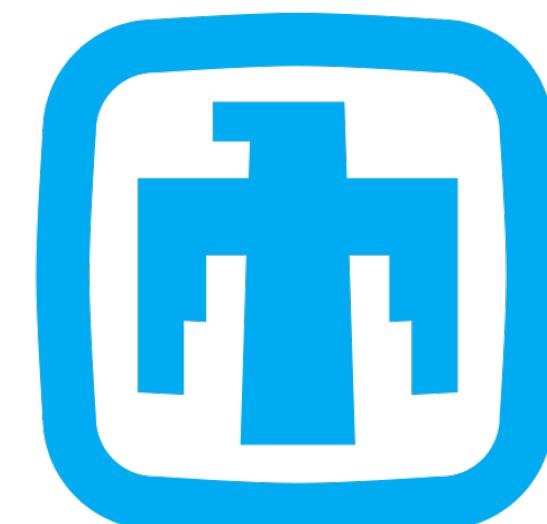


# Nonlinear Forces and Omitted Masses: Mass-Spring-Damper Models and Their Model-Form Errors

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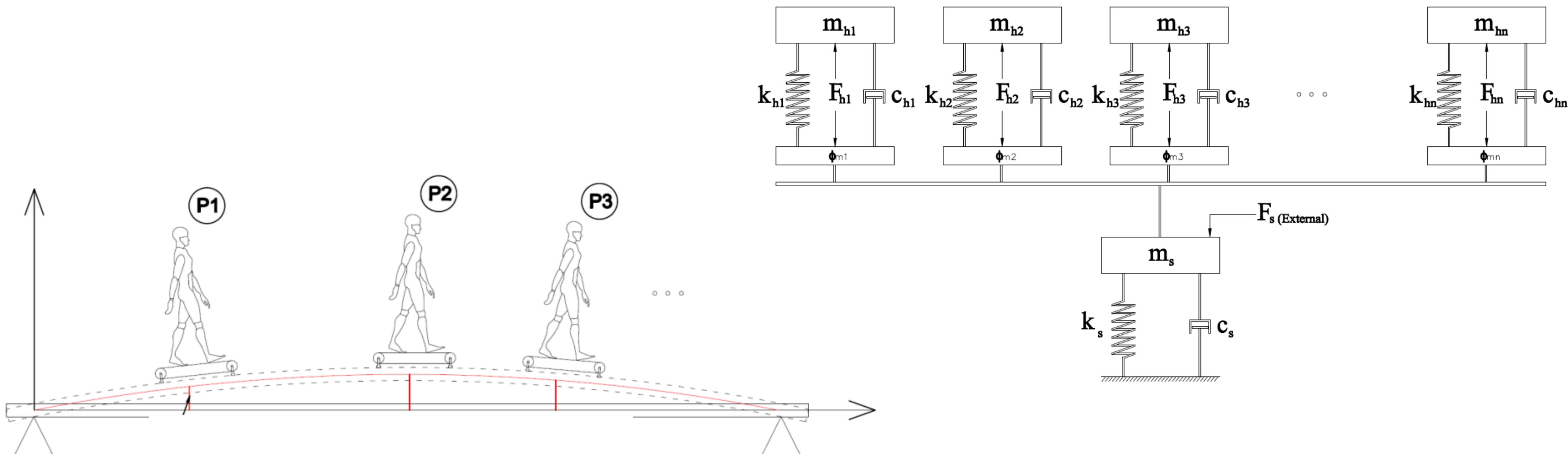
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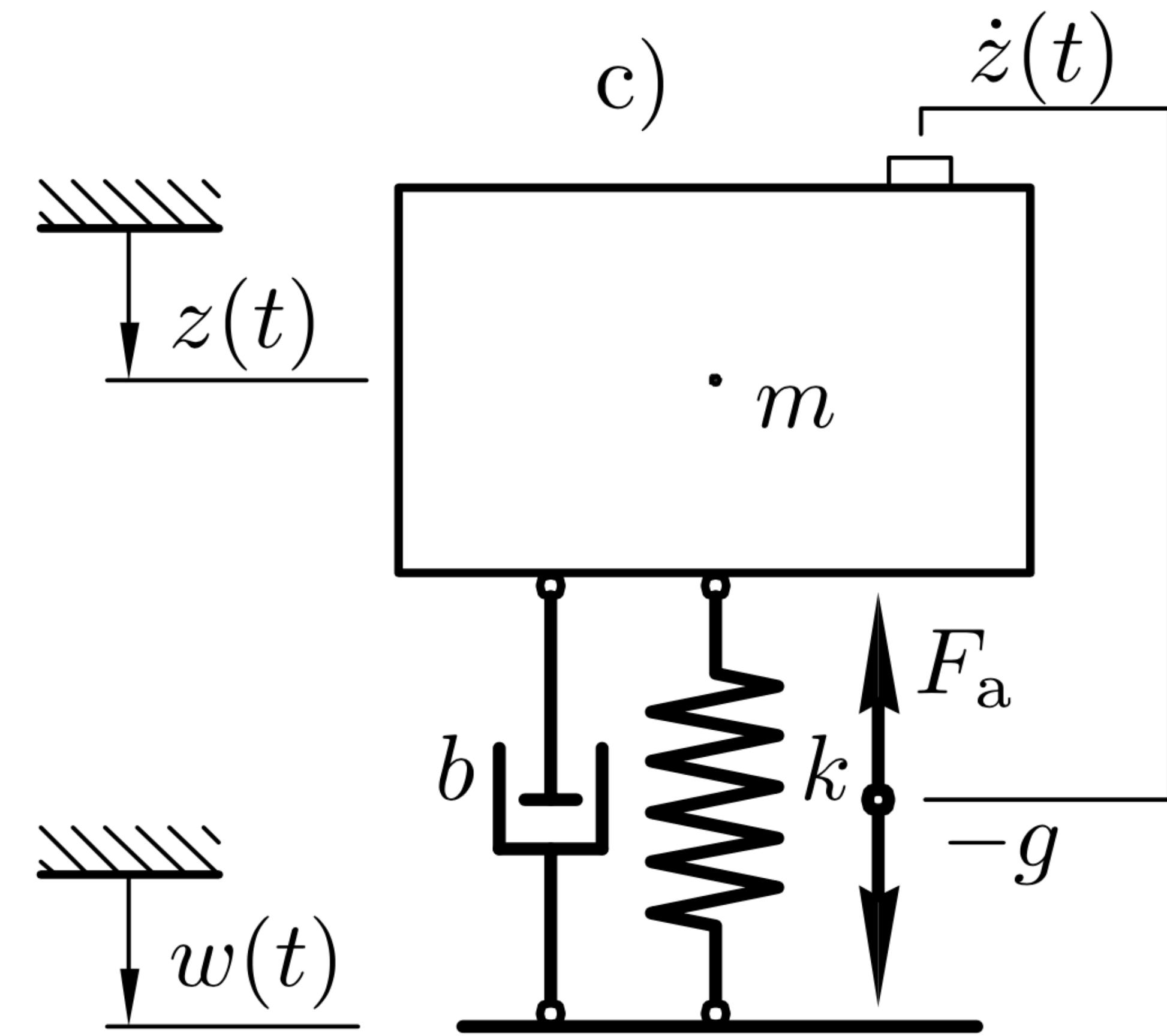
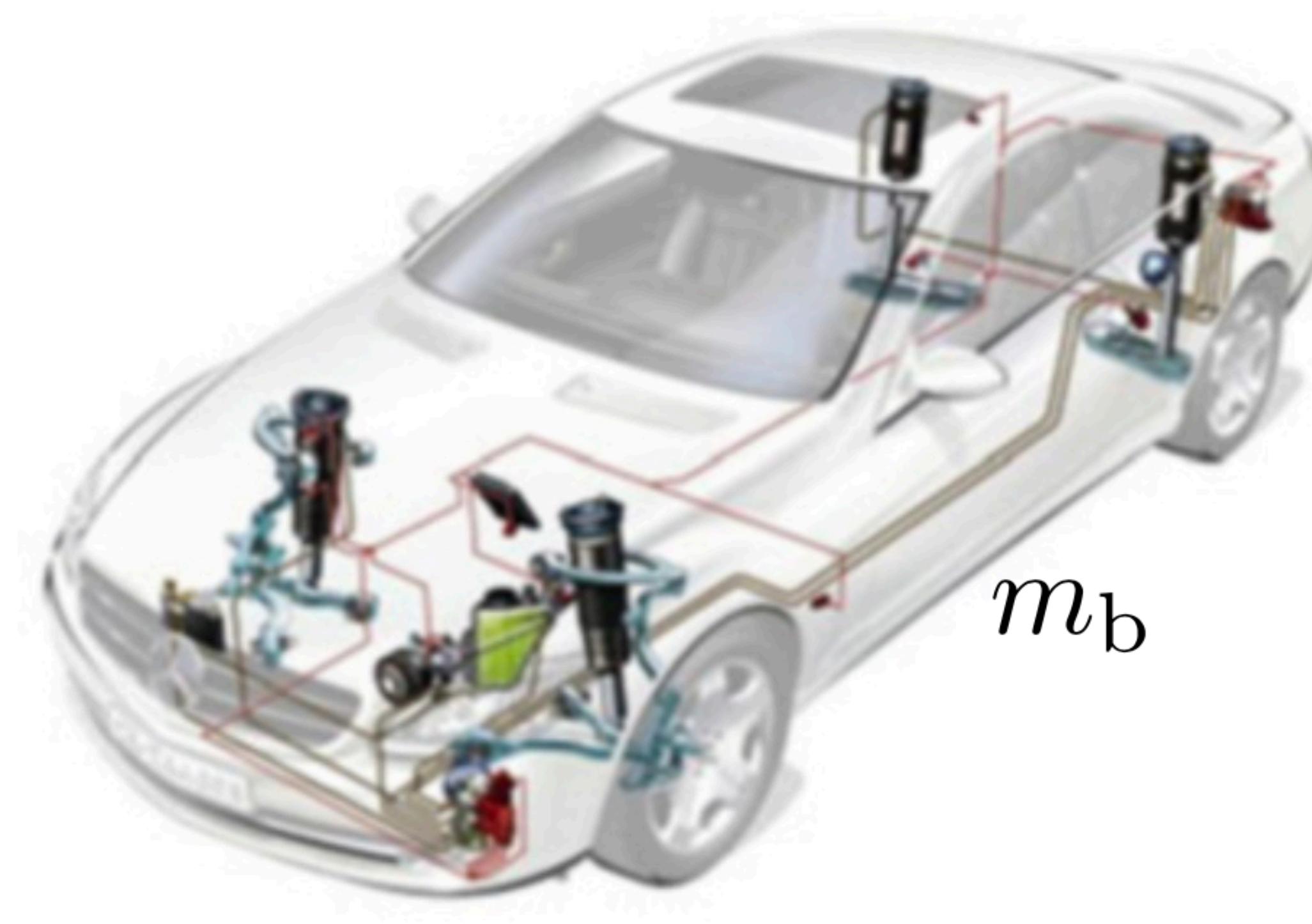
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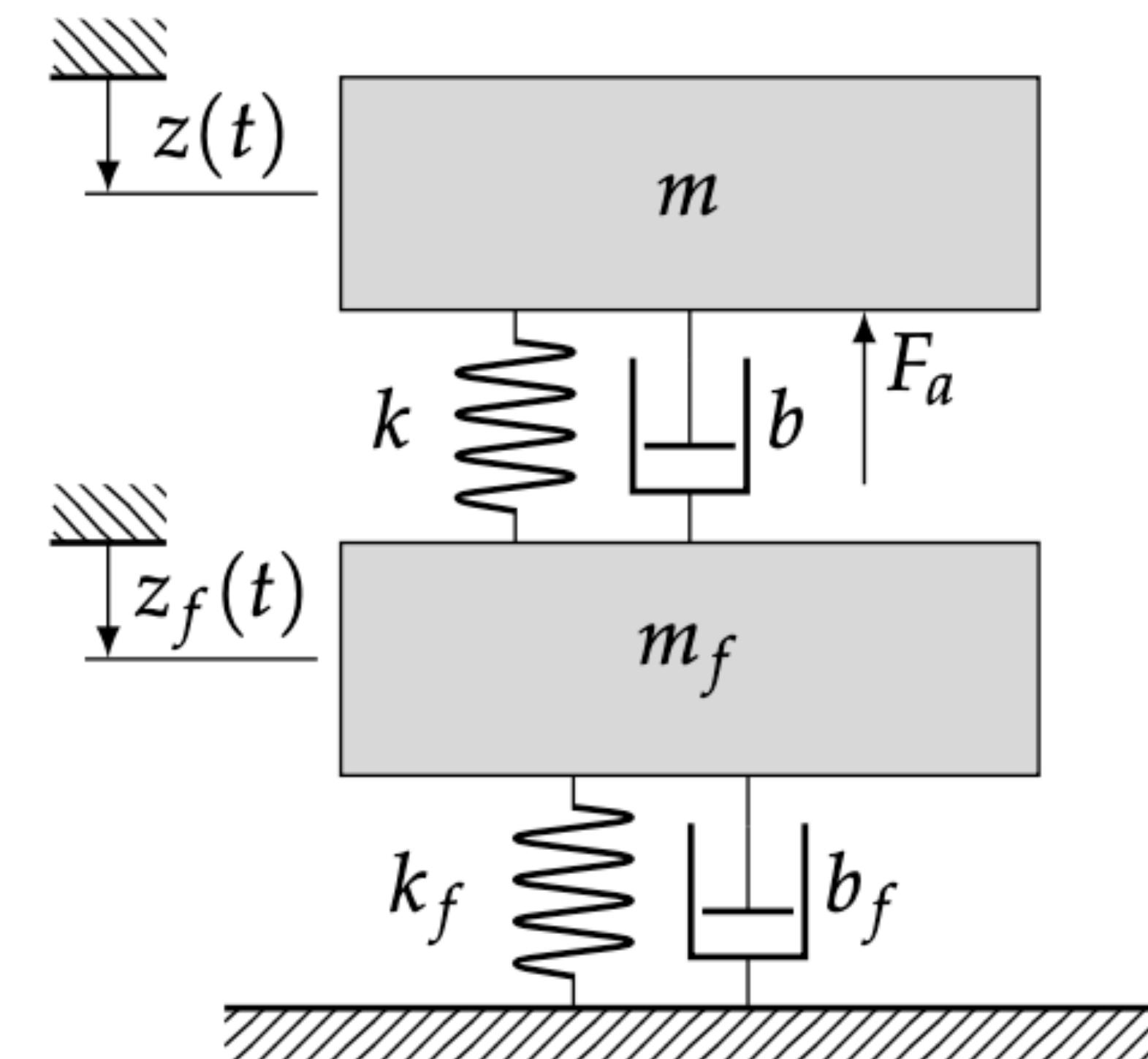
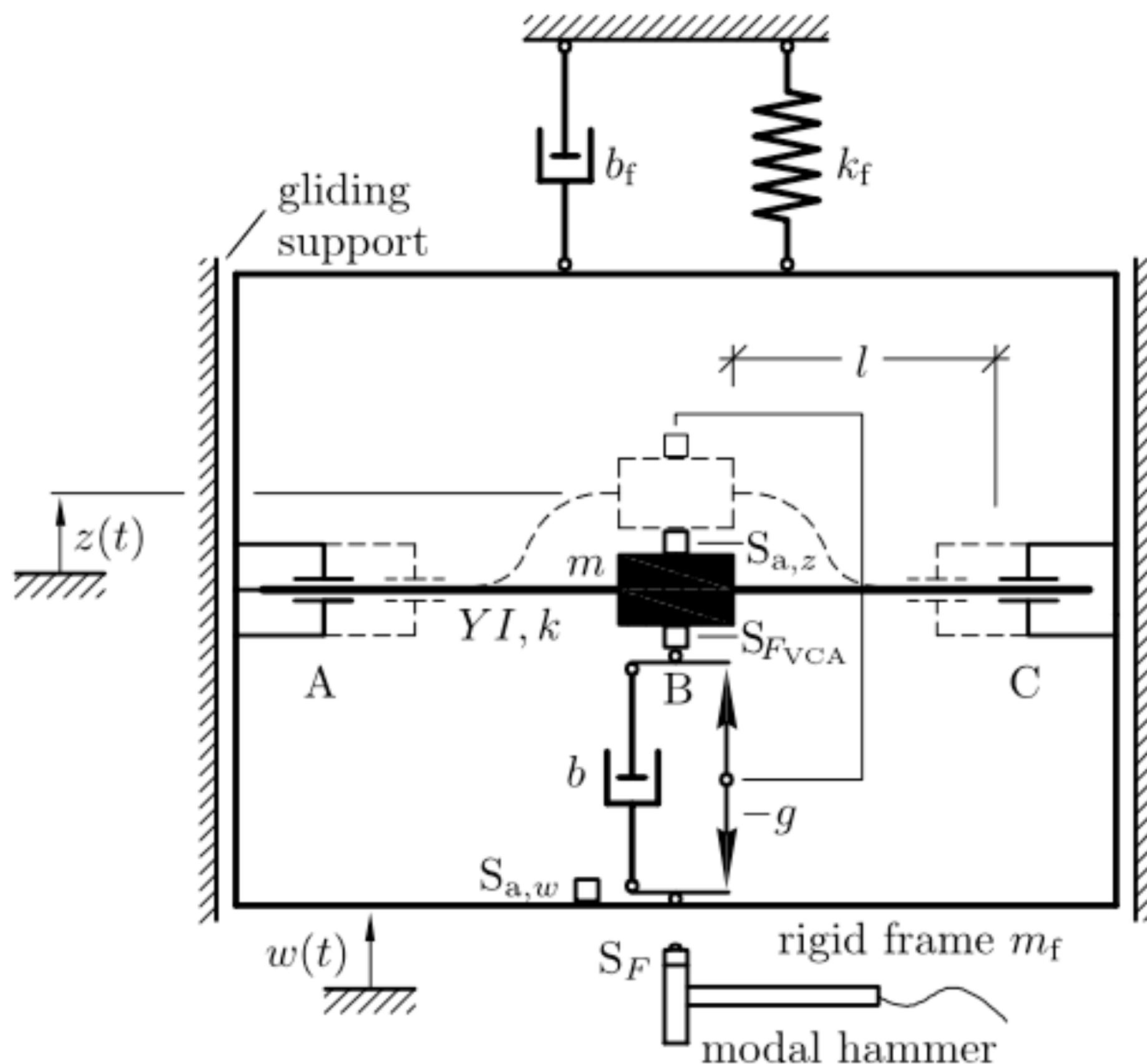
# Mass-spring-damper models approximate the impact load on bridges



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

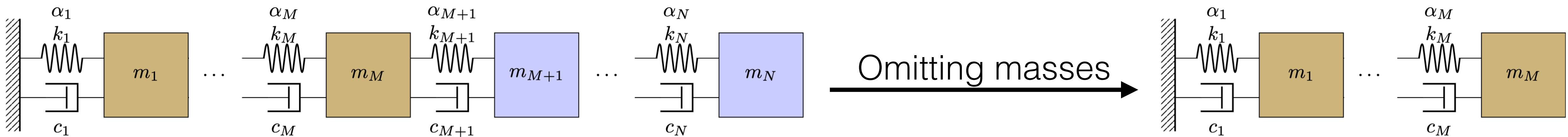


# A two mass oscillator approximates the vibration isolation experiment

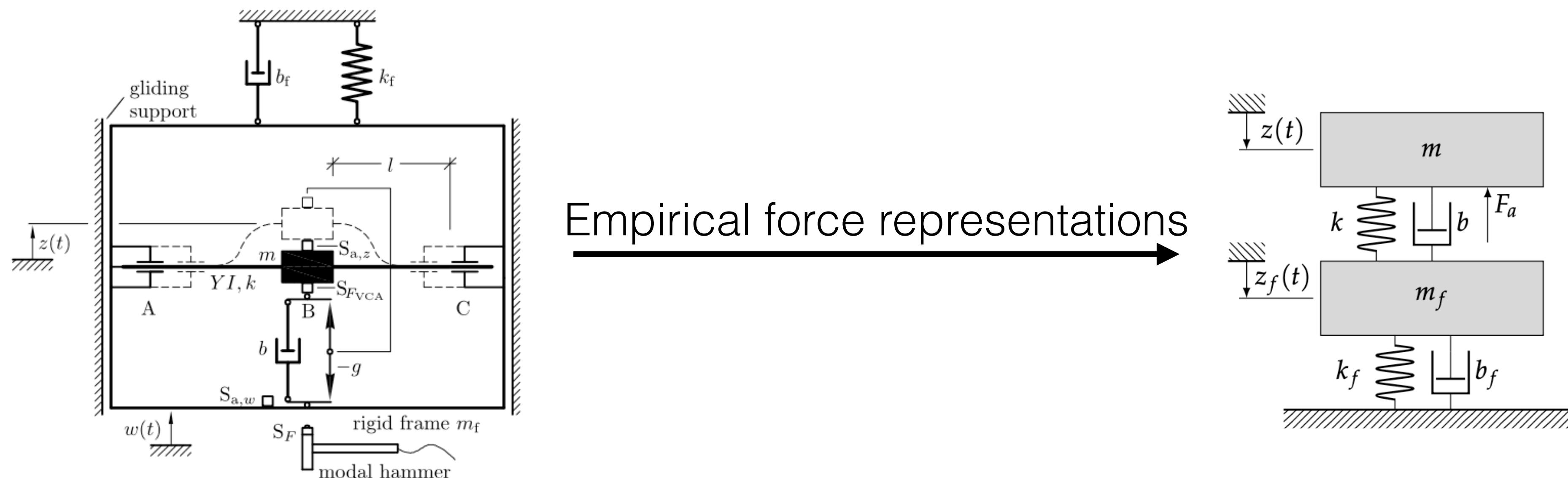


# How do model-form errors arise?

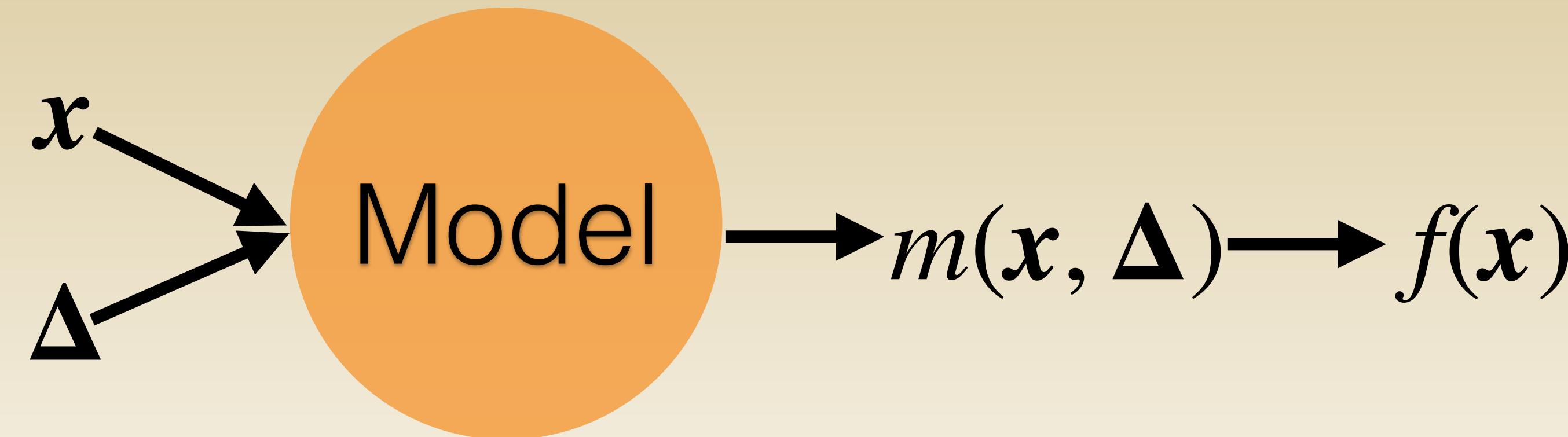
## Case 1



## Case 2

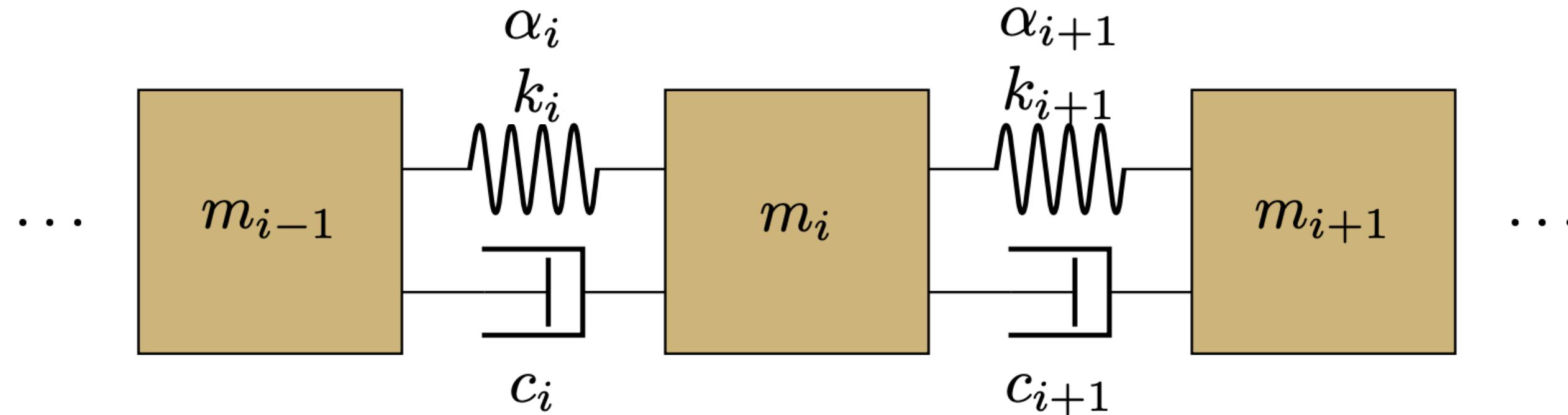


Embedded model enrichments reduce the model-form error and enable extrapolative predictions



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

# Case 1: forming mass-spring-damper models



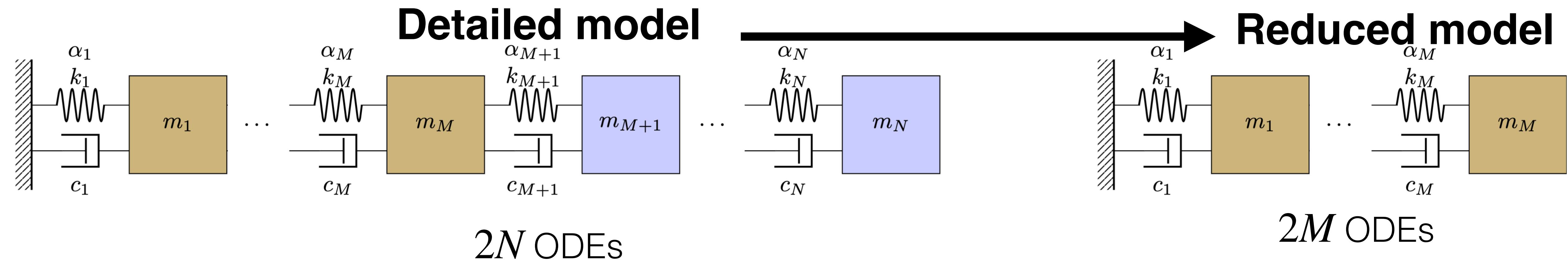
Newton's Second Law of Motion:

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

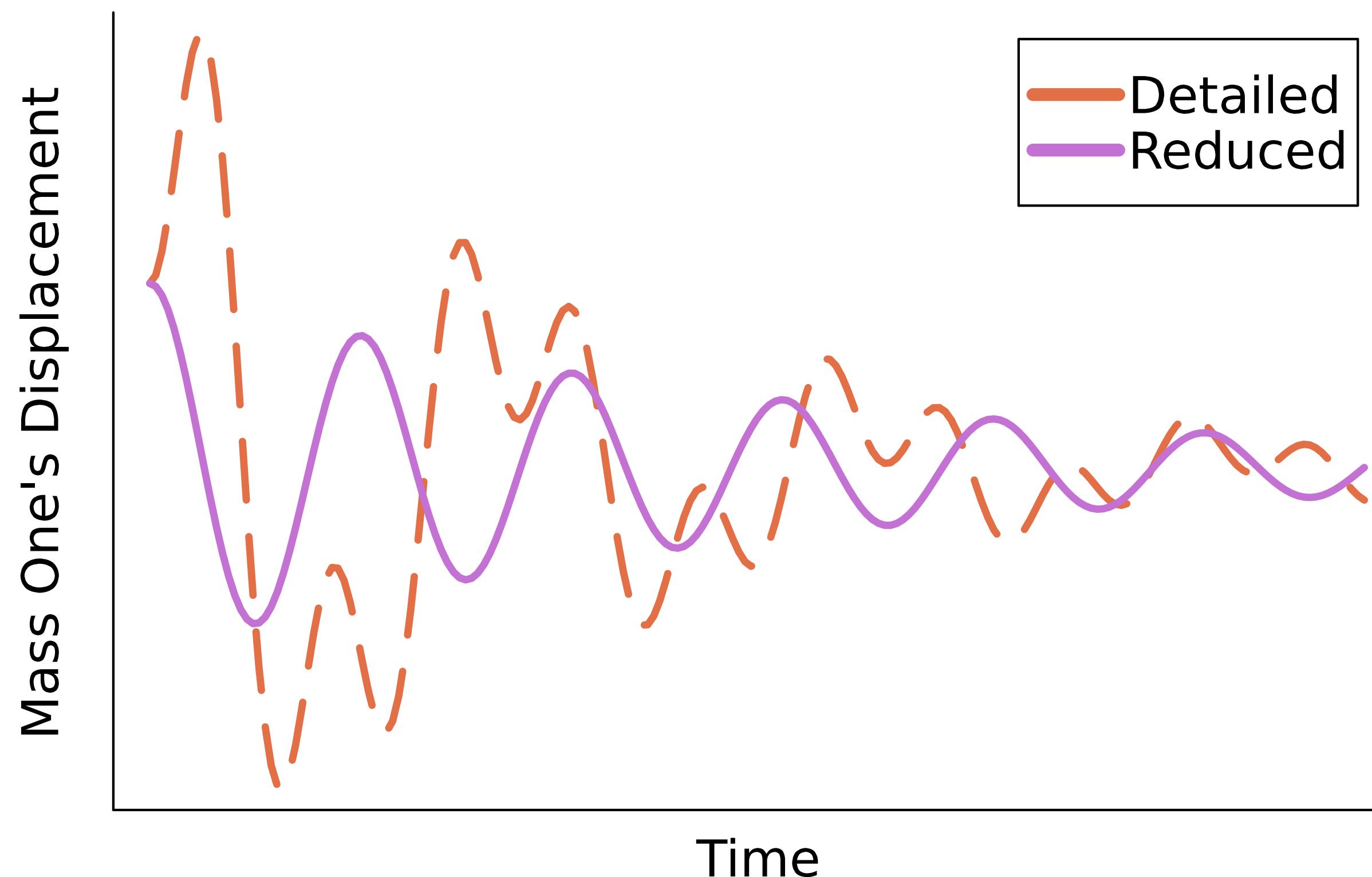
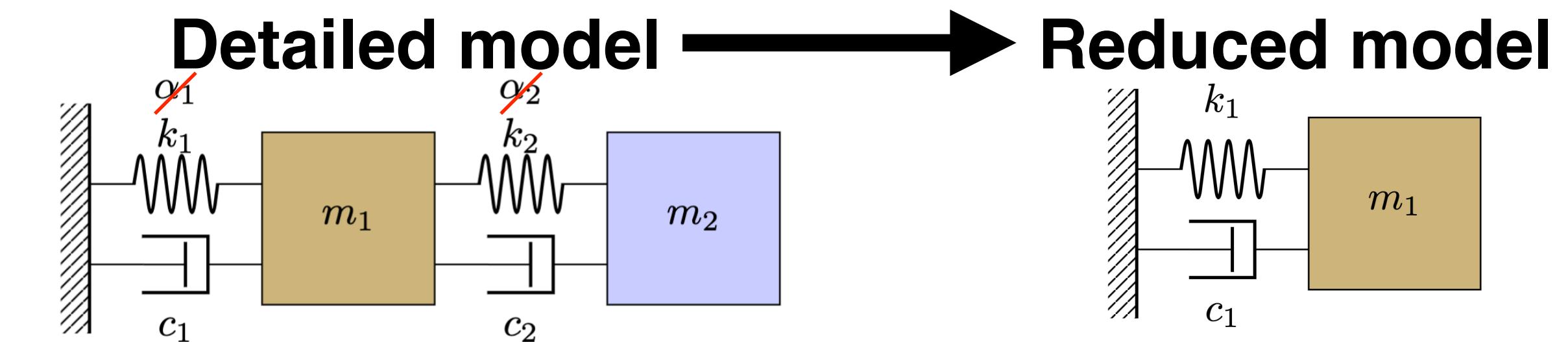
**Damping**  $\rightarrow f_{d,i}(t) = -c_i \dot{x}_i(t)$

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

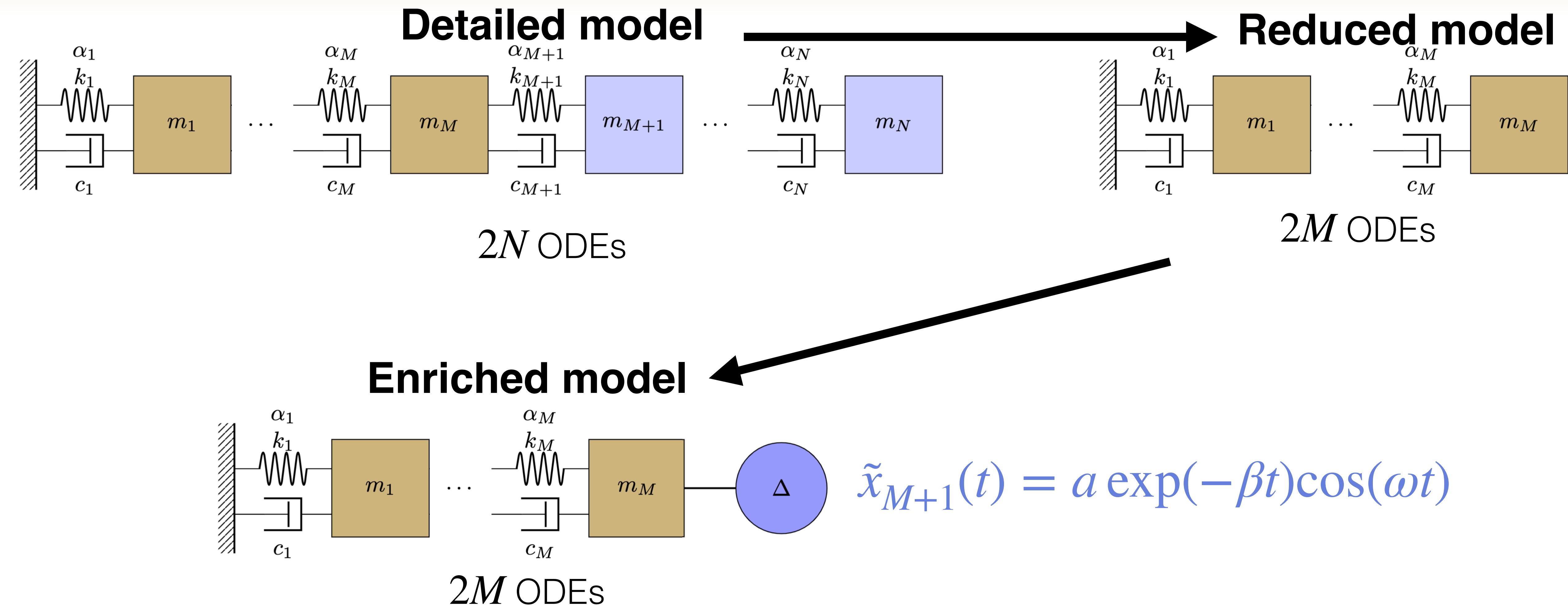
The reduced model is formed  
by subsampling the detailed model



# Omitting masses causes error in the reduced model

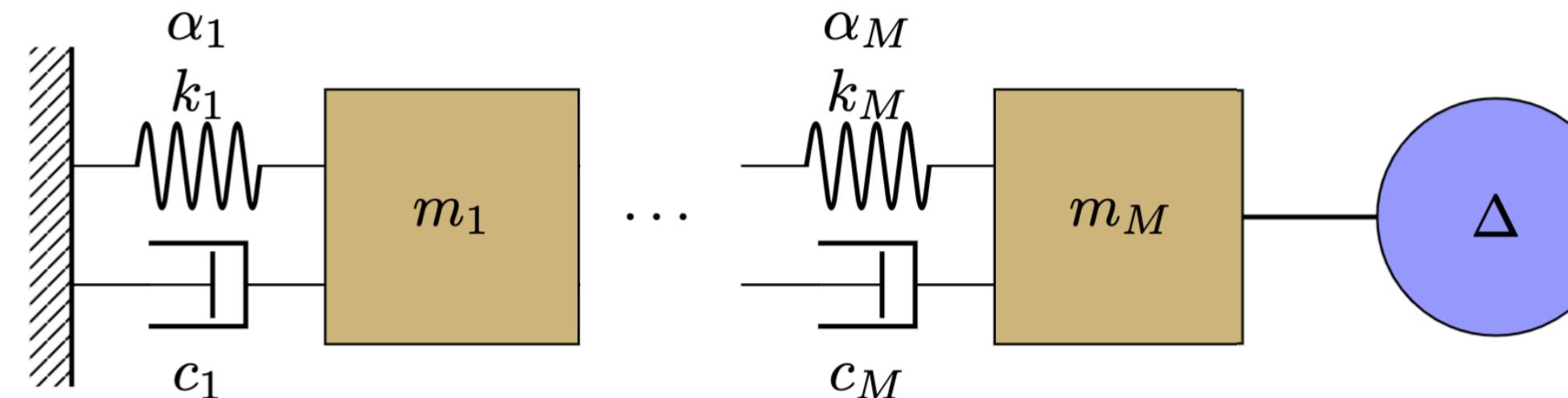


An enrichment operator is added to the reduced model to form the enriched models



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

## Enriched model

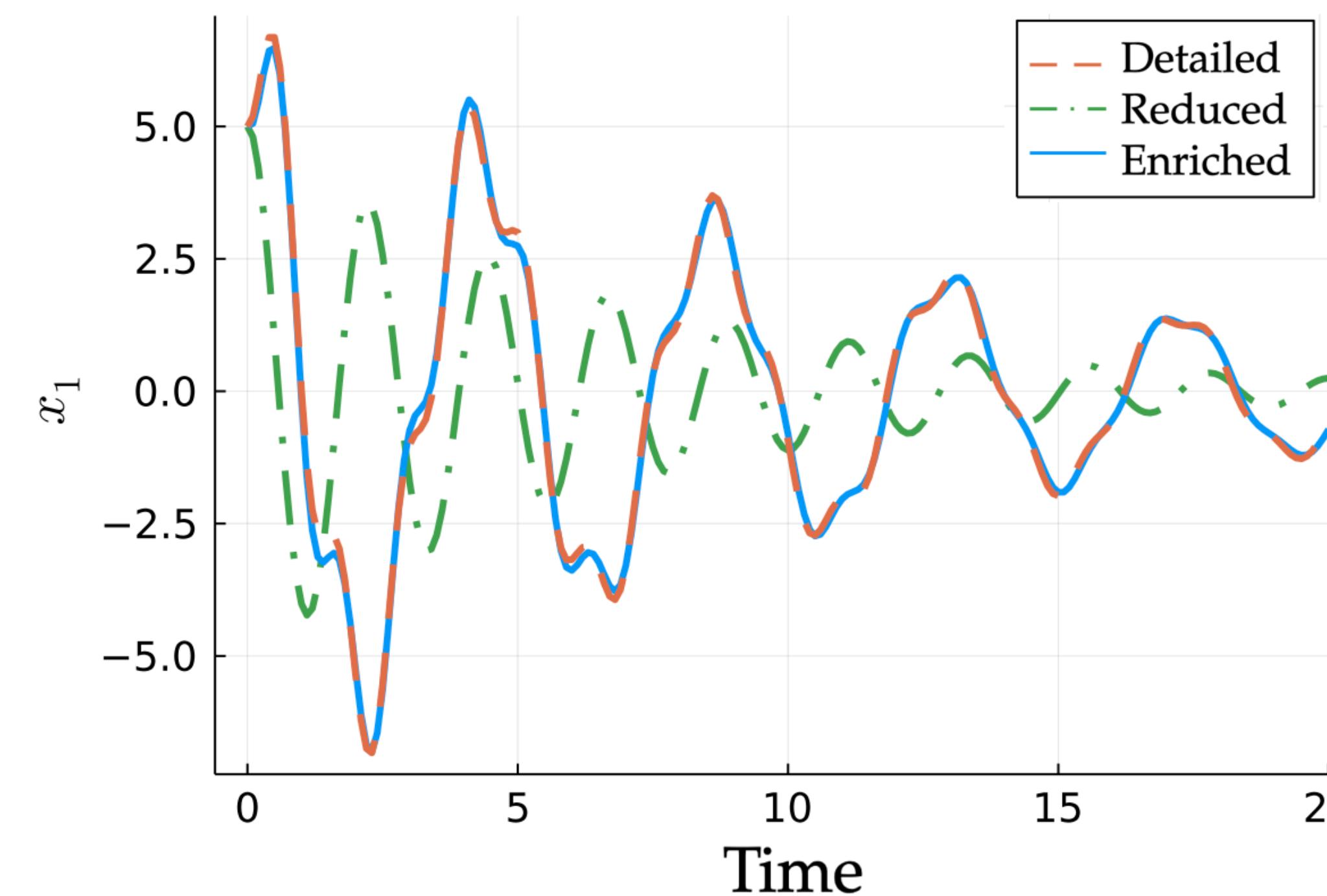
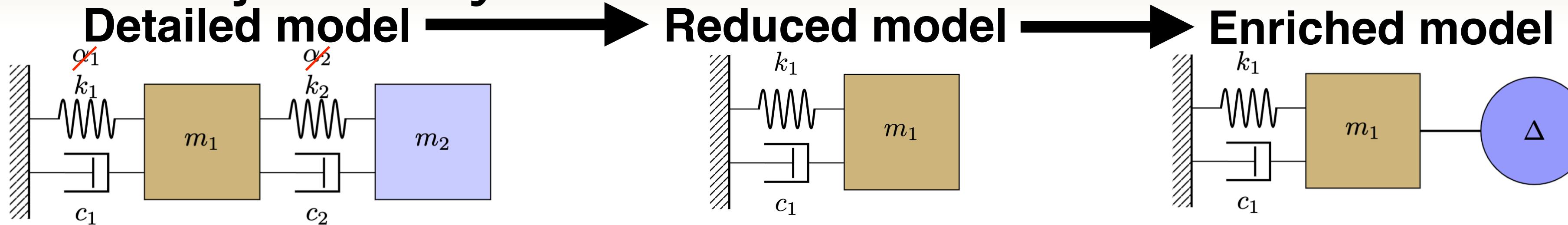


$$\tilde{x}_{M+1}(t) = a \exp(-\beta t) \cos(\omega t)$$

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

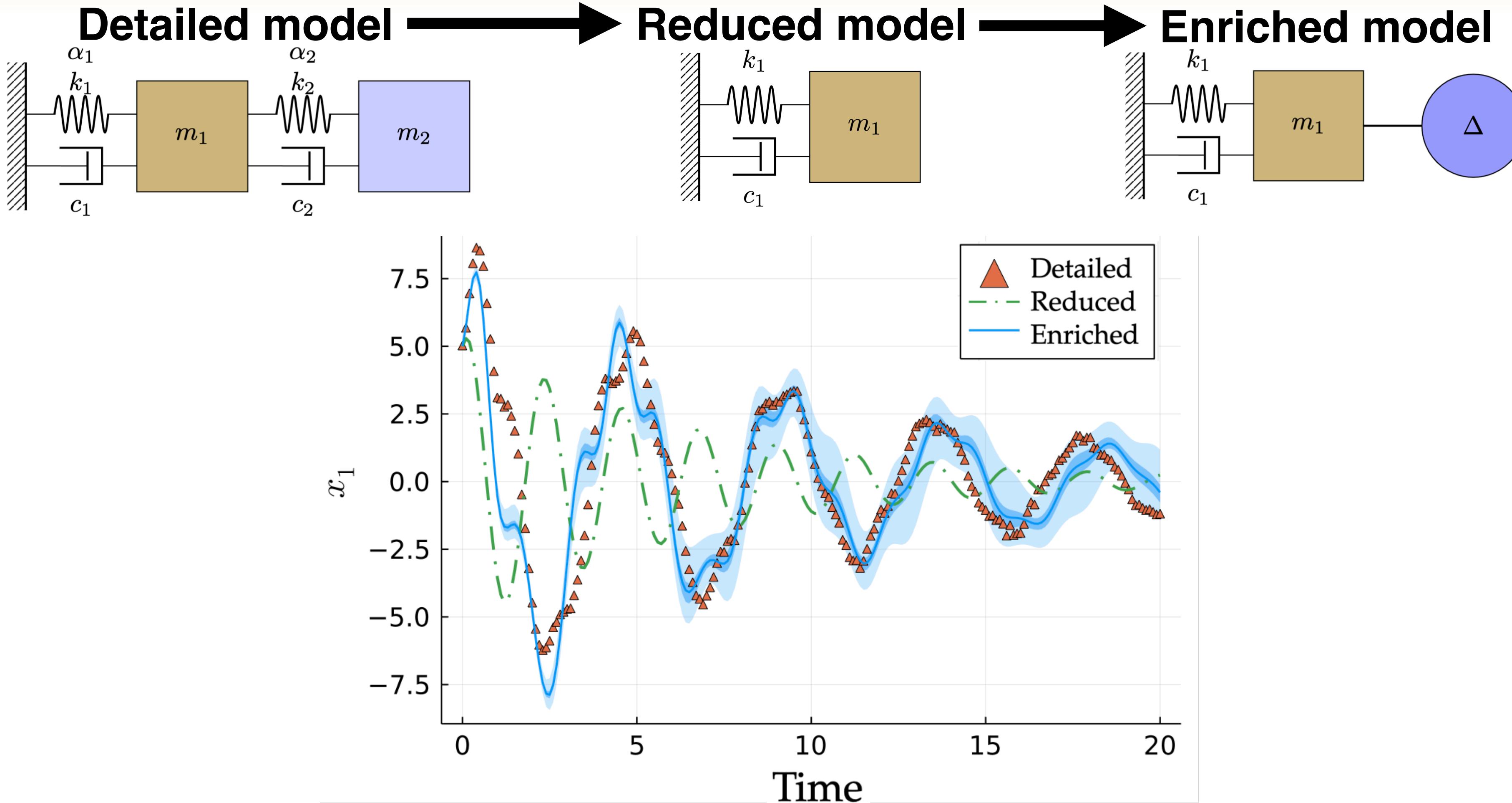
- Model parameters:  $\theta = (\delta, a, \beta, \omega)$
- $\theta_i \sim \mathcal{N}(\mu_i, \sigma_i^2)$ , where  $\mu_i \in \mathbb{R}$ ,  $\sigma_i \in \mathbb{R}_{\geq 0}$ , and  $i = \{1, 2, 3, 4\}$
- Hyperparameters:  $\phi = (\phi_1, \phi_2, \phi_3, \phi_4)$ , where  $\phi_i = (\mu_i, \sigma_i)$
- We use hierarchical Bayesian calibration to sample a posterior on  $\phi$

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

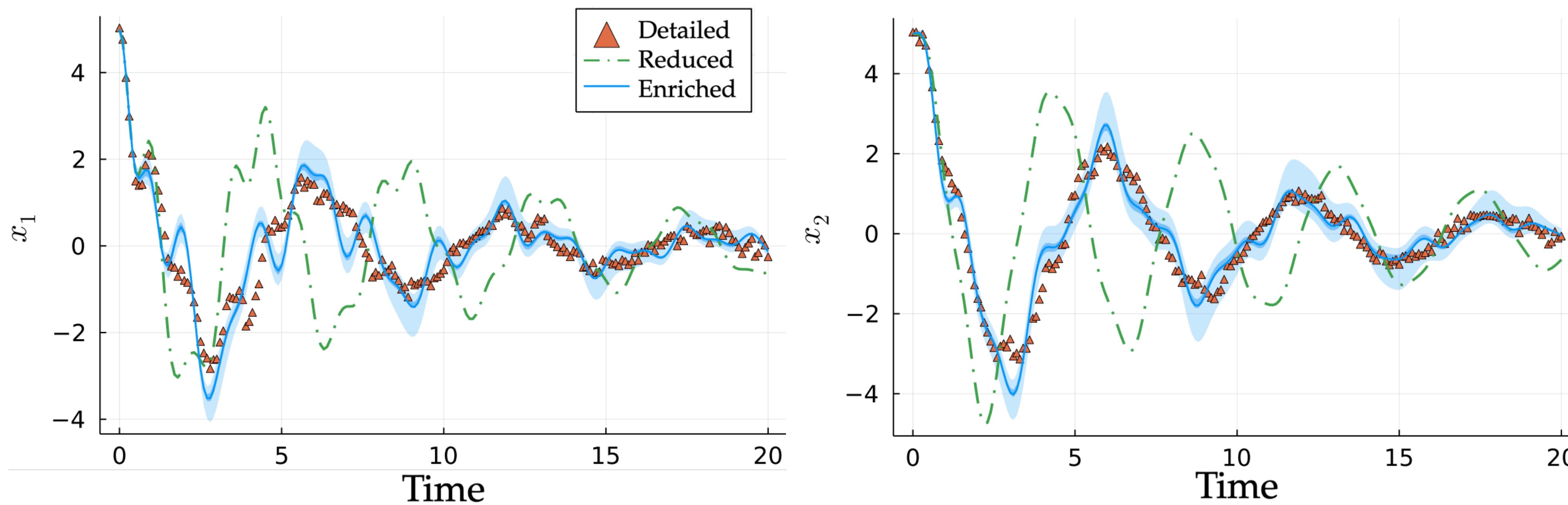
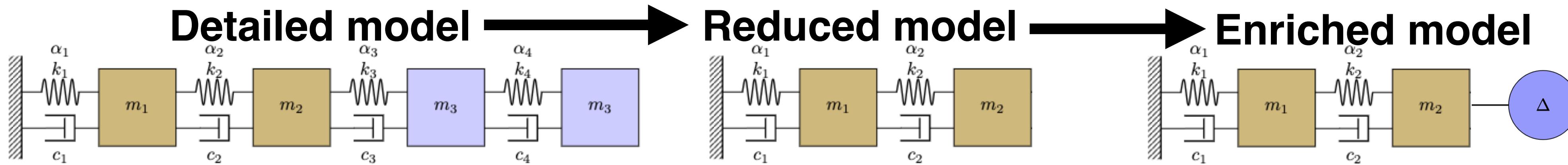


Enriched model is calibrated with mass one's displacement data.

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

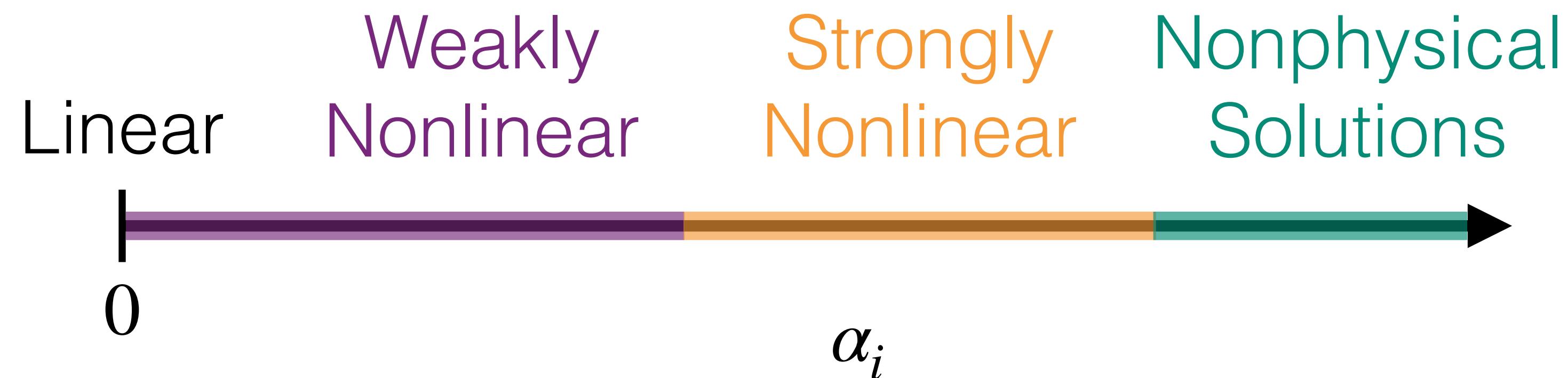


# Discrepancies emerge between the enriched model and the **nonlinear** detailed model

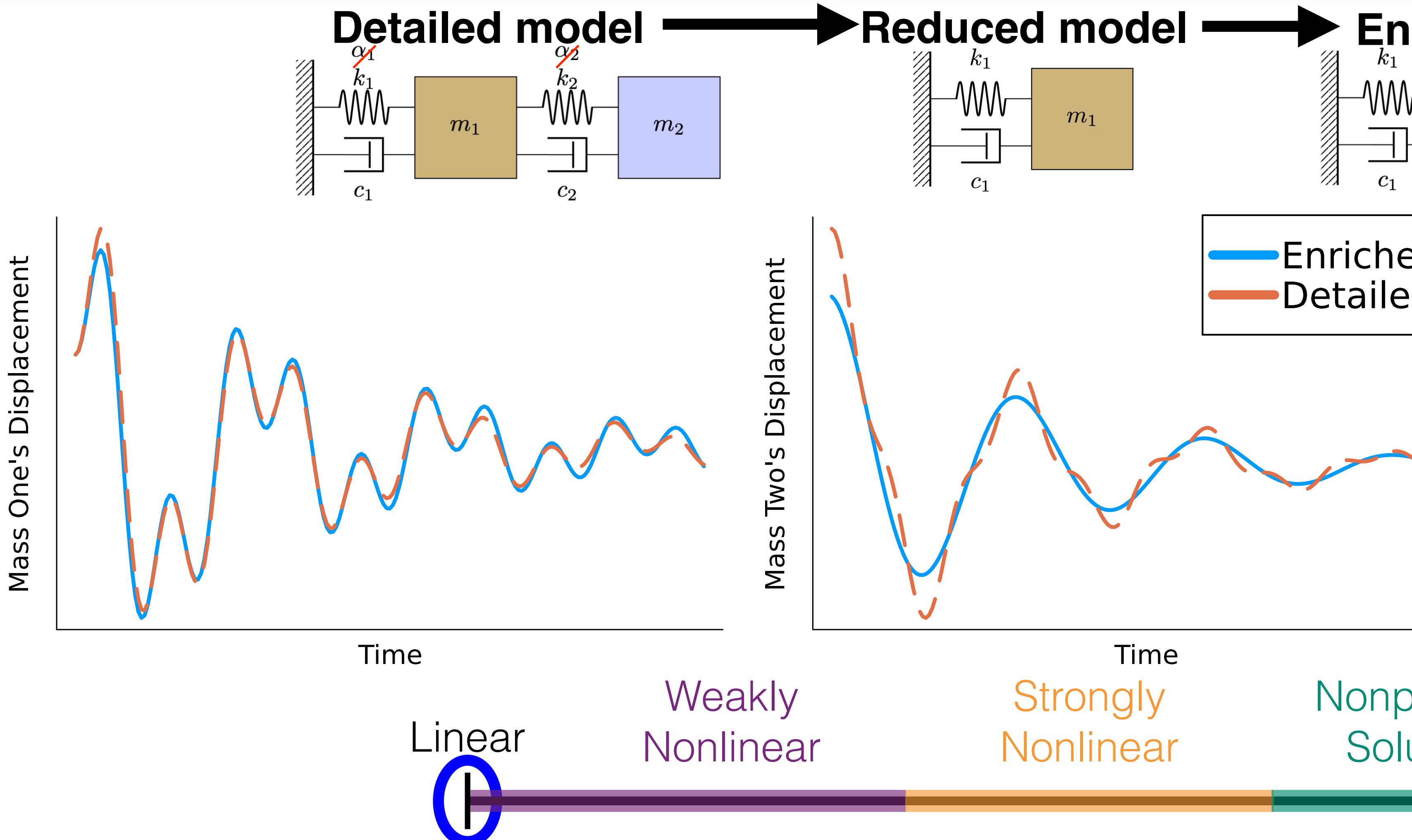


How does the degree of nonlinearity impact model discrepancy?

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

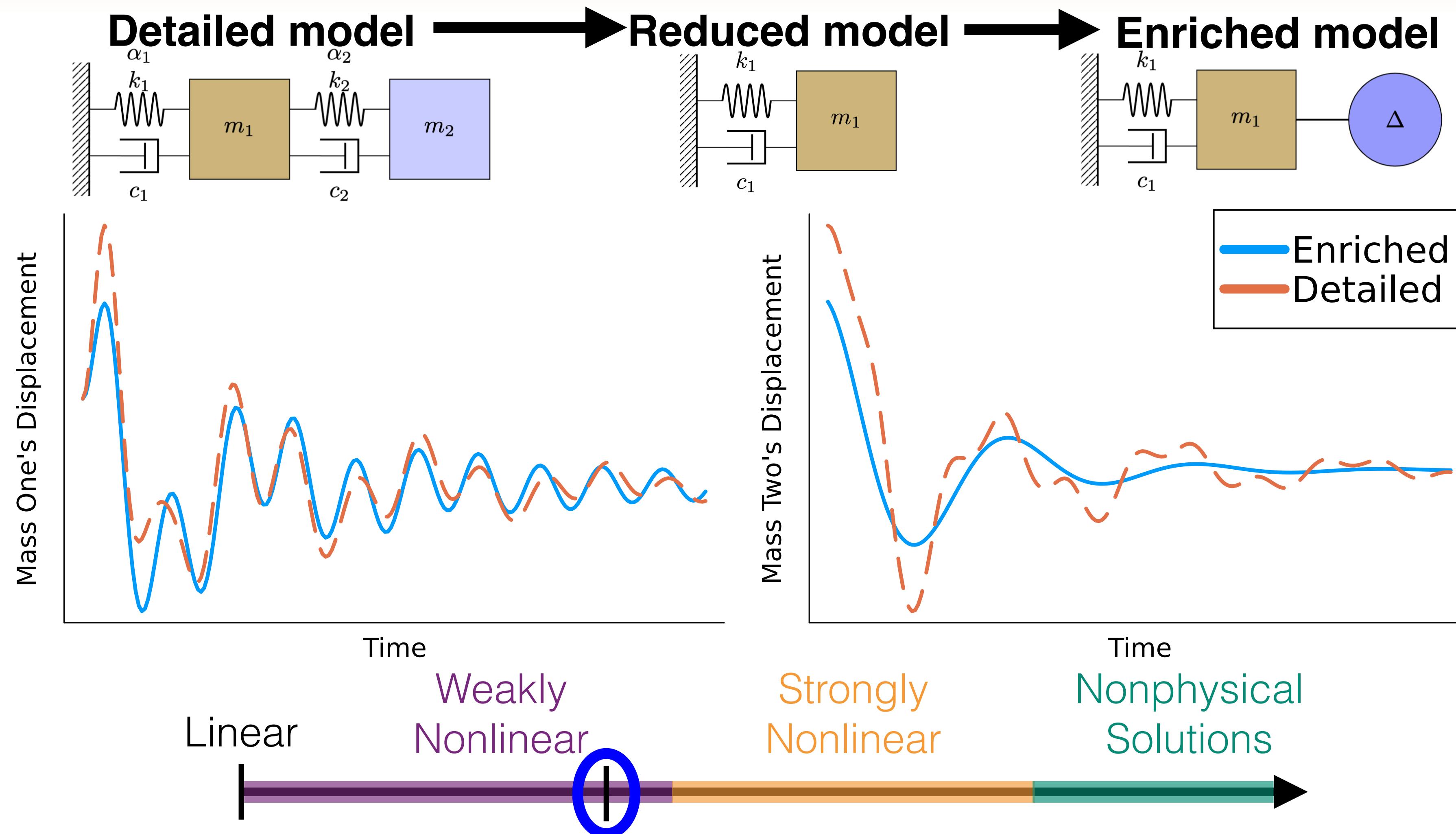


# The enriched model closely matches the **linear** detailed trajectories

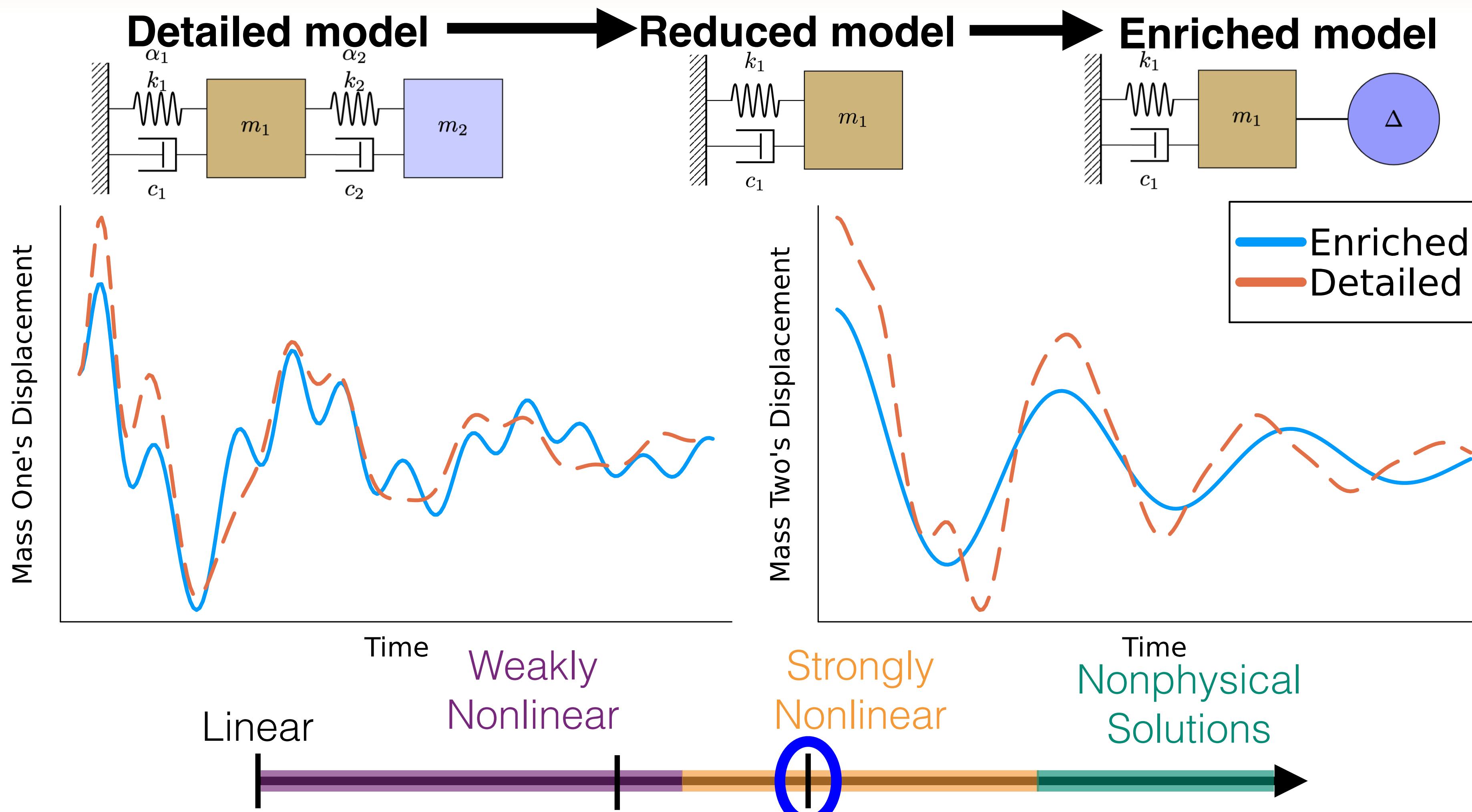


Enriched model predictions are using MAP estimates of model parameters.

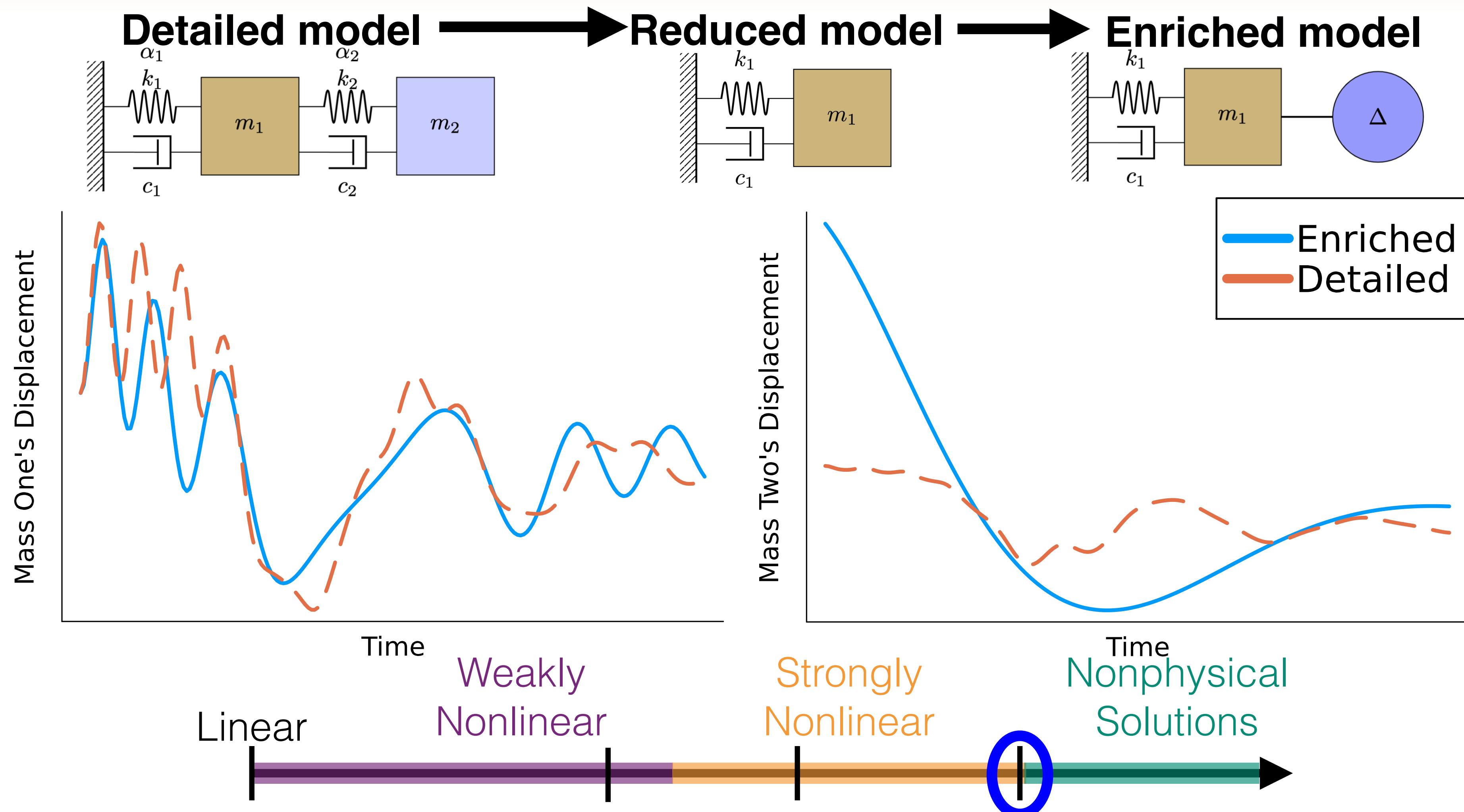
The enriched model overestimates the dampening of mass two for the **weakly nonlinear** model



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

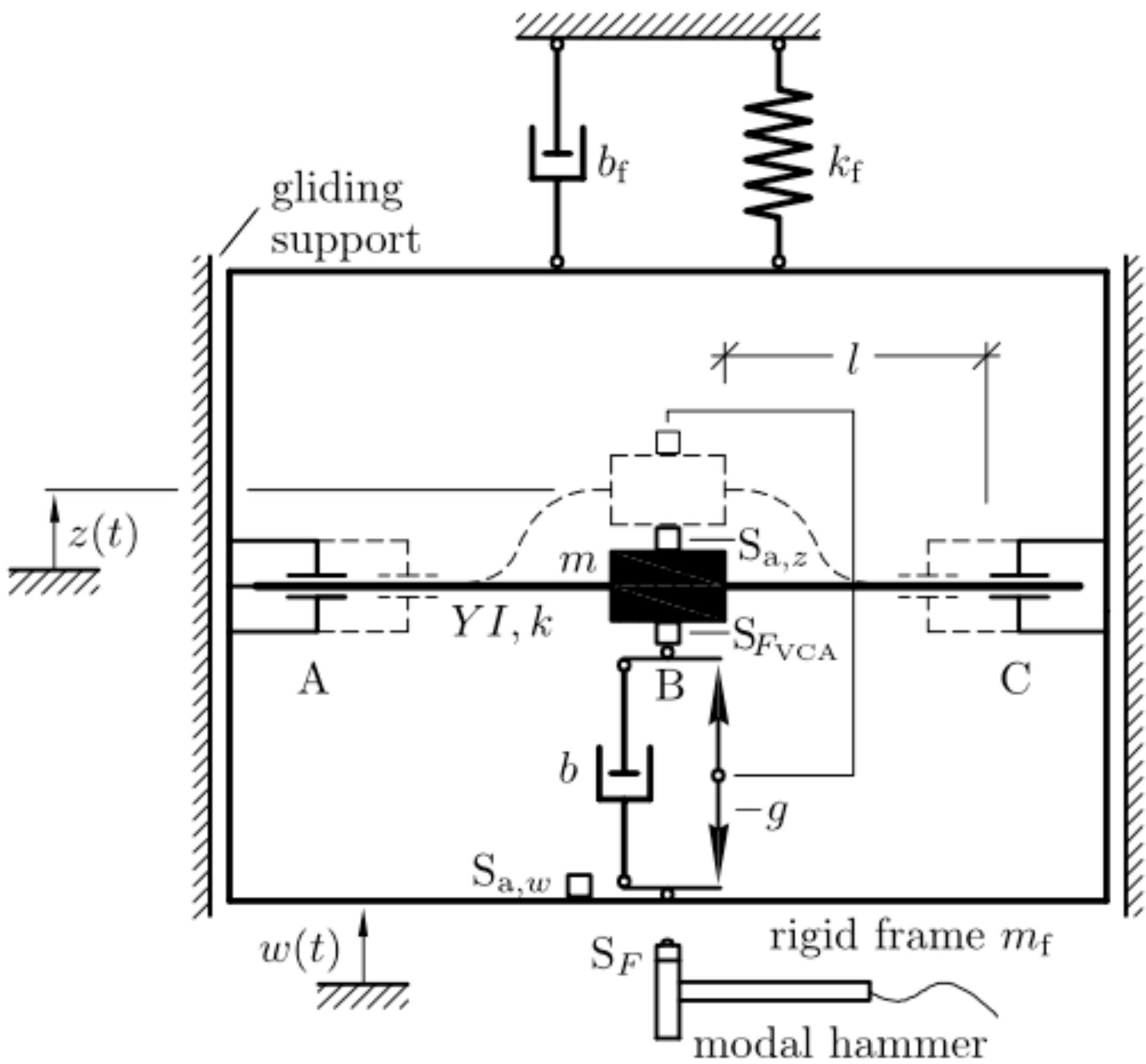


The enriched model overestimates the initial amplitude of mass two for the **strongest nonlinear** model

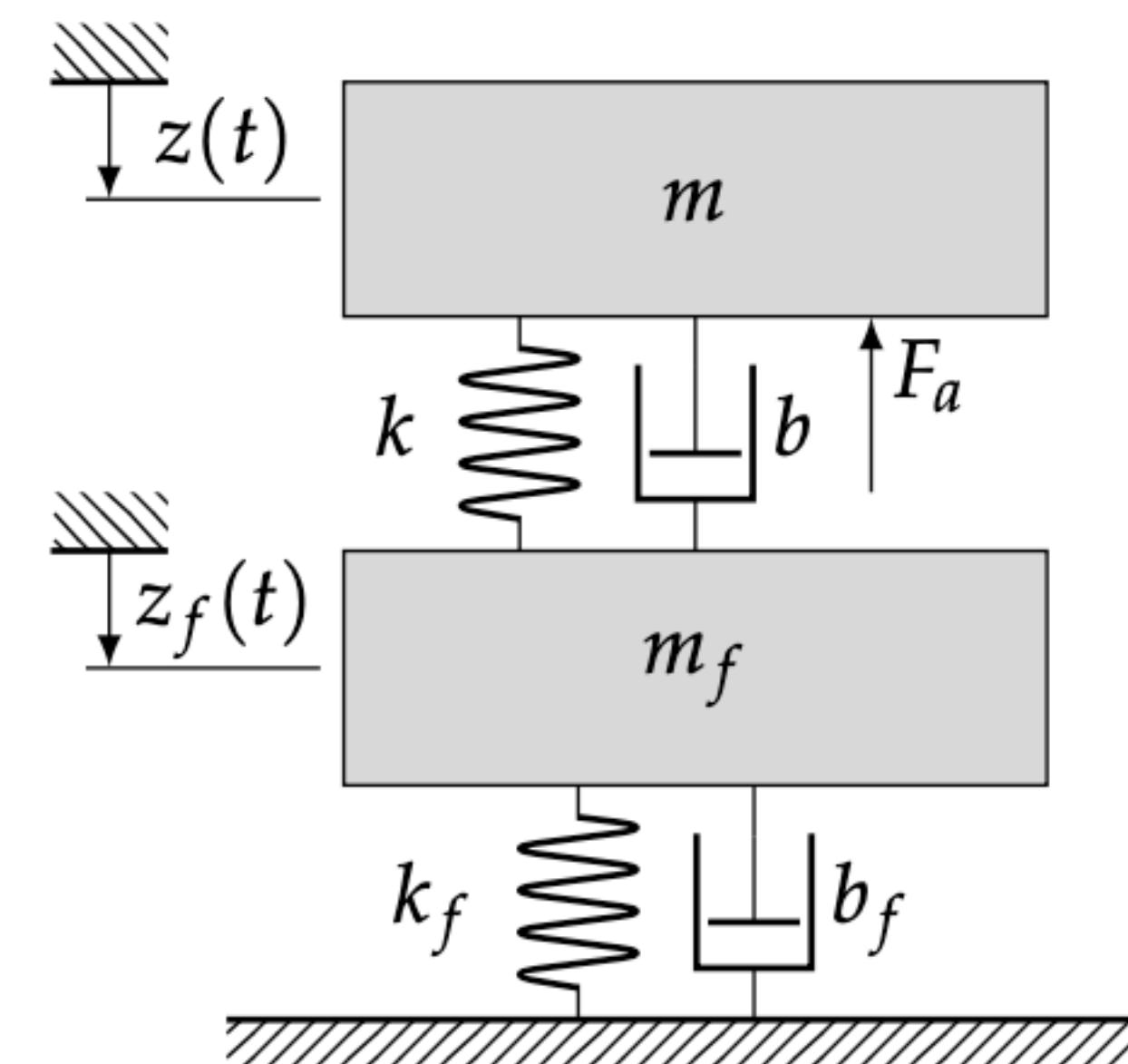


# Case 2

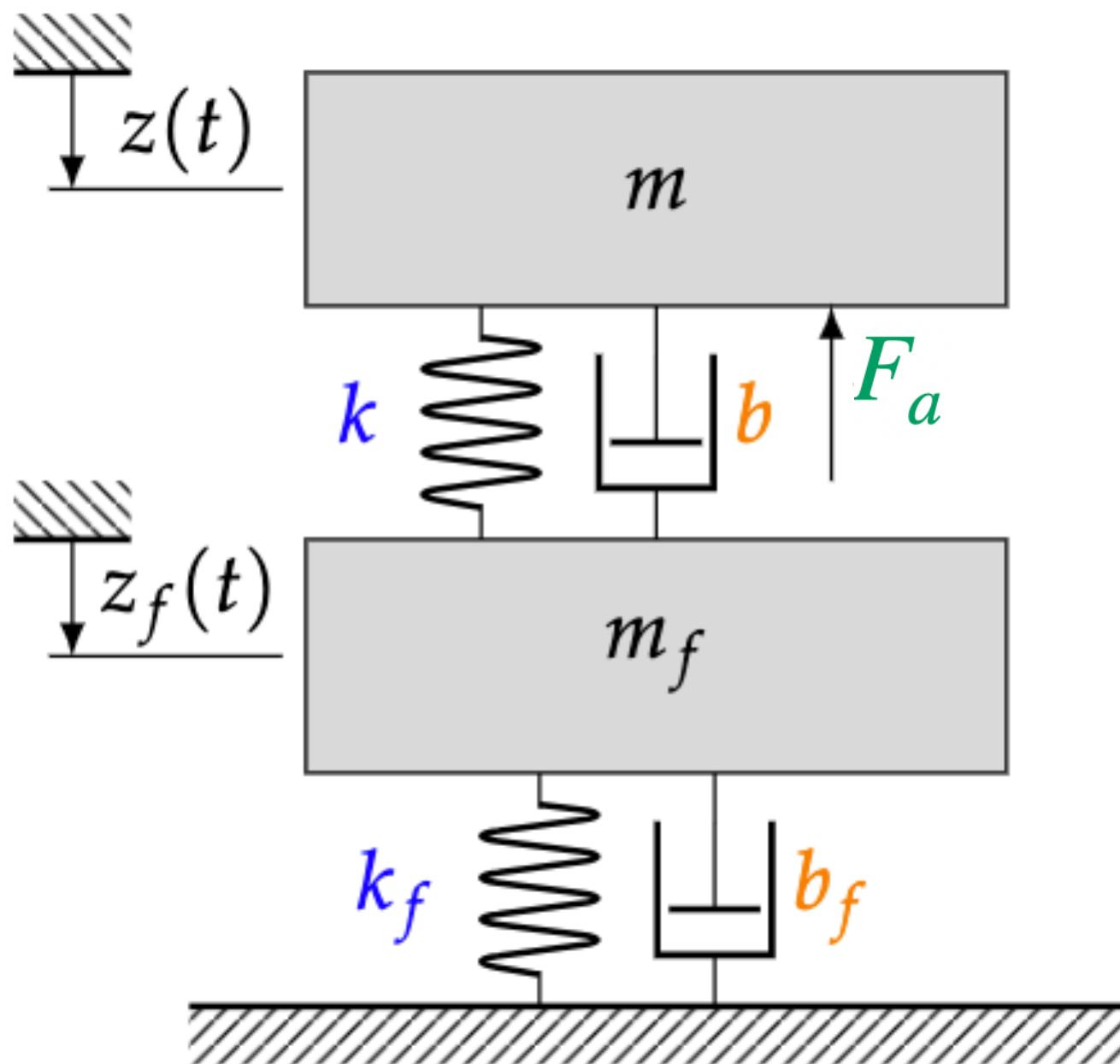
## Experiment



## Low-fidelity model



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



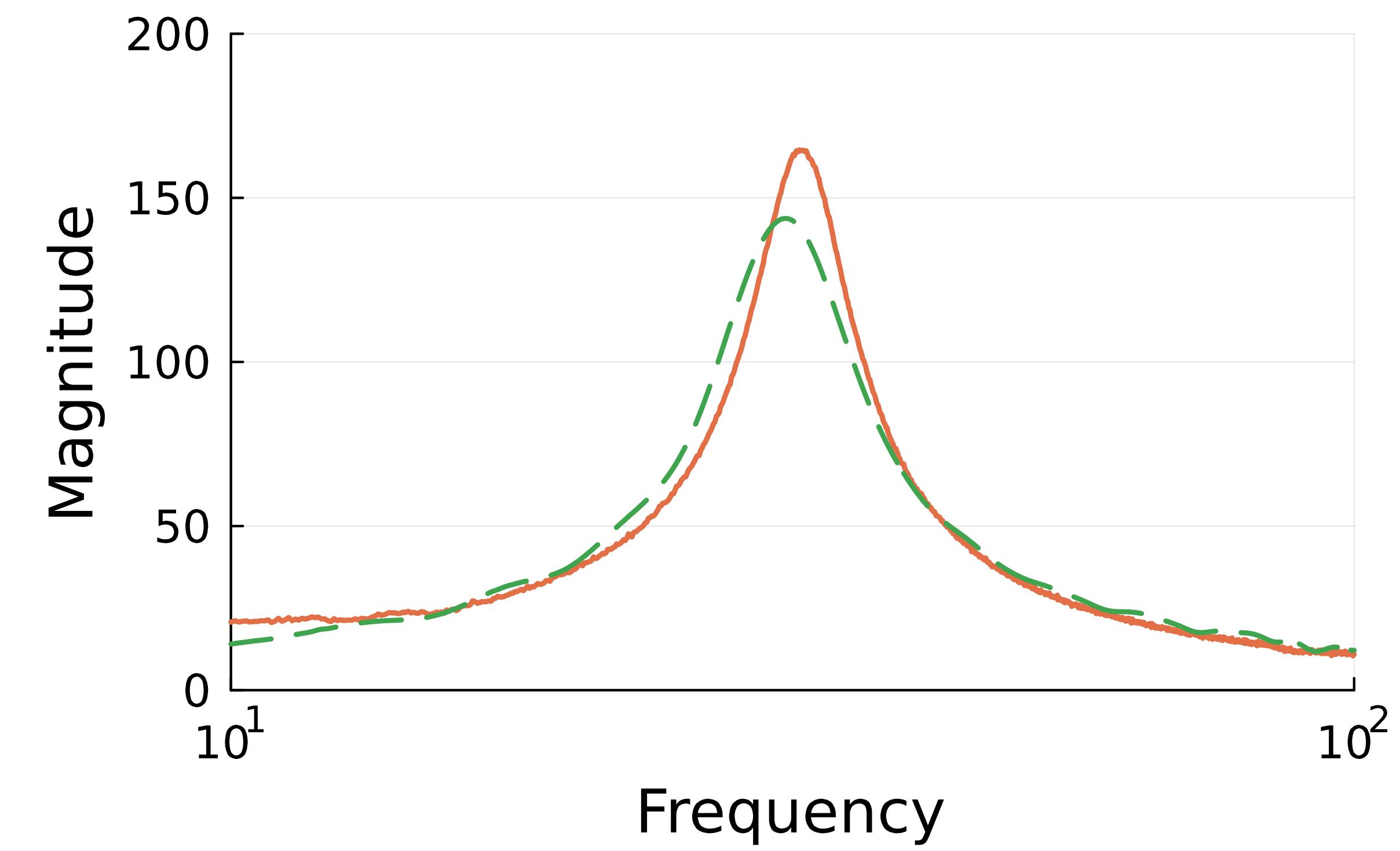
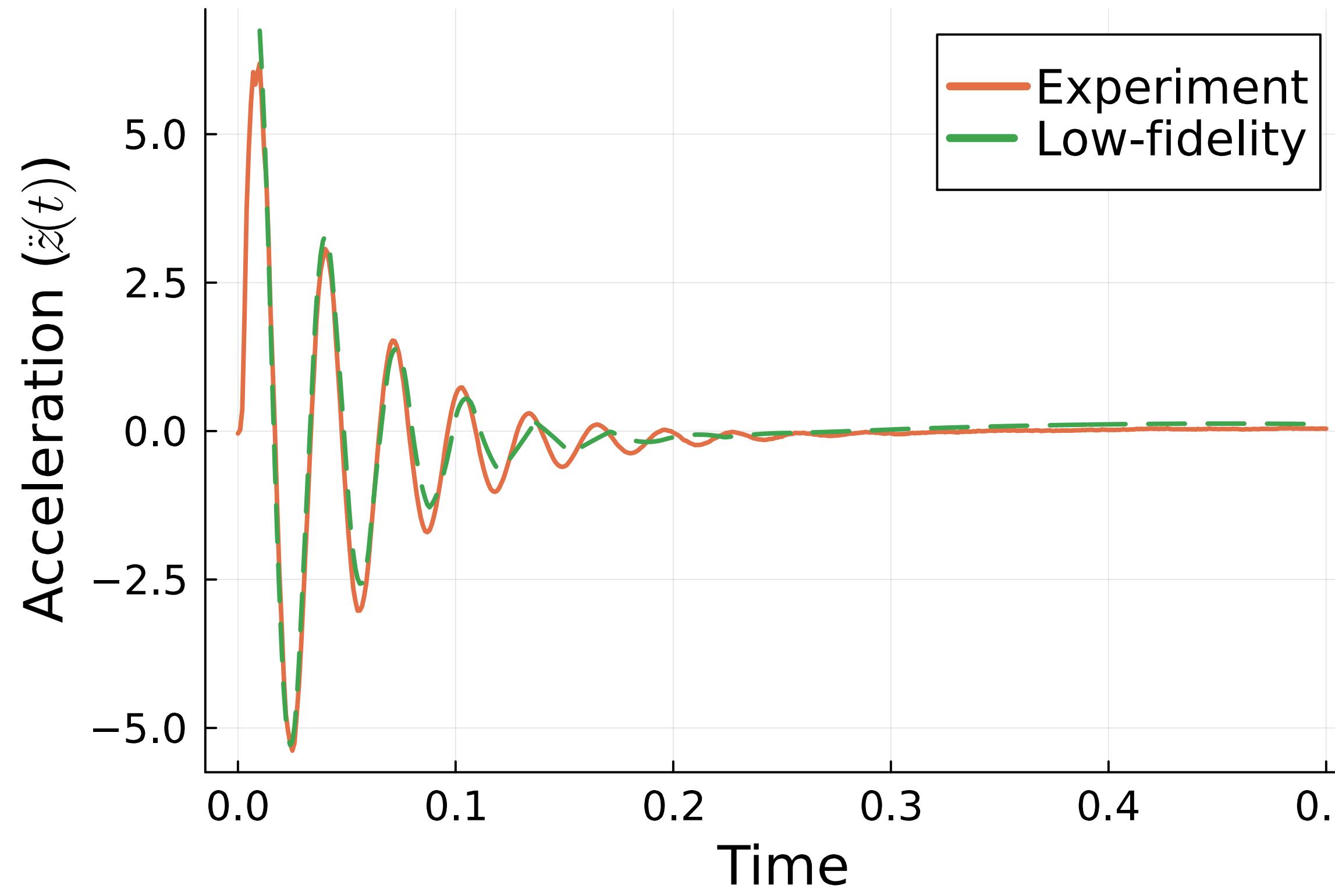
$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - F_a \left( \dot{z}(t) \right) - k \left( z(t) - z_f(t) \right) \right\}$$

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

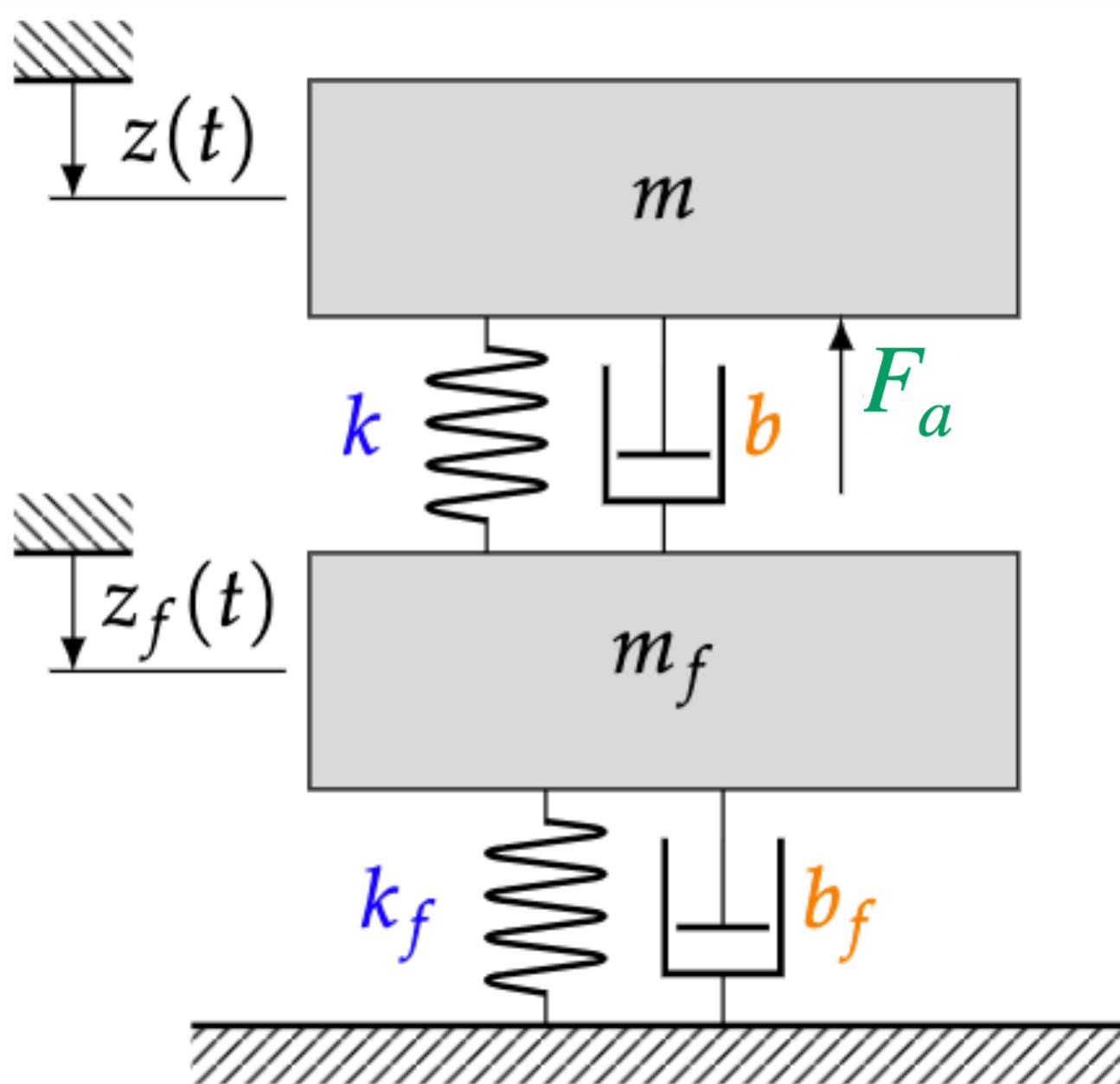
Spring force  
Passive damping force  
Active damping force

$g_0 \dot{z}(t)$

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



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

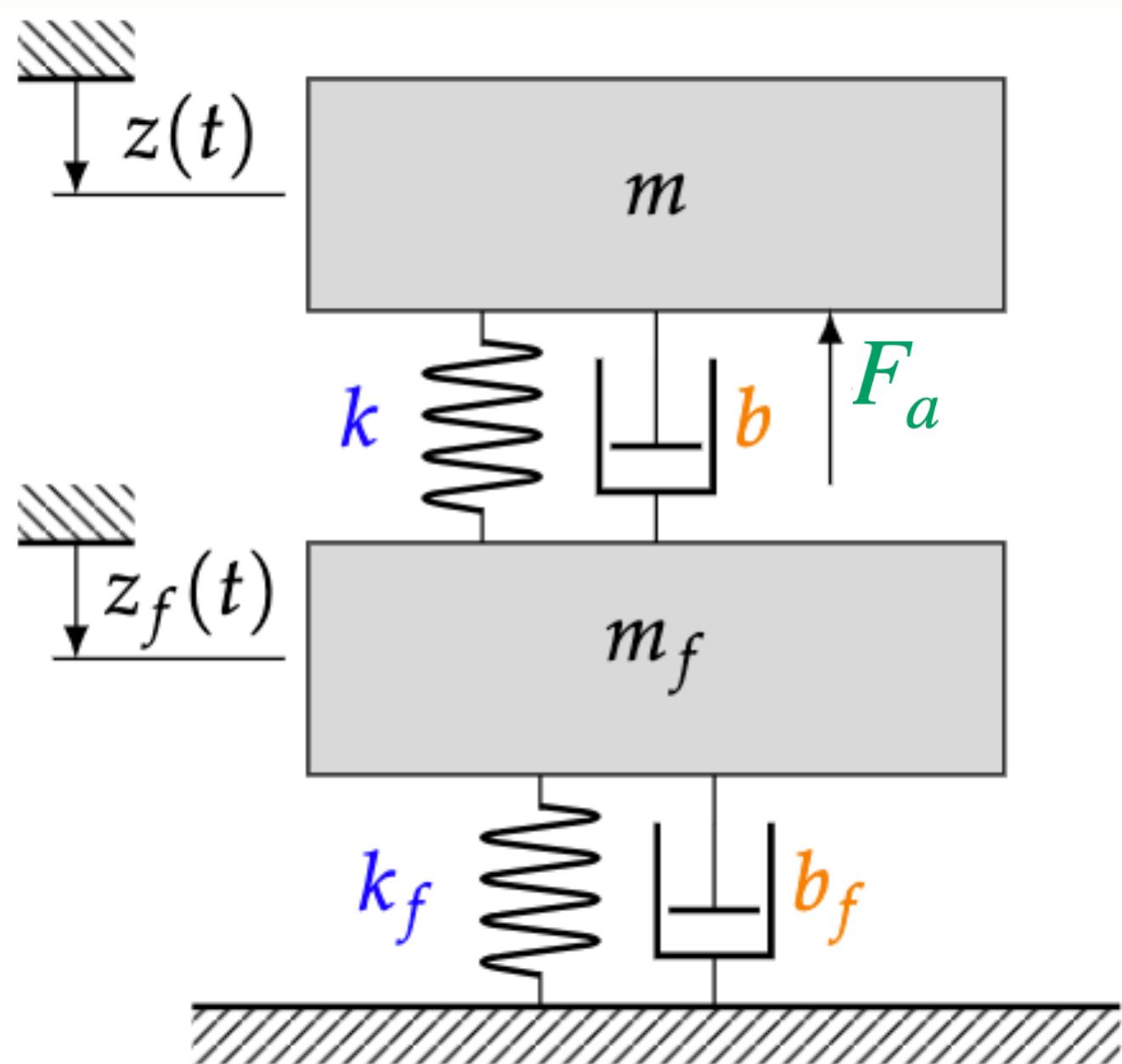


$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - F_a \left( z(t) \right) - k \left( z(t) - z_f(t) \right) \right\}$$

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

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

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



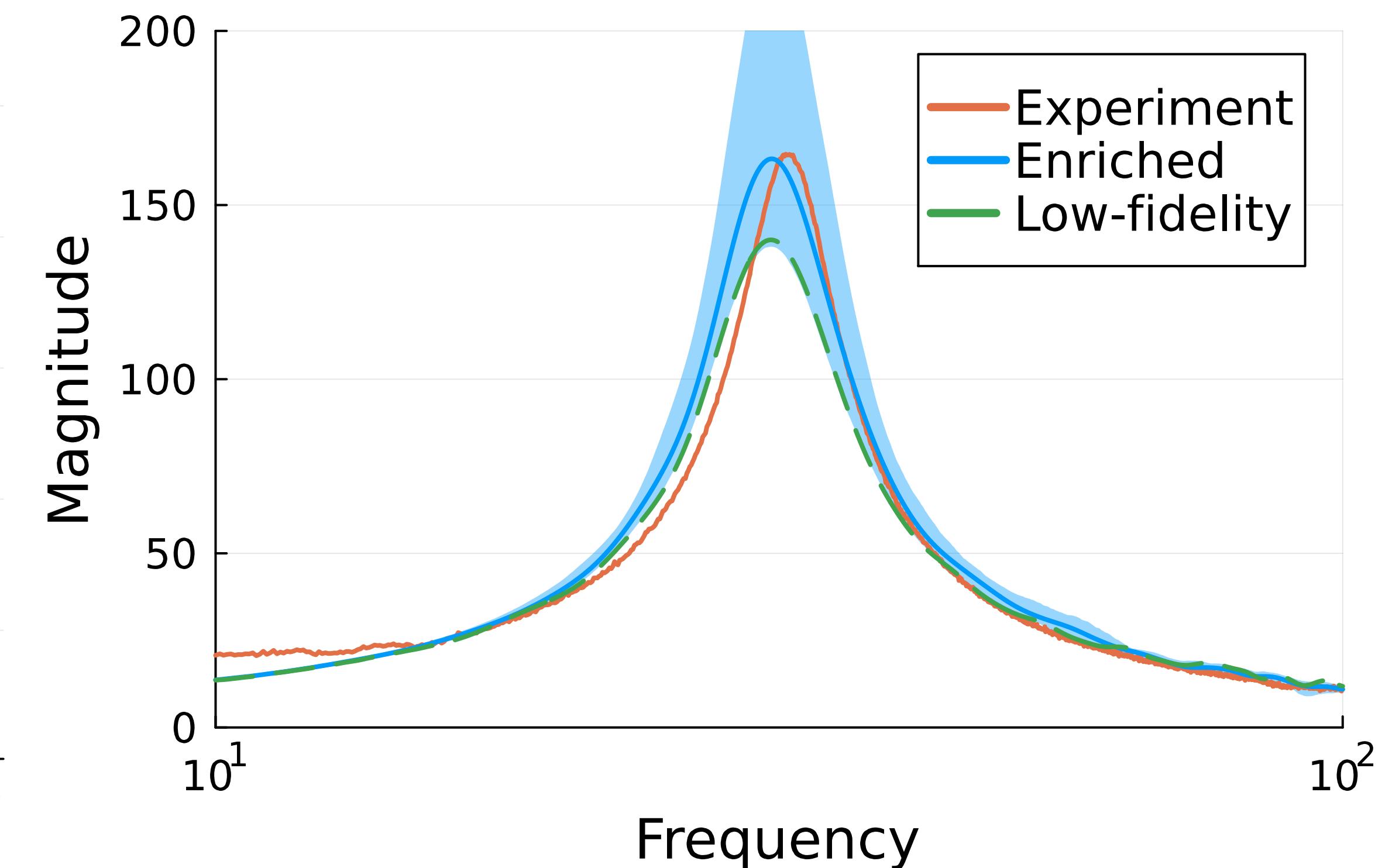
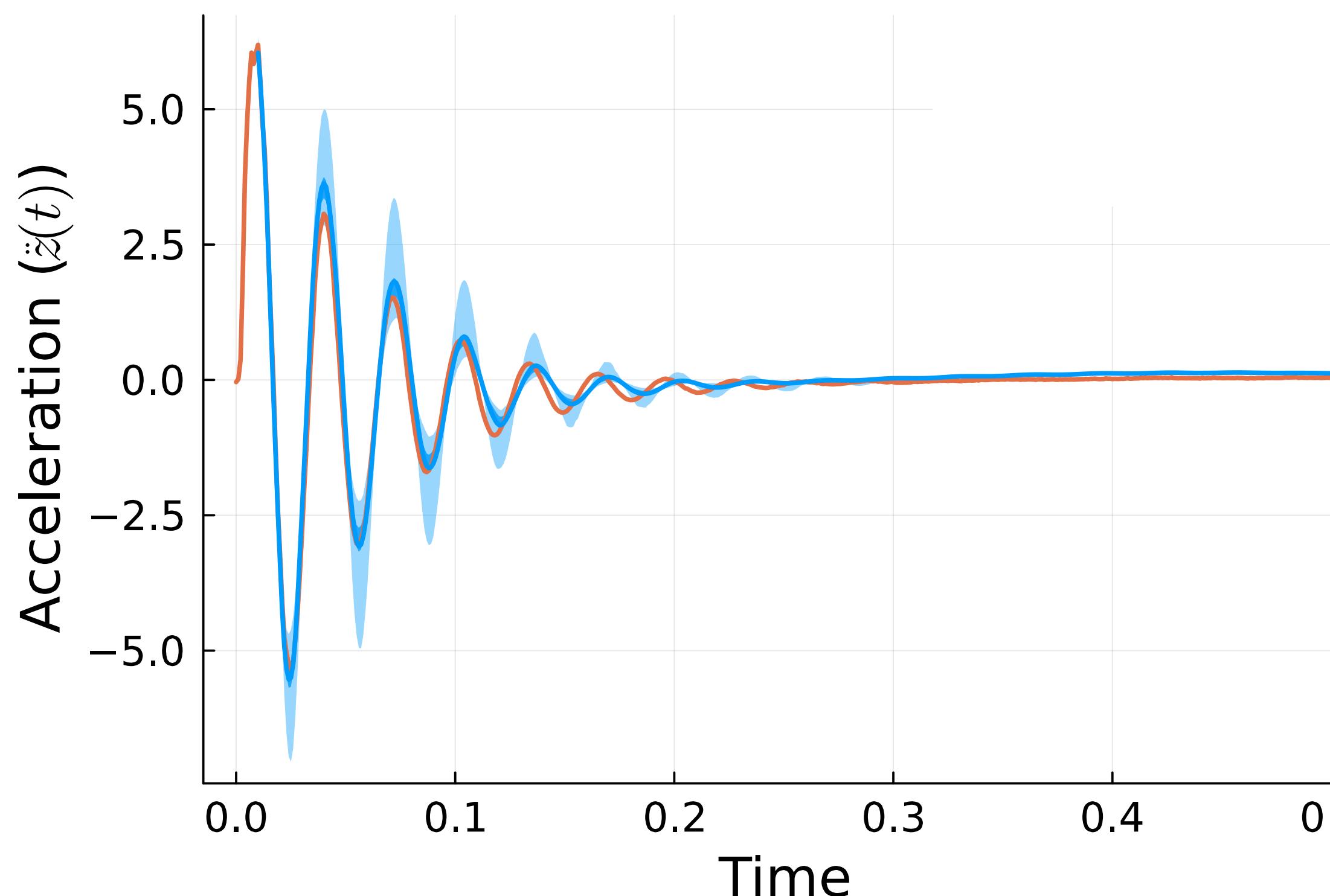
$$\ddot{z}(t) = \frac{1}{m} \left\{ -b \left( \dot{z}(t) - \dot{z}_f(t) \right) - F_a \left( \dot{z}(t) \right) - k \left( z(t) - z_f(t) \right) \right\}$$

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

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

$$g_i \sim \mathcal{N}(\mu_i, \sigma_i^2)$$
, where  $\mu_i \in \mathbb{R}$ ,  $\sigma_i \in \mathbb{R}_{\geq 0}$ , and  $i = \{1,2\}$
- Hyperparameters:  $\phi = (\phi_1, \phi_2)$ , where  $\phi_i = (\mu_i, \sigma_i)$
- We use hierarchical Bayesian calibration to sample a posterior on  $\phi$

# The enriched model covers most experimental observations



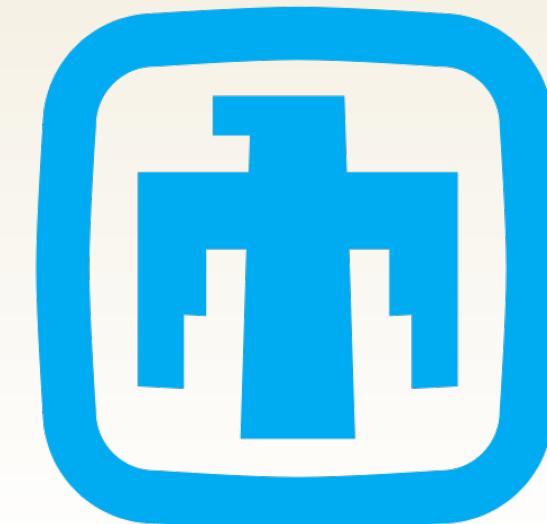
[5] R. BANDY, T. PORTONE, AND R. MORRISON, *Stochastic model correction for the adaptive vibration isolation round-robin challenge*, *Model Validation and Uncertainty Quantification*, Volume 3, Conference Proceedings of the Society for Experimental Mechanics Series, (to be released).

# 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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