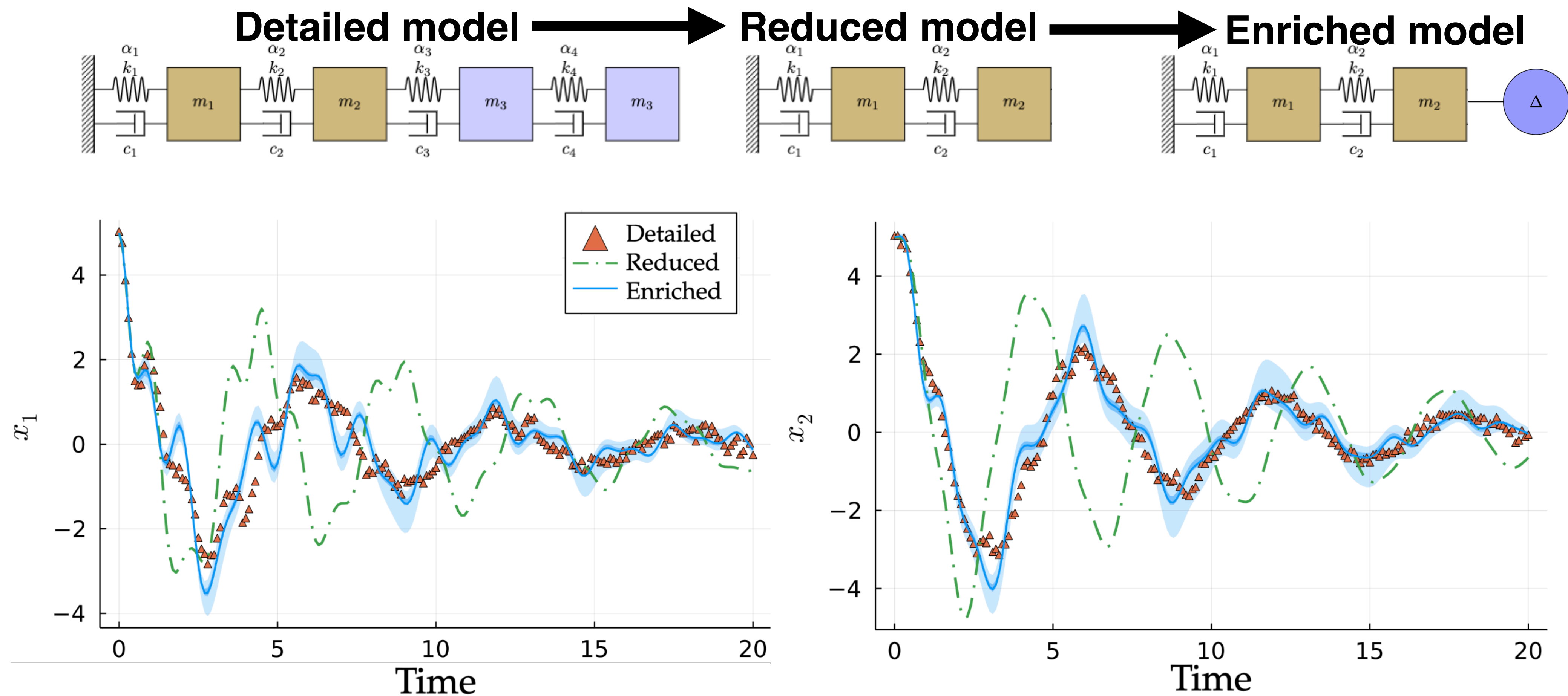
Enriched model is calibrated with mass one's displacement data.



# Discrepancies emerge between the enriched model and the **nonlinear** detailed model

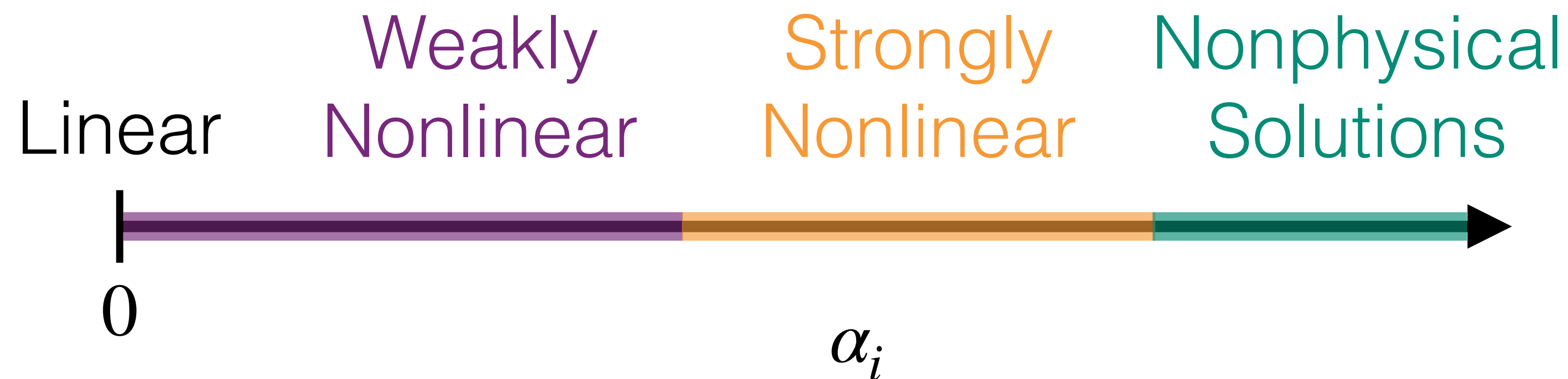


# Discrepancies emerge between the enriched model and the **nonlinear** detailed model



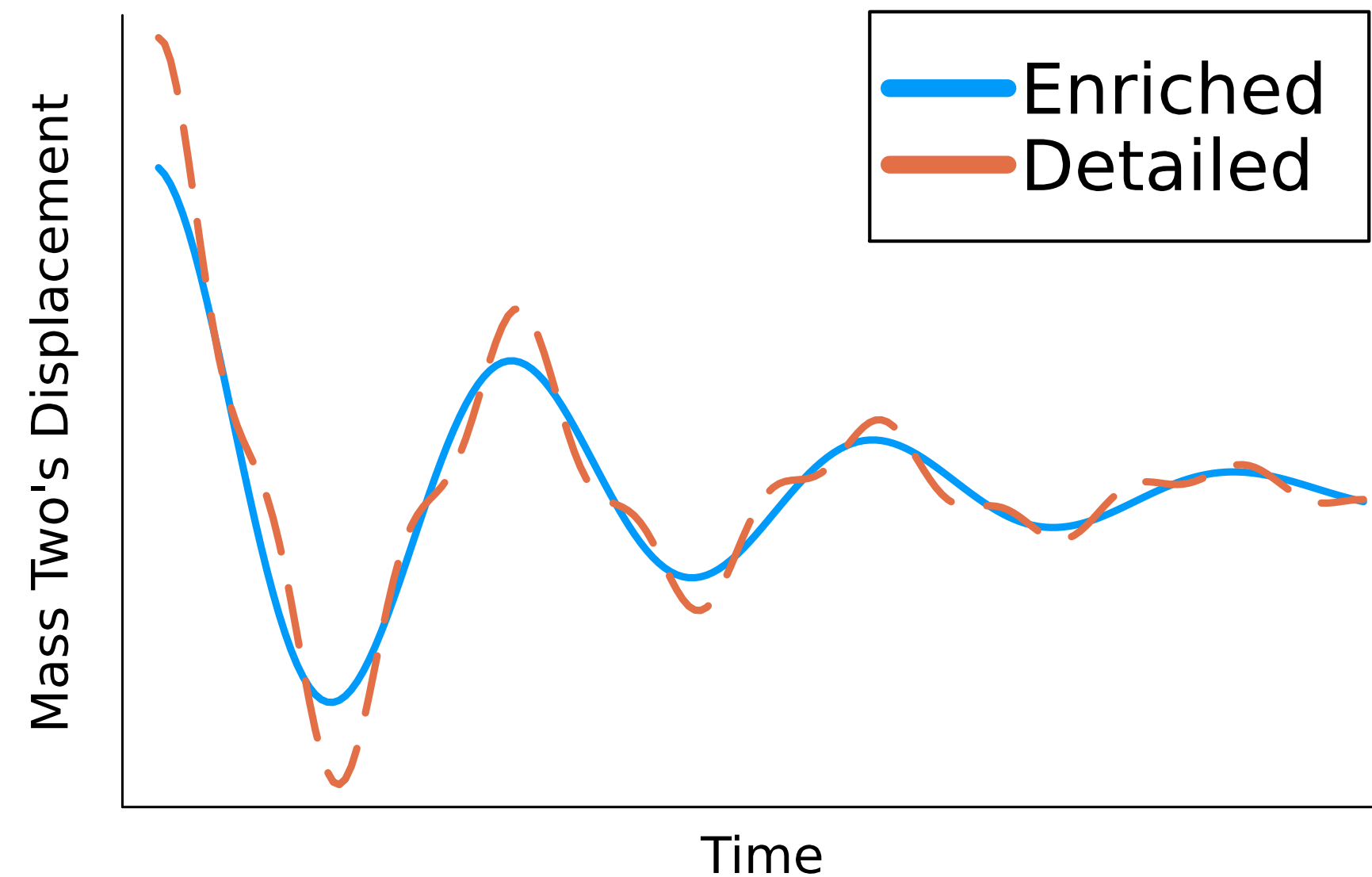
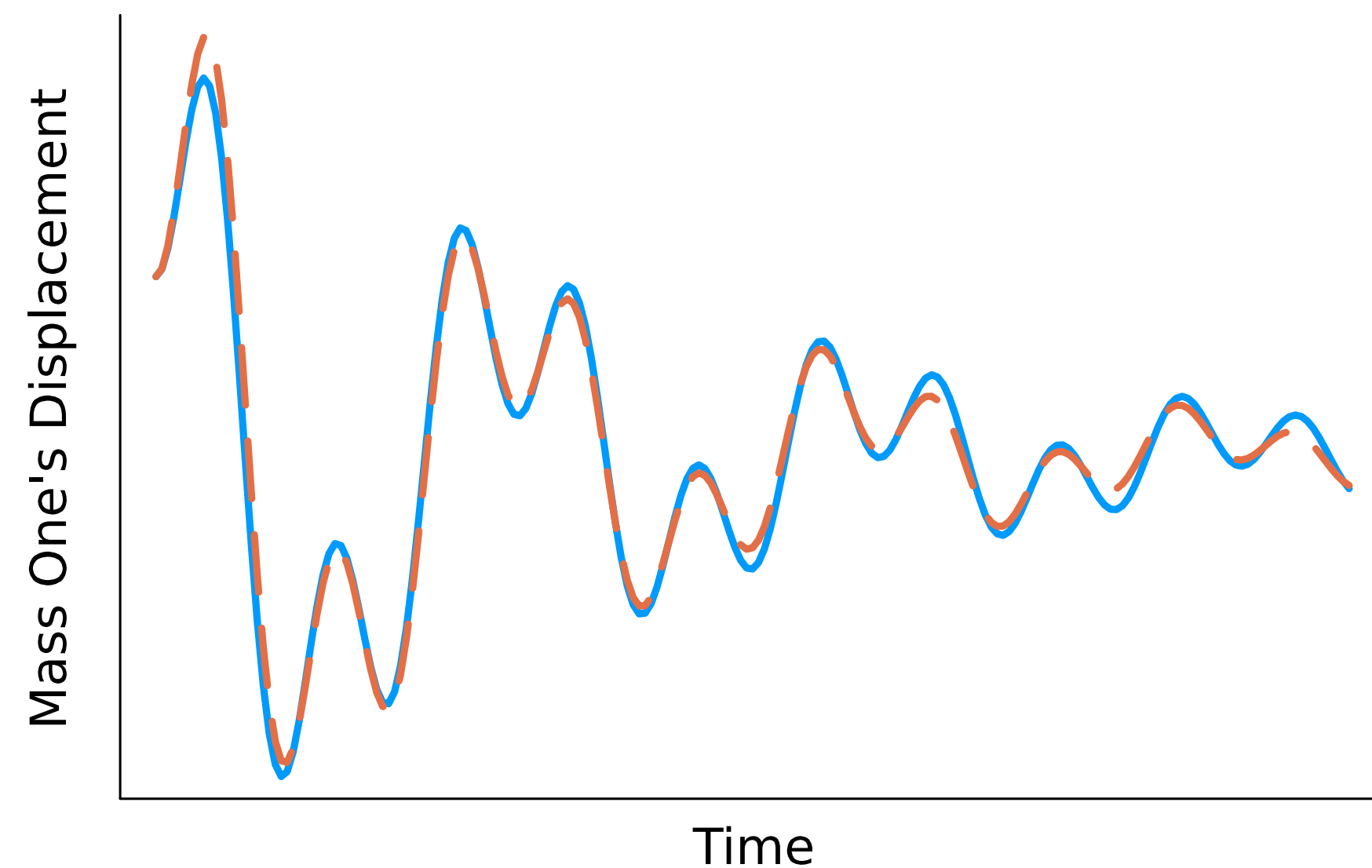
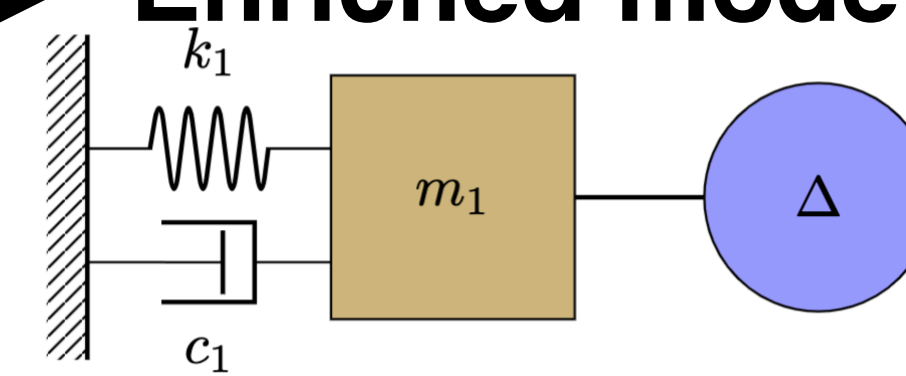
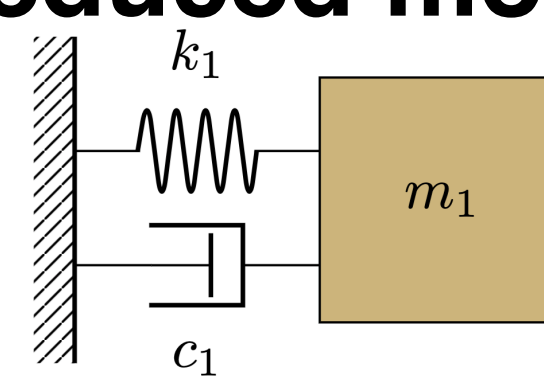
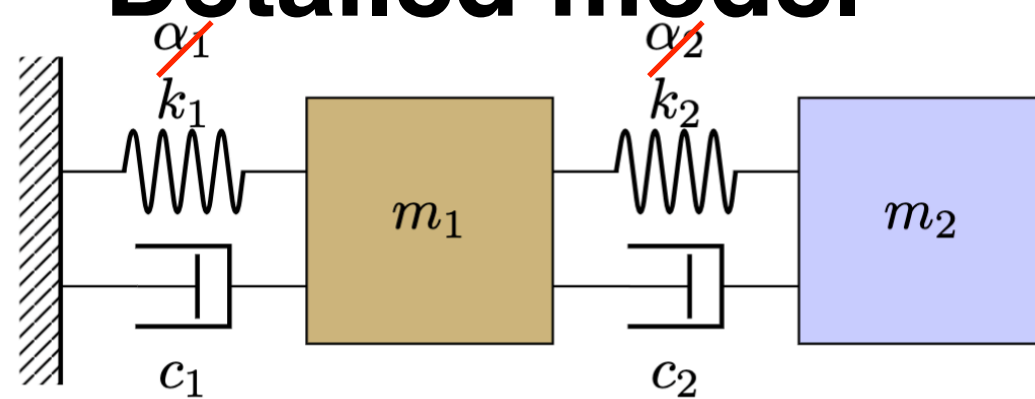
# How does the degree of nonlinearity impact model discrepancy?

$$f_{s,i}(t) = \left[ \underbrace{-k_i(x_i(t) - x_{i-1}(t)) + k_{i+1}(x_{i+1}(t) - x_i(t))}_{\text{Hooke's law}} \right] \left( 1 + \underbrace{\alpha_i(x_{i+1}(t) - x_{i-1}(t))}_{\text{nonlinear term}} \right)$$



# The enriched model closely matches the **linear** detailed trajectories

Detailed model → Reduced model → Enriched model

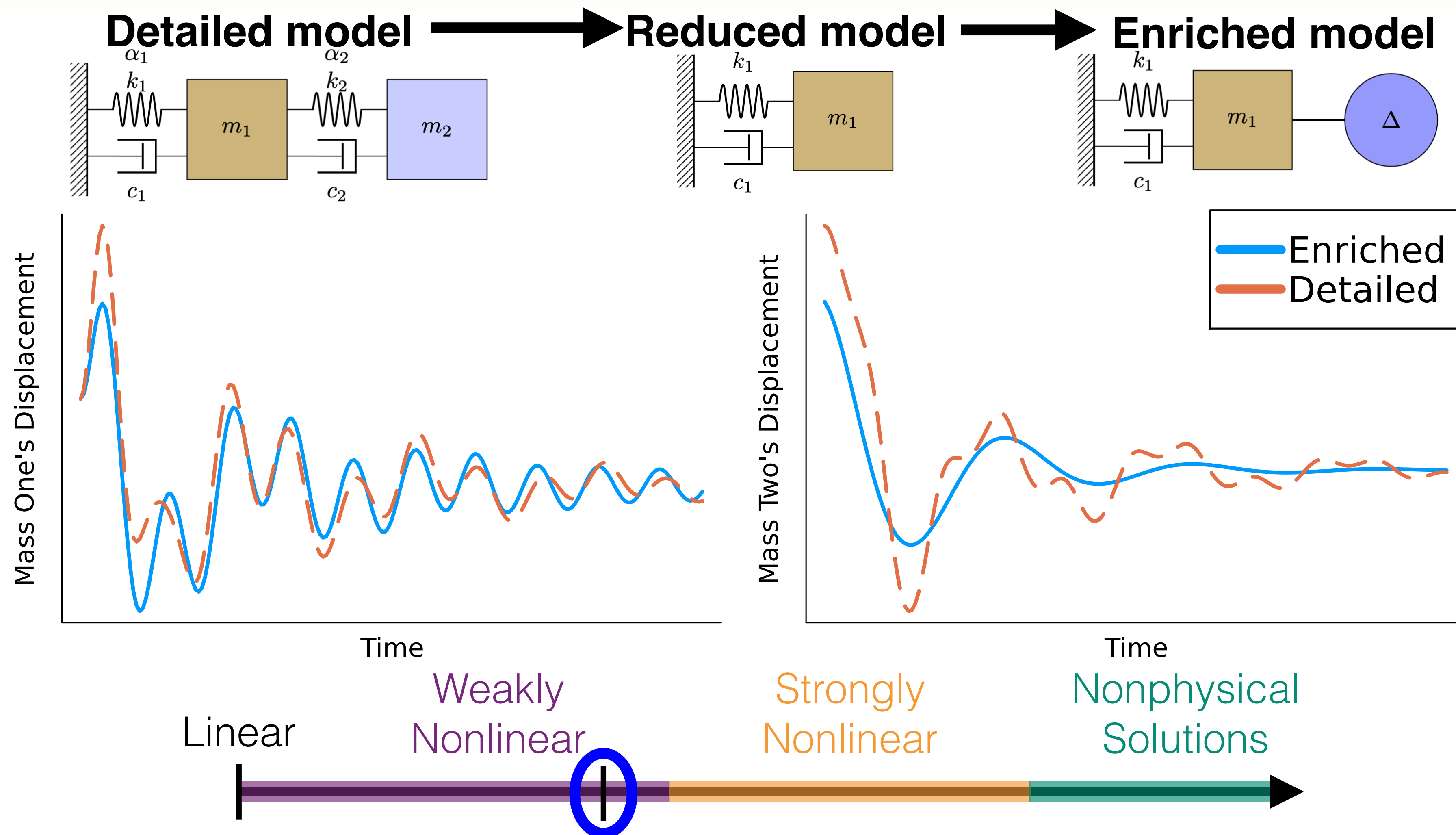


Enriched model predictions are using MAP estimates of model parameters.

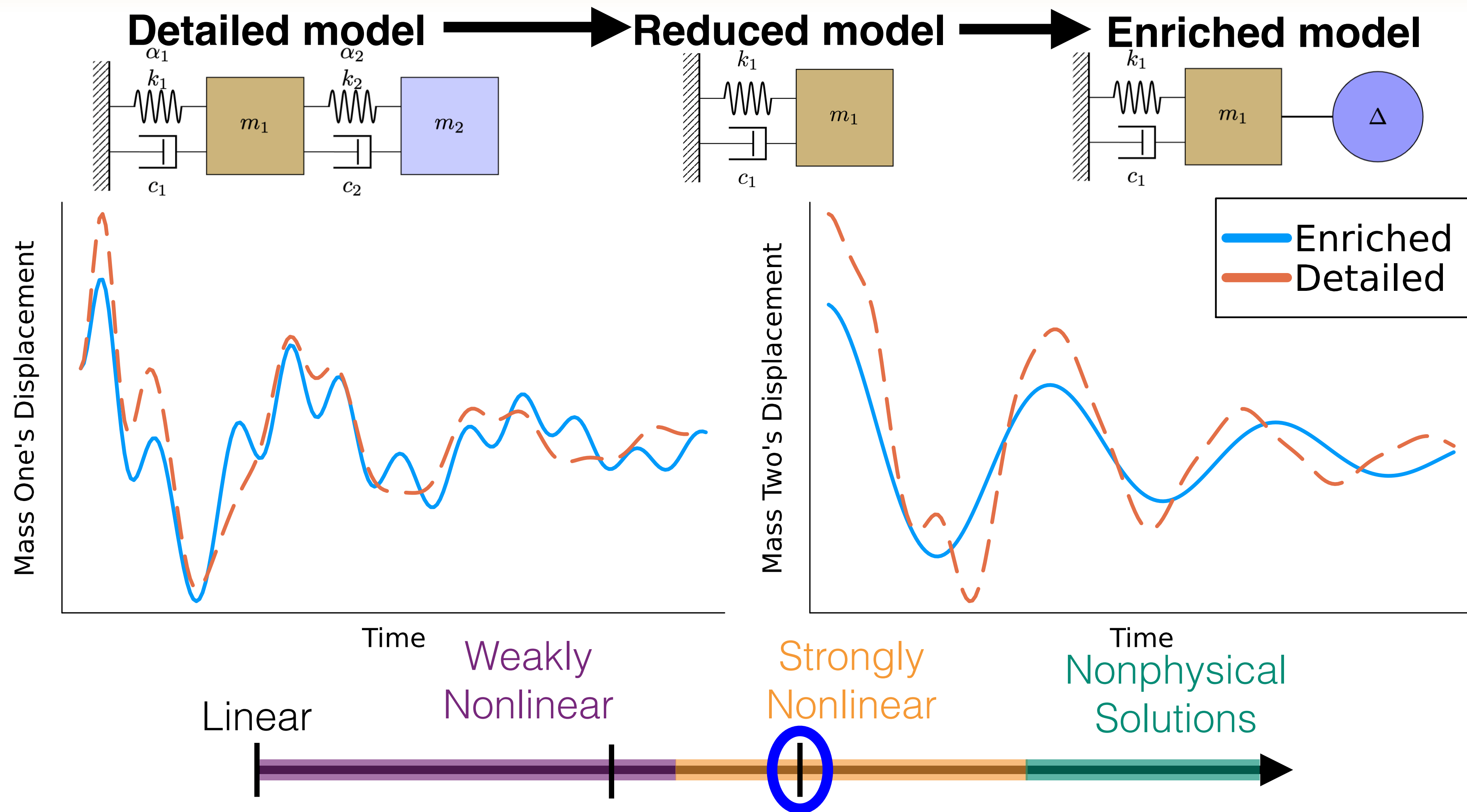




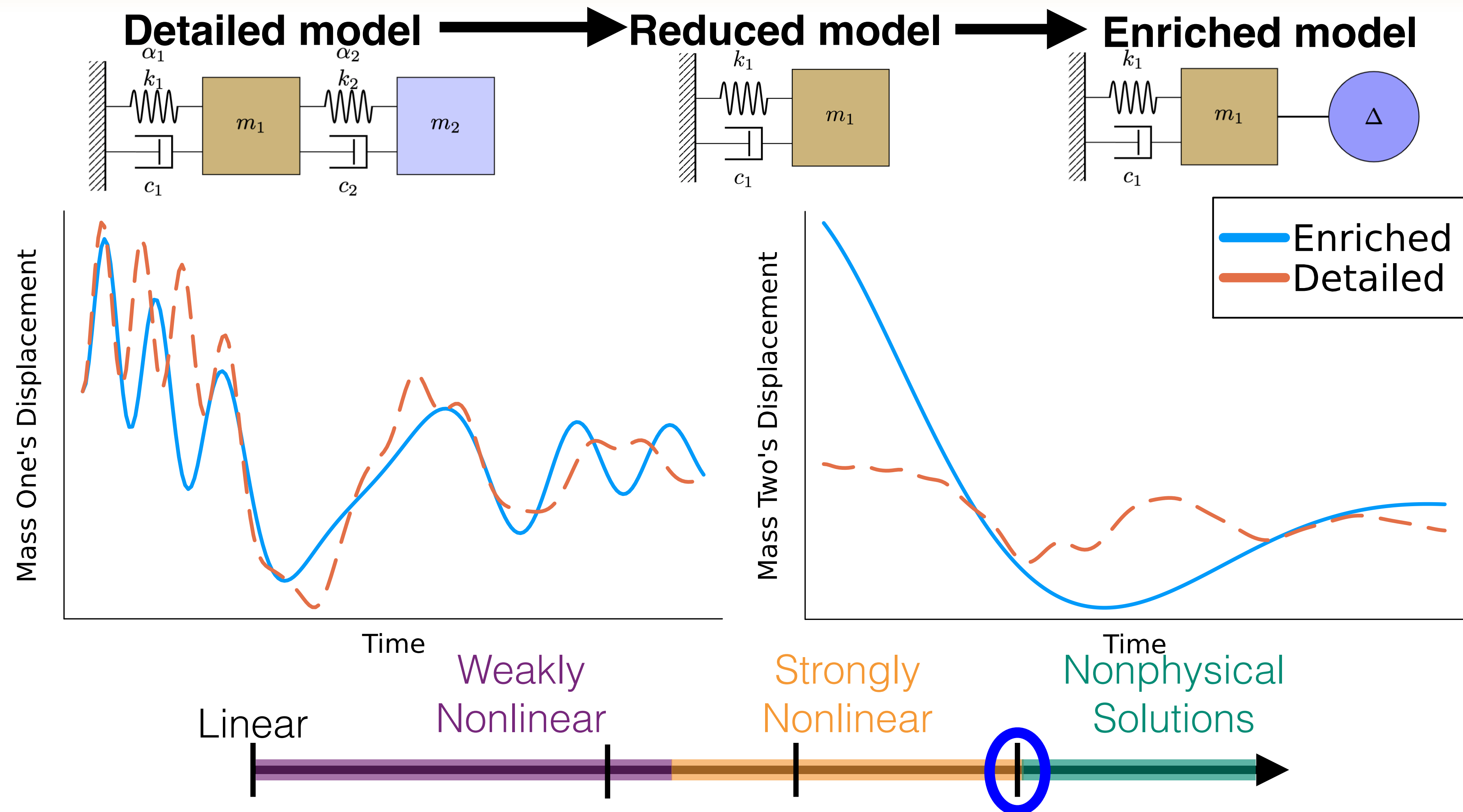
The enriched model overestimates the dampening of mass two for the **weakly nonlinear** model



The enriched model overestimates the dampening and frequency of mass two for the **stronger nonlinear** model



The enriched model overestimates the initial amplitude of mass two for the **strongest nonlinear** model

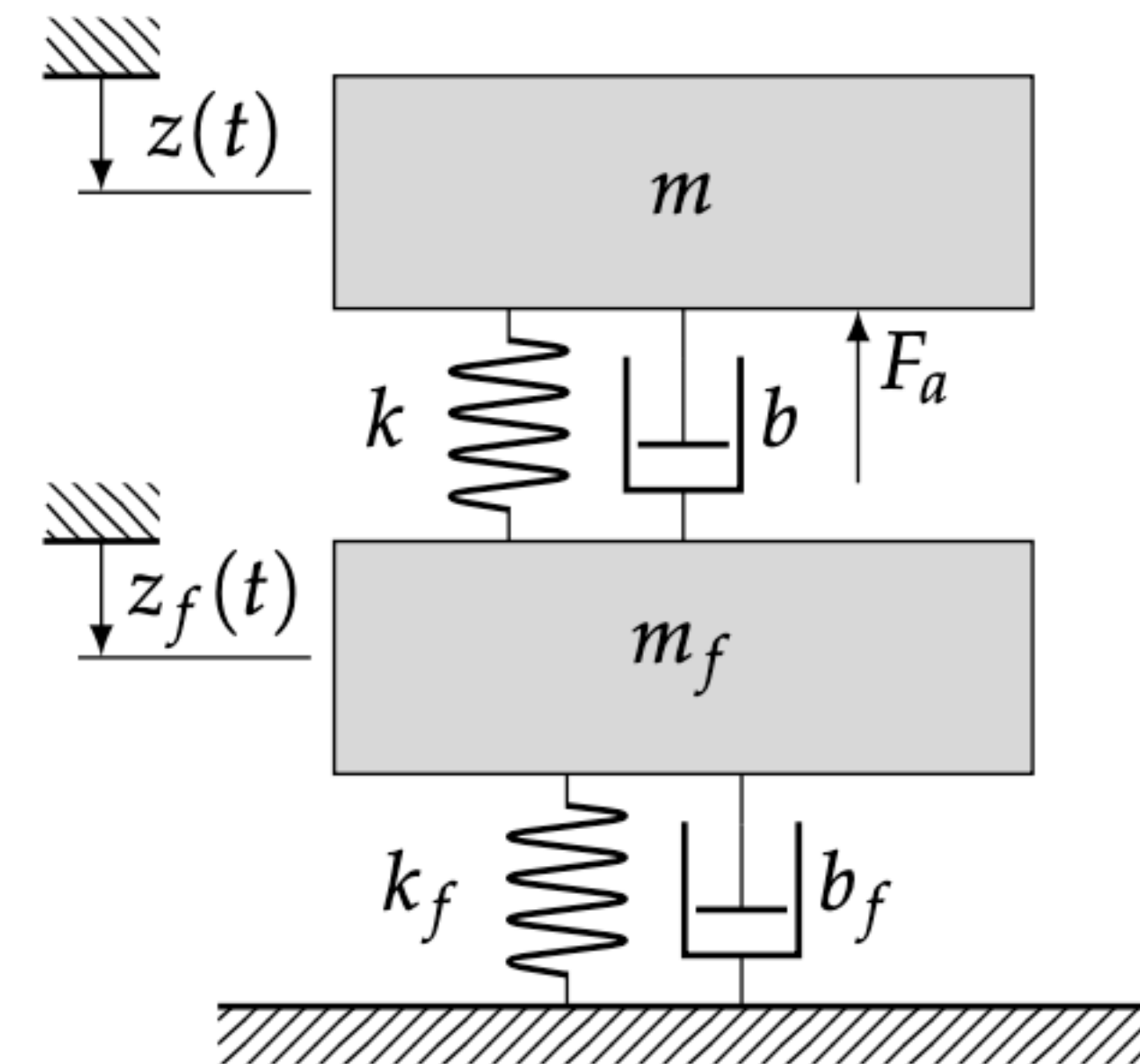
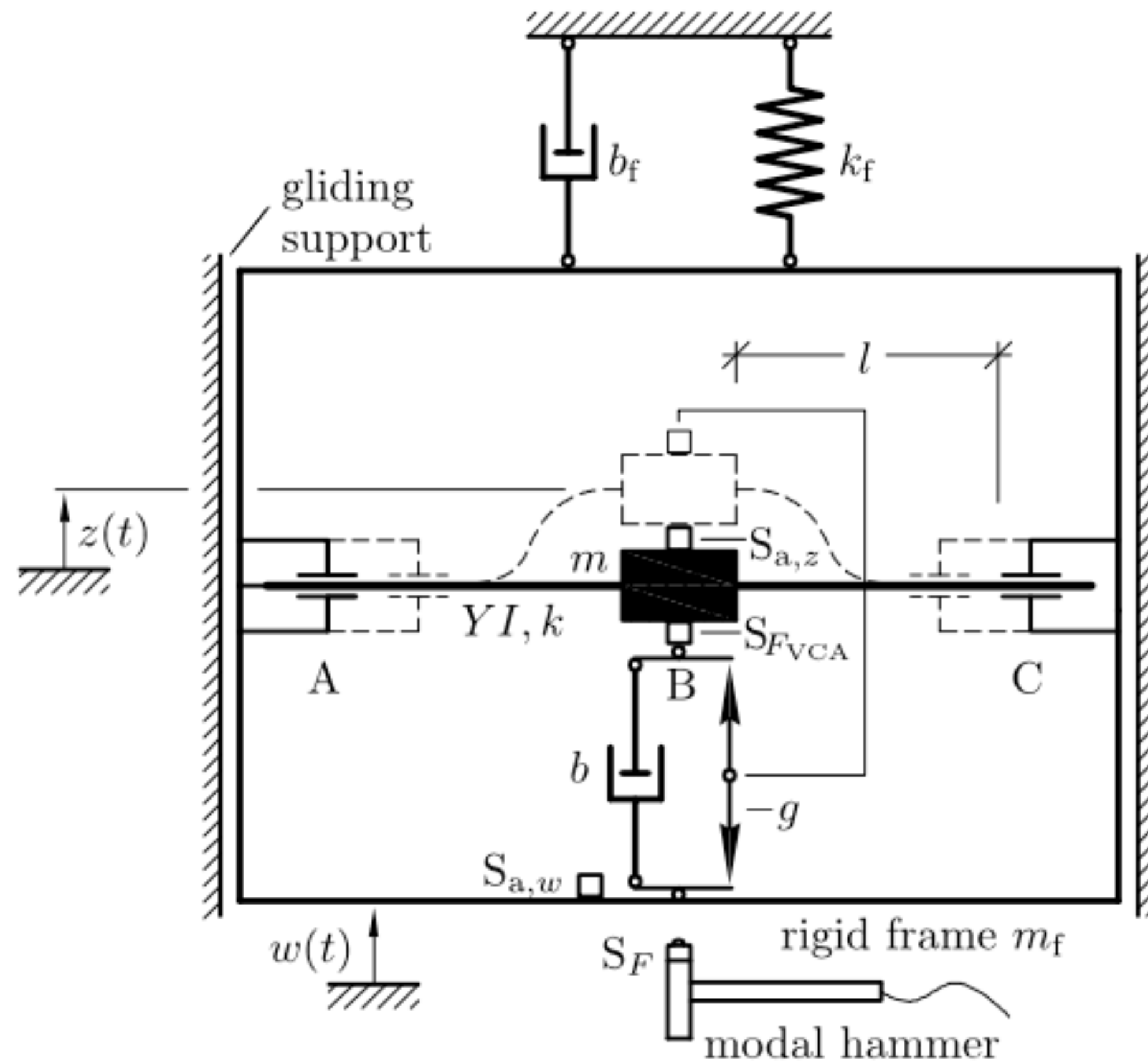


# Case 2

Experiment

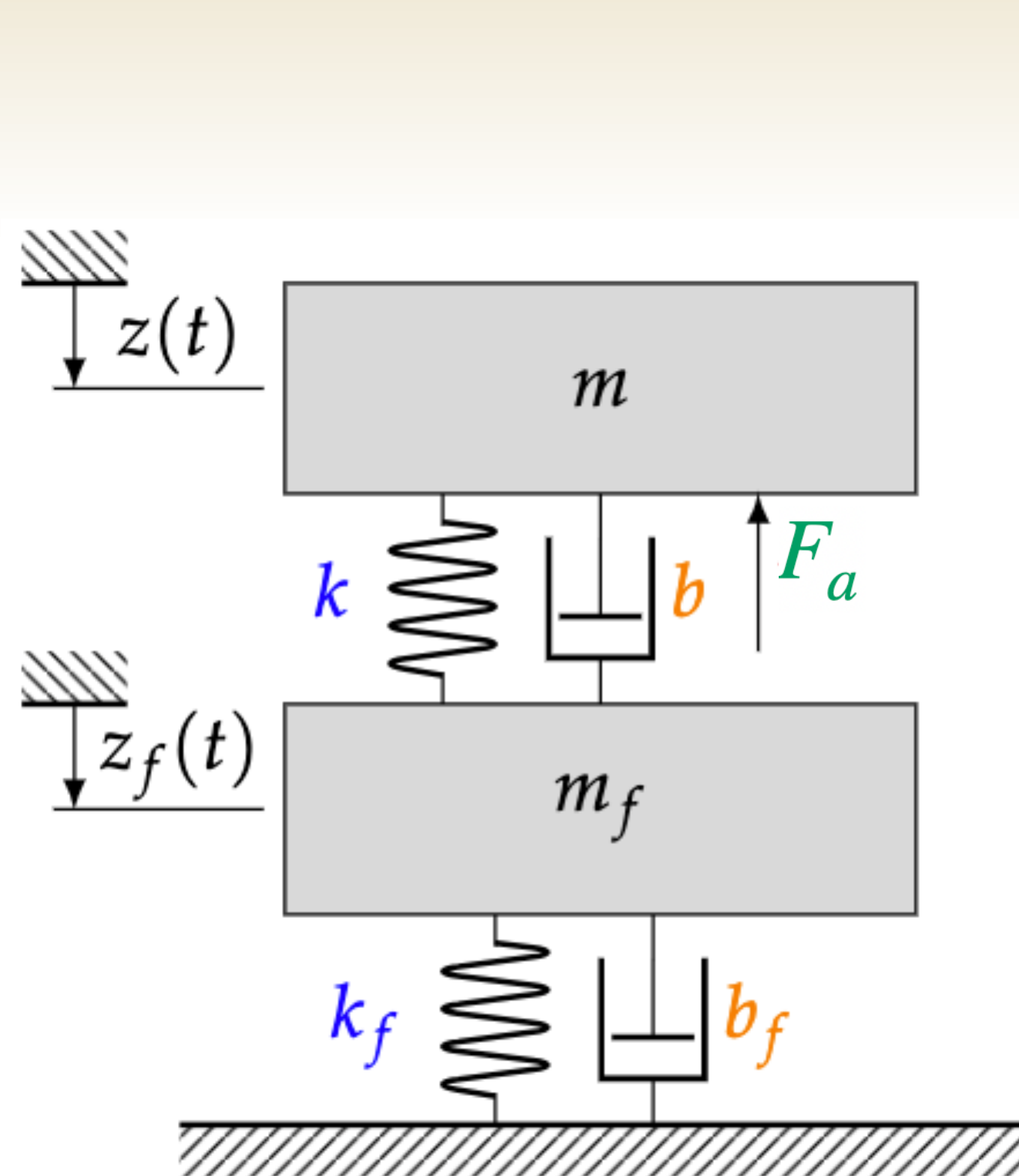


Low-fidelity  
model





The low-fidelity model is a two-mass oscillator with ***linear*** forces



$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - F_a \left( \dot{z}(t) \right) - k \left( z(t) - z_f(t) \right) \right\}$$

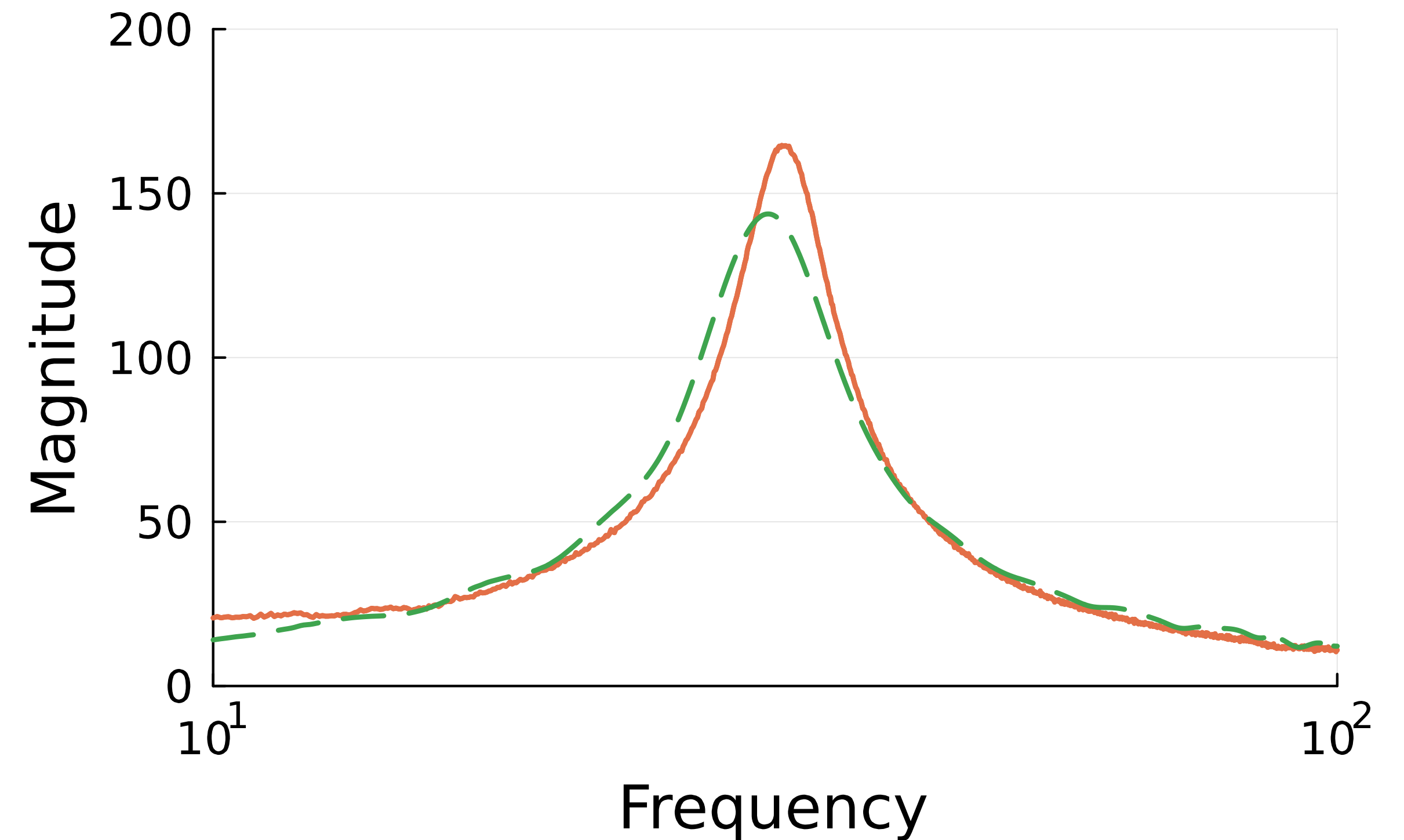
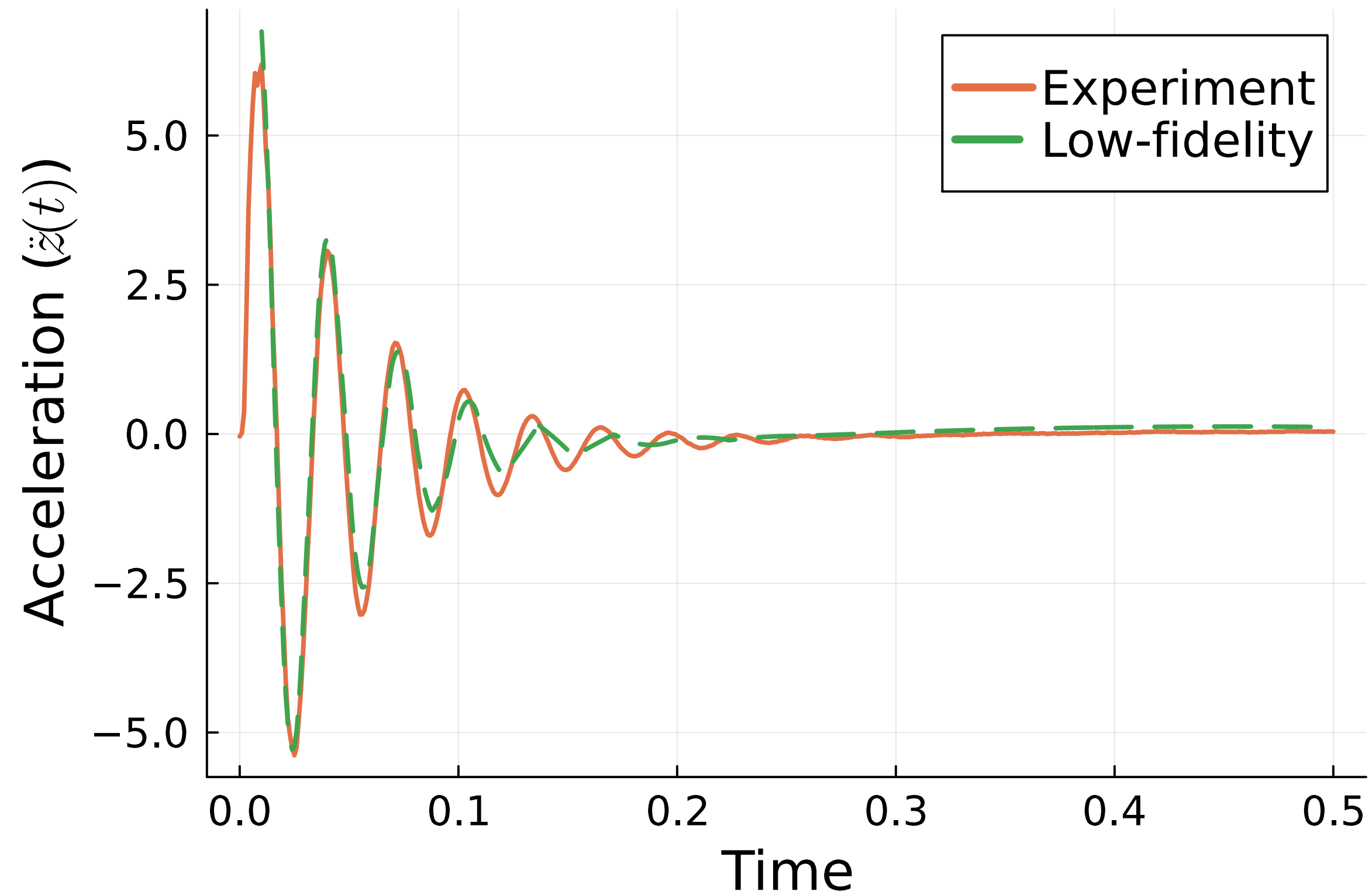
$$\ddot{z}_f(t) = \frac{1}{m_f} \left\{ b \left( \dot{z}(t) - \dot{z}_f(t) \right) + k \left( z(t) - z_f(t) \right) - b_f \dot{z}_f - k_f z_f \right\}$$

Spring force

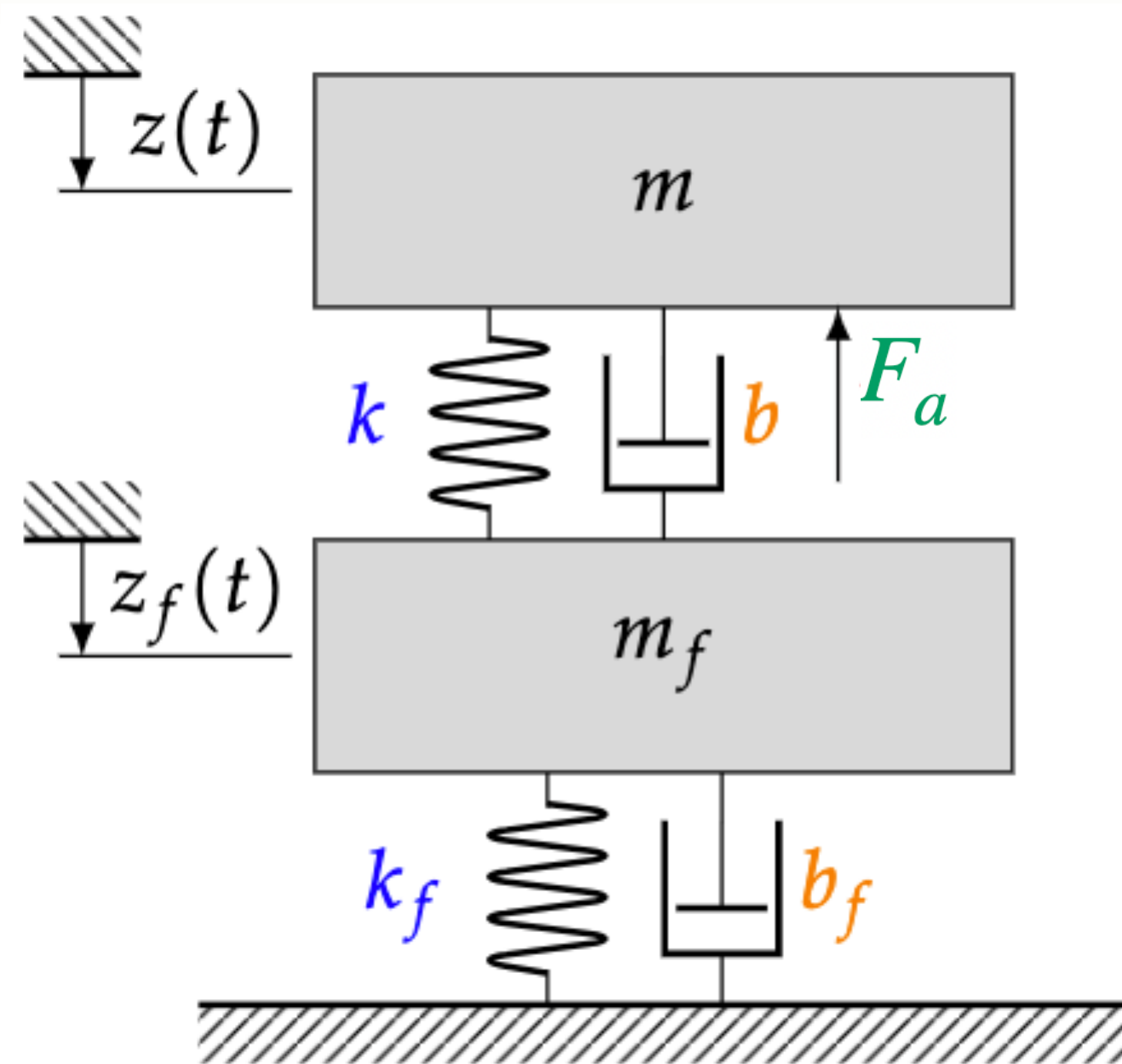
Passive damping force

Active damping force

# Assumed *linear* forces cause error in the low-fidelity model



# An enrichment operator is embedded into the low-fidelity model

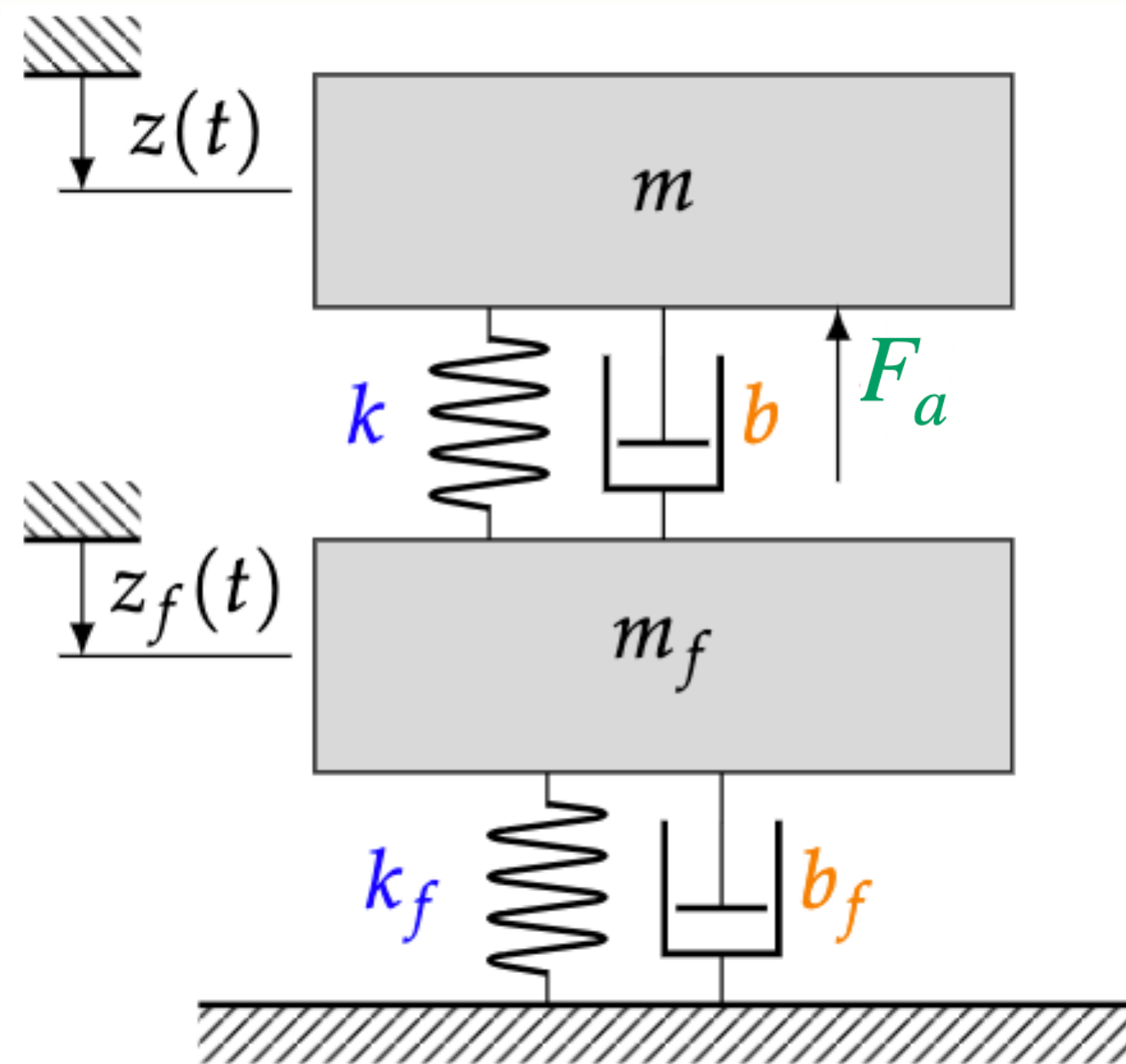


$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - \underbrace{F_a \left( \dot{z}(t) \right)}_{g_0 \dot{z}(t) + g_1 |\dot{z}(t)| \dot{z}(t) + g_2 \dot{z}(t)^3} - k \left( z(t) - z_f(t) \right) \right\}$$

Enrichment operator

$$\ddot{z}_f(t) = \frac{1}{m_f} \left\{ b \left( \dot{z}(t) - \dot{z}_f(t) \right) + k \left( z(t) - z_f(t) \right) - b_f \dot{z}_f - k_f z_f \right\}$$

# The enriched model is a two-mass oscillator with *nonlinear* active damping



$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - \underbrace{F_a \left( \dot{z}(t) \right)}_{\text{Enrichment operator}} - k \left( z(t) - z_f(t) \right) \right\}$$

$g_0 \dot{z}(t) + g_1 |\dot{z}(t)| \dot{z}(t) + g_2 \dot{z}(t)^3$

- Model parameters:  $\boldsymbol{\theta} = (g_1, g_2)$

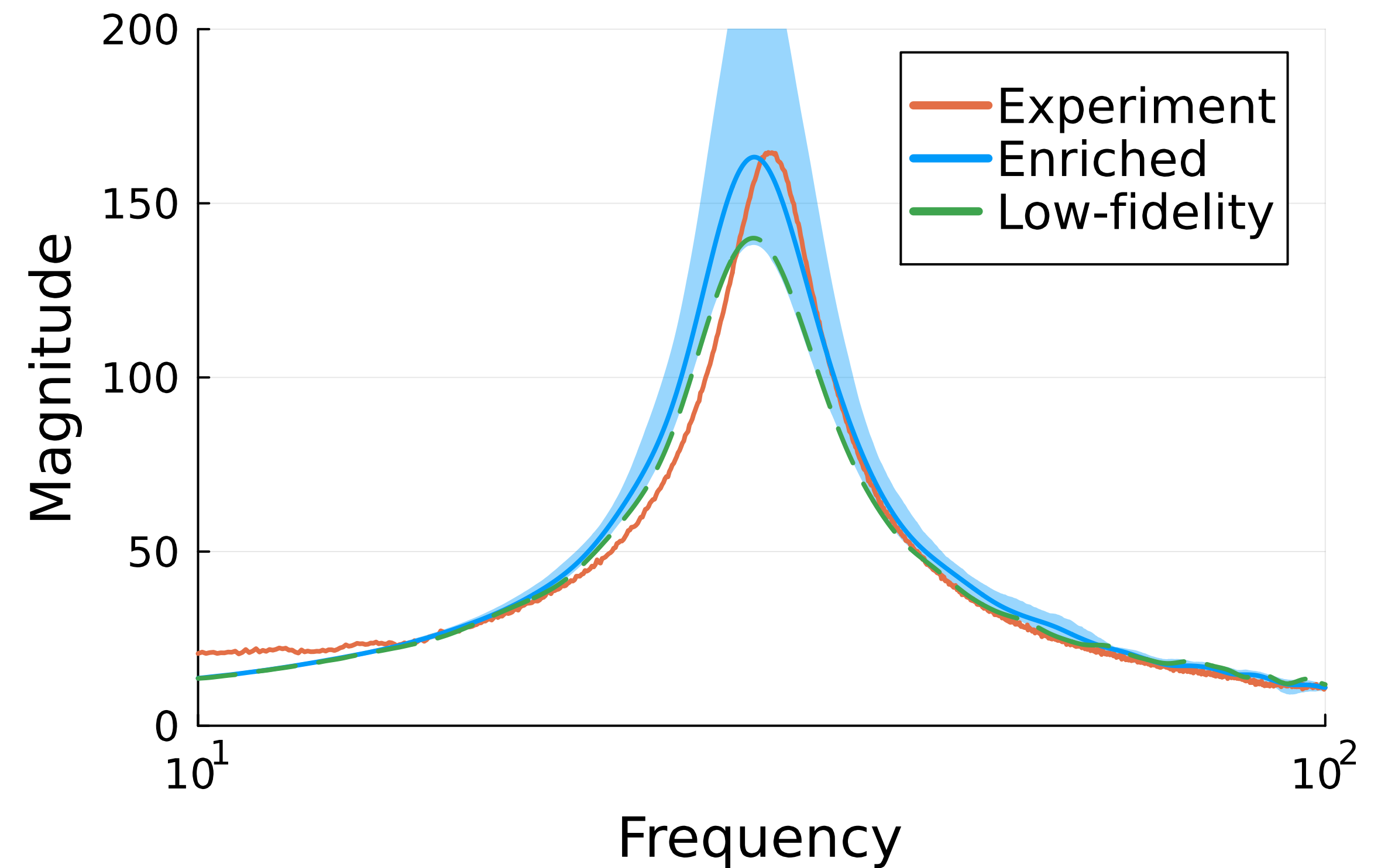
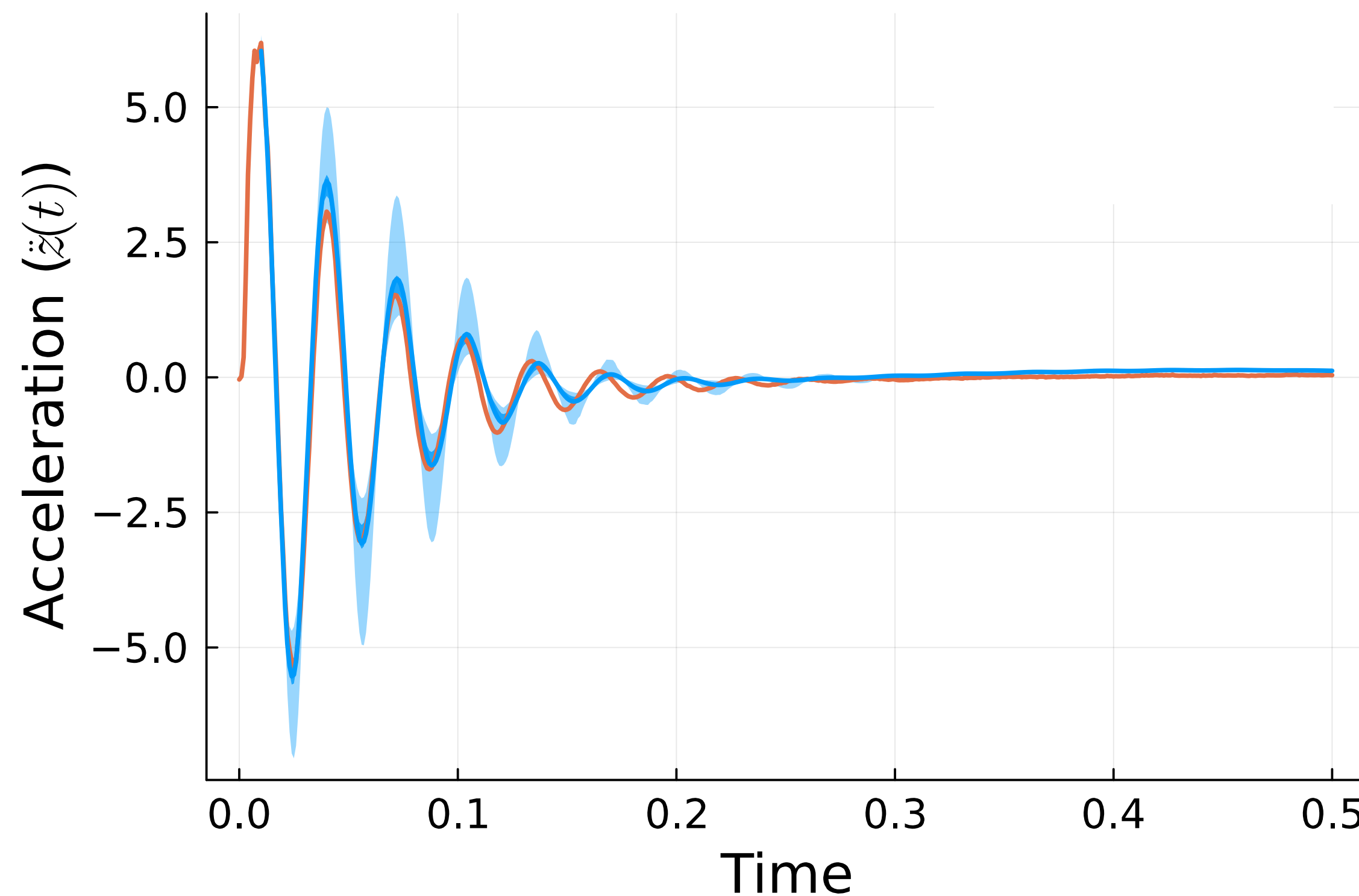
$$g_i \sim \mathcal{N}(\mu_i, \sigma_i^2), \text{ where } \mu_i \in \mathbb{R}, \sigma_i \in \mathbb{R}_{\geq 0}, \text{ and } i = \{1, 2\}$$

- Hyperparameters:  $\boldsymbol{\phi} = (\boldsymbol{\phi}_1, \boldsymbol{\phi}_2)$ , where  $\boldsymbol{\phi}_i = (\mu_i, \sigma_i)$

- We use hierarchical Bayesian calibration to sample a posterior on  $\boldsymbol{\phi}$



# The enriched model covers most experimental observations



[5] R. BANDY, T. PORTONE, AND R. MORRISON, *Stochastic model correction for the adaptive vibration isolation round-robin challenge, Model Validation and Uncertainty Quantification*, Volume 3, Conference Proceedings of the Society for Experimental Mechanics Series, (to be released).

# Conclusions

- Mass-spring-damper models illustrate model-form error that can arise in many structural dynamics applications.
- Expert knowledge about a potential source of model-form error informs the enrichment operators.
- Enriched models decrease discrepancies and retains interpretability.

# Questions

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