



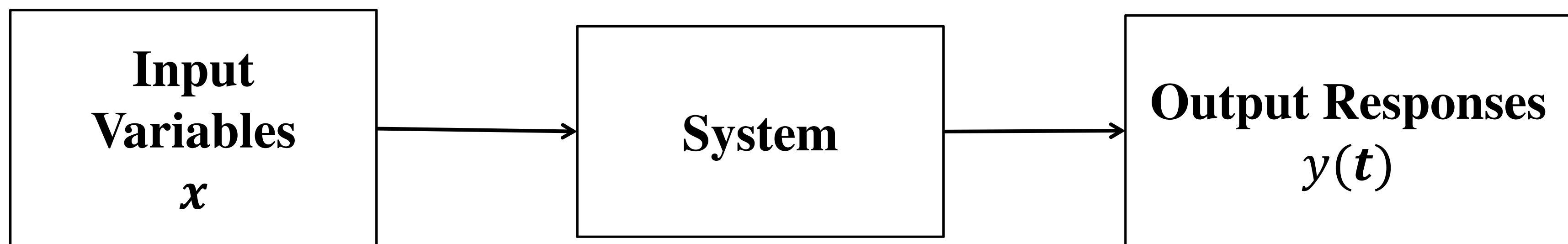
# Tractable Modeling of Systems with Functional Outputs with Nonseparable Covariance Structures



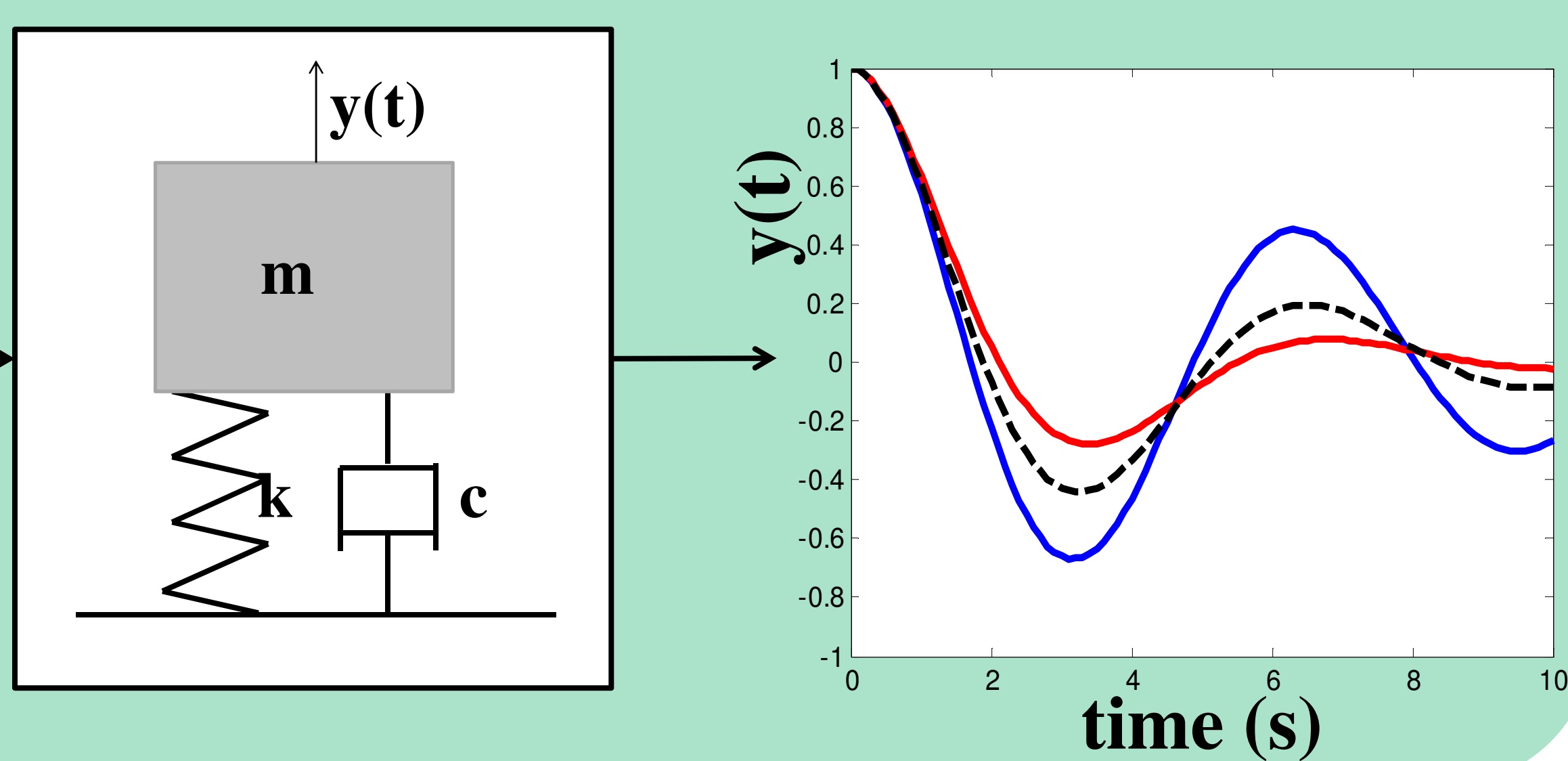
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## Motivation



$x$   
 $m$ := Mass of the Weight  
 $k$ := Spring Constant  
 $c$ := Damping Coefficient



## Previous Approach

### Separable Covariance

$$C((x, t), (x', t')) = C_x(x, x') C_t(t, t')$$

Used in (recent examples): Williams et al. (2006), Genton (2007), Rougier (2008), Bayarri et al. (2009), and Conti and O'Hagan (2010)

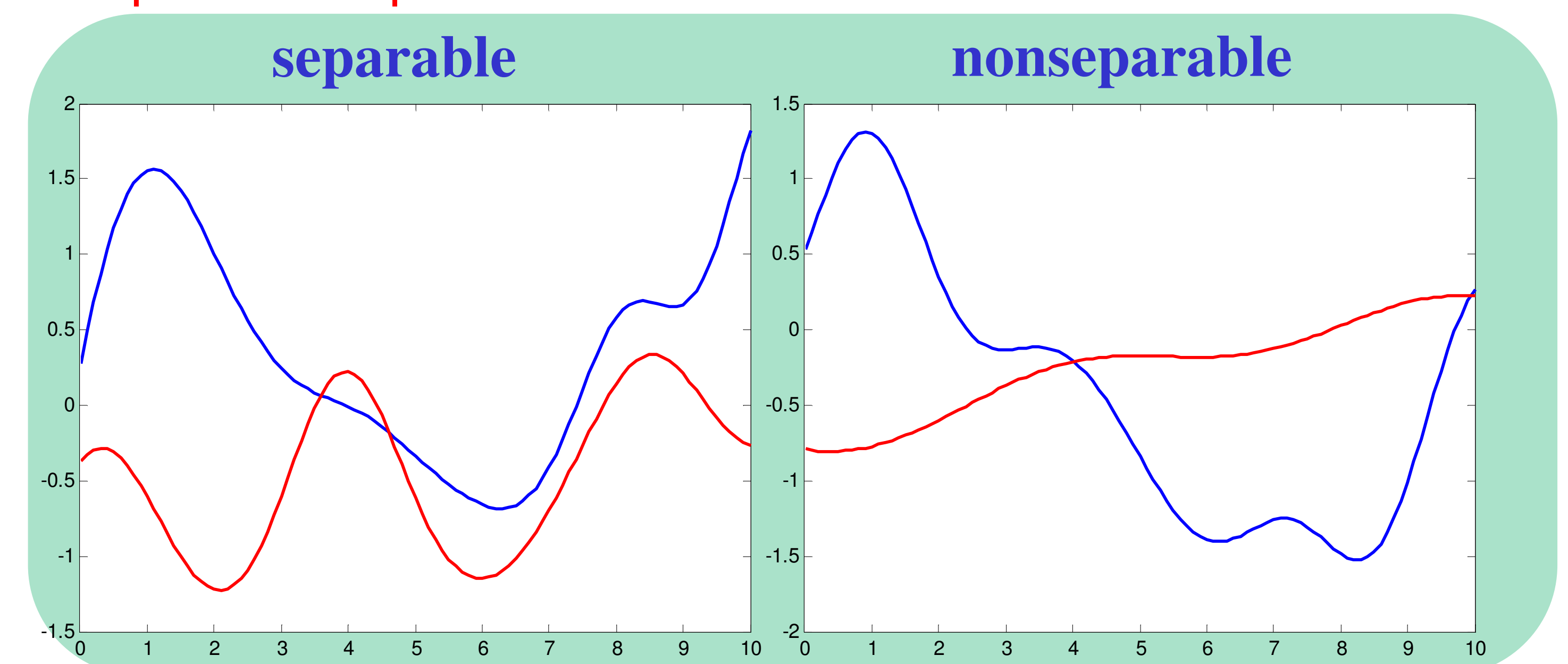
Under this assumption:

$$\Sigma = \Sigma_x \otimes \Sigma_t \rightarrow \Sigma^{-1} = \Sigma_x^{-1} \otimes \Sigma_t^{-1}$$

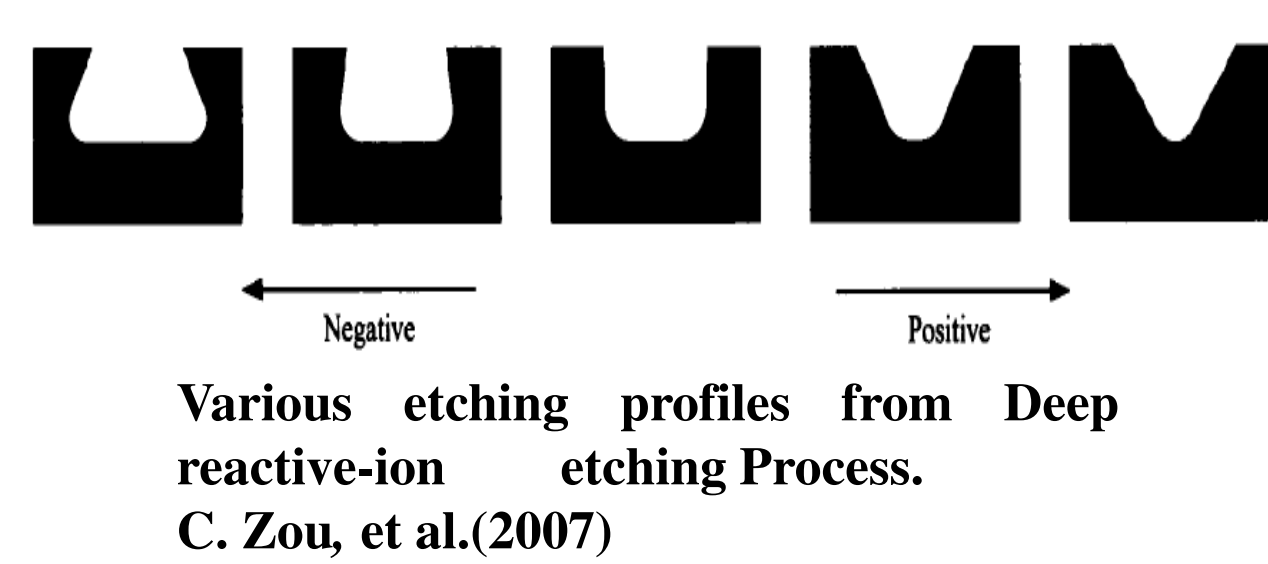
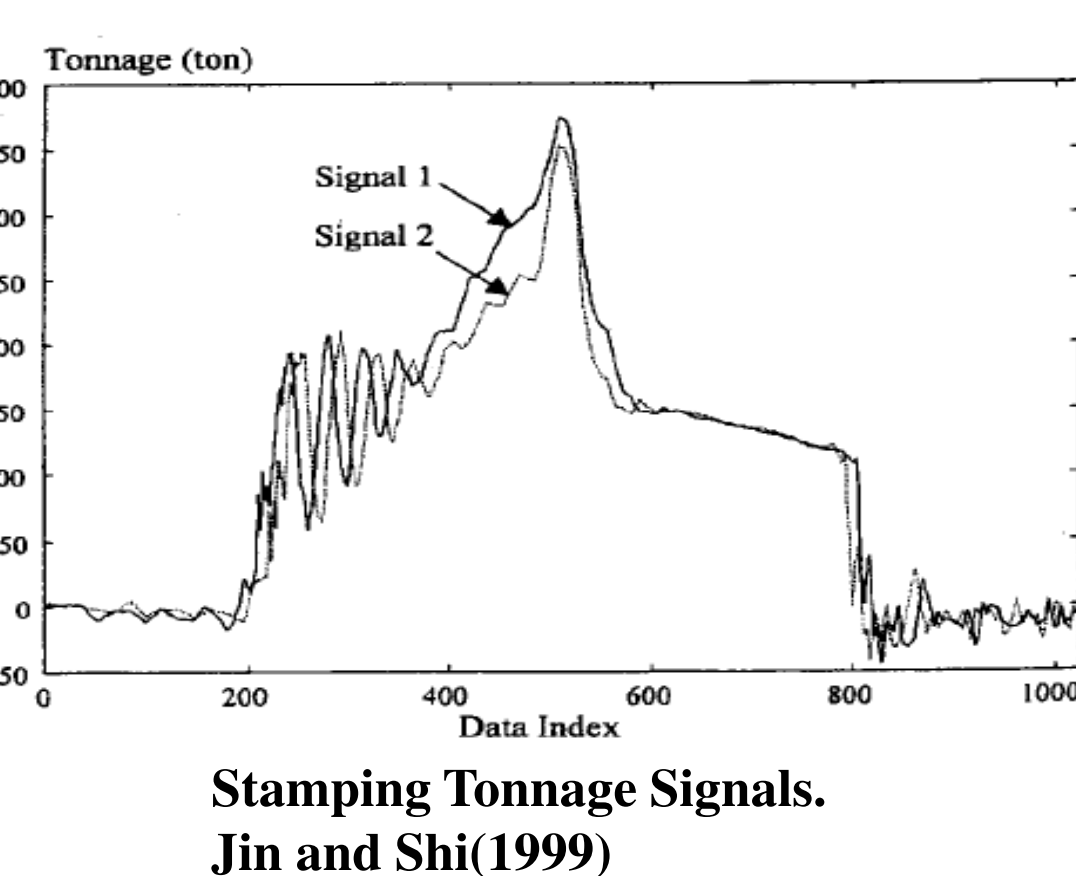
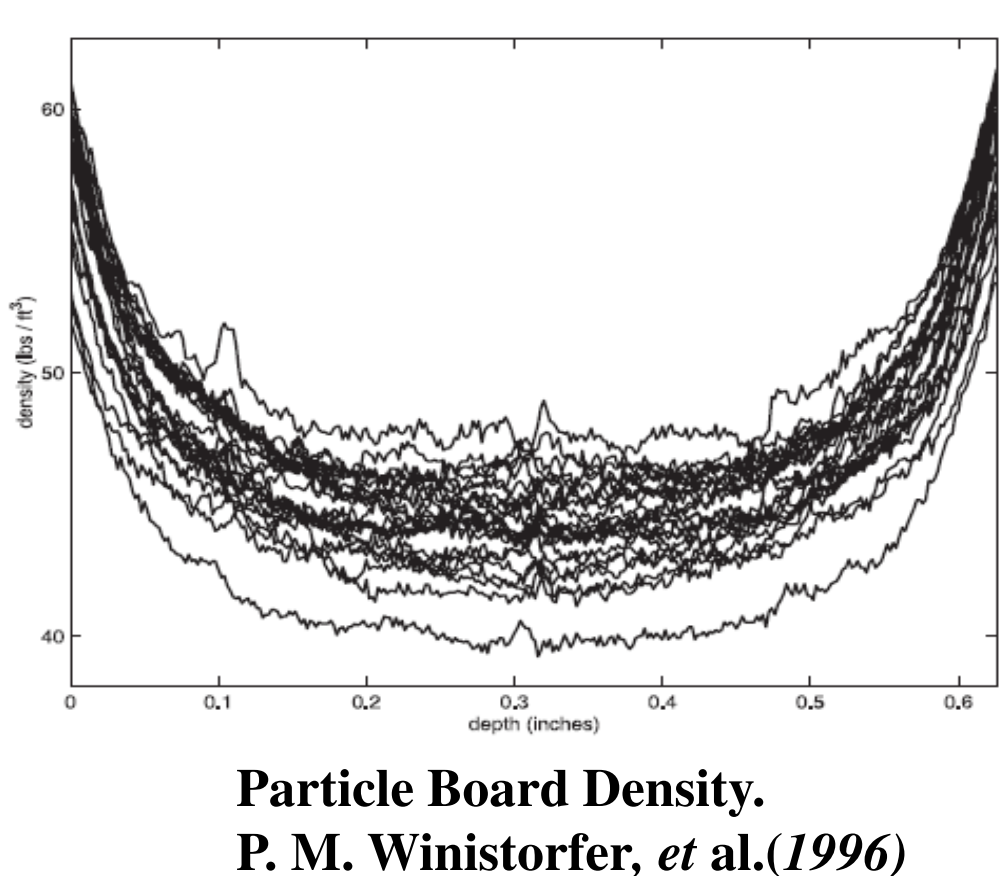
Likelihood evaluation and prediction are much faster.

$\Sigma_t$  is  $n \times n$  and  $\Sigma_x$  is  $m \times m$   
operations:  $\mathcal{O}(\max(n^3, m^3))$   
bytes of storage:  $\mathcal{O}(\max(n^2, m^2))$

However, Since  $C_t(t, t')$  is independent of  $x$ , the covariance of the output is independent of  $x$ .



## Examples of Functional Data Responses



## Objective

To predict a new functional response using previously observed functional responses.

## Accomplishments

We propose a nonstationary and nonseparable covariance function to model a wide variety of functional responses.

$$C((x, t), (x', t')) = C_x(x, x') C_{t|x}(t, t'|x, x')$$

conditionally warped

$$C_{t|x}(t, t'|x, x') = \sigma_t^\top(t|x) \Sigma_t^{-\frac{1}{2}}(x) \Sigma_t^{-\frac{1}{2}}(x') \sigma_t(t'|x')$$

where,  $\sigma_t^\top(t|x) = [C_t(t, t_1|x), C_t(t, t_2|x), \dots, C_t(t, t_n|x)]$   
and  $\{\Sigma_t(x)\}_{ij} = C_t(t_i, t_j|x)$

Using this model, inference is still tractable.

operations:  $\mathcal{O}(\max(mn^3, m^3))$   
bytes of storage:  $\mathcal{O}(\max(n^2, m^2))$

## Introduction

### Gaussian Process Model

A Gaussian process (GP) is a probability distribution for an unknown function

- Colloquially, an infinite dimensional multivariate normal distribution

$$Z(x) \sim GP[m(\cdot), C(\cdot, \cdot)]$$

$m(\cdot)$  is the mean function

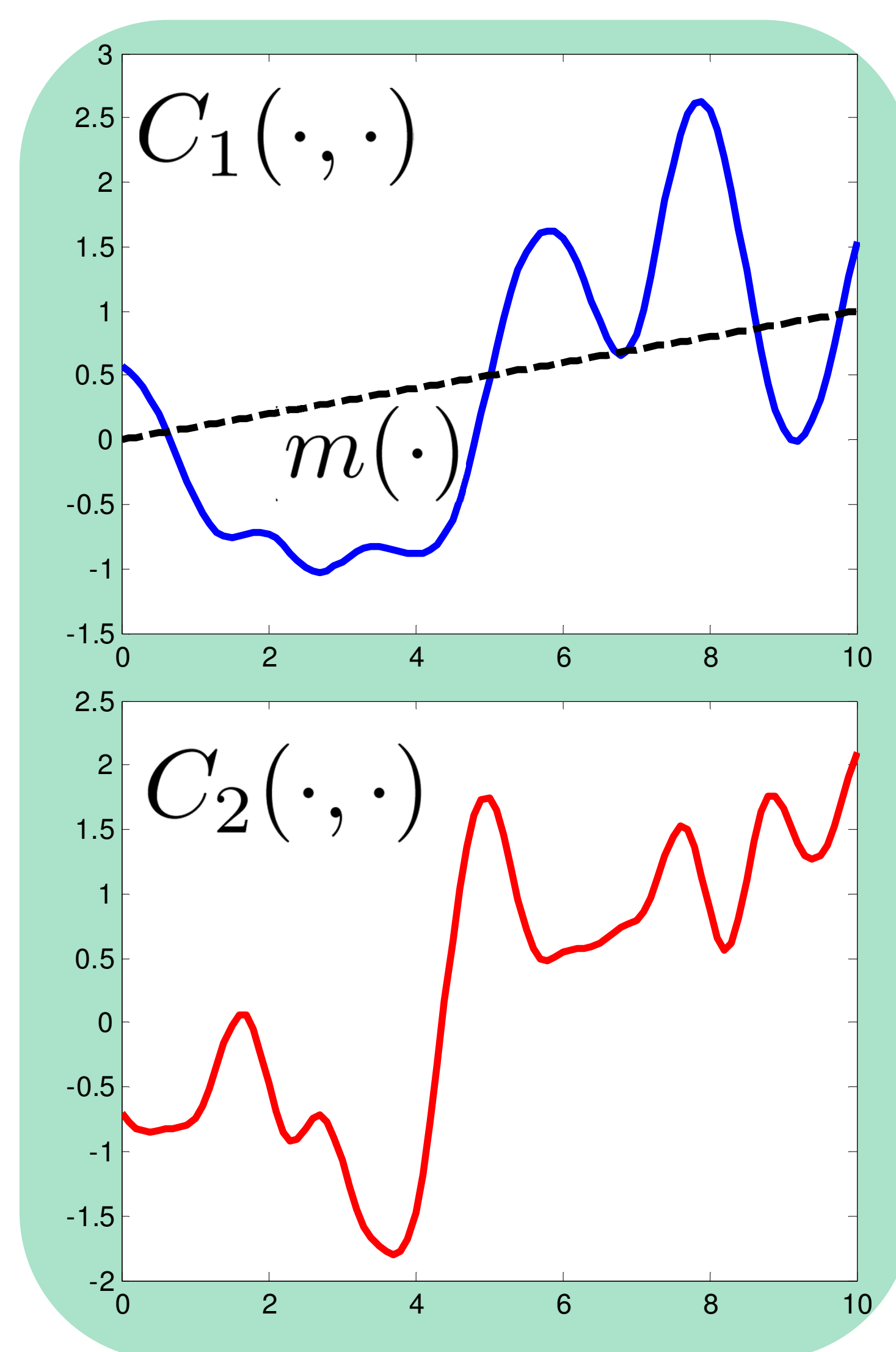
$C(\cdot, \cdot)$  is the covariance function

GP model for functional responses

$$y(x, t) = f^\top(x, t)\beta + z(x, t)$$

We observe responses:

$$Y = [y_1^\top, \dots, y_m^\top]^\top$$



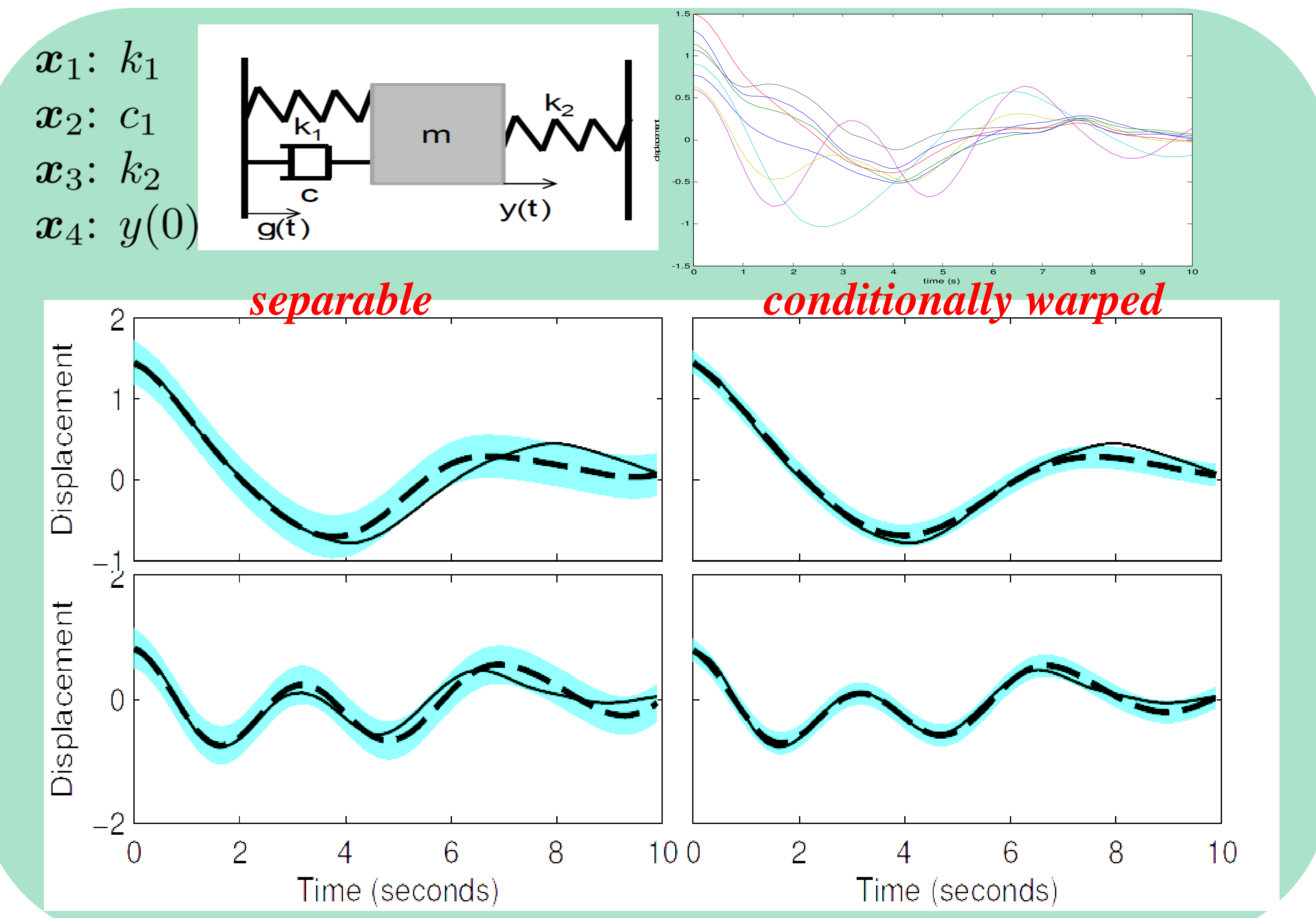
## Prediction

- Estimate the parameters using the likelihood (Maximum likelihood, Bayesian, etc.):

$$l \propto \exp\left[-\frac{1}{2} \left\{ \log |\Sigma| + (Y - F\beta)^\top \Sigma^{-1} (Y - F\beta) \right\}\right]$$

- Develop the best linear unbiased predictor to predict new responses:

$$\hat{y}(x_0, t_0) = f^\top(x_0, t_0)\beta + \sigma^\top(x_0, t_0) \Sigma^{-1} (Y - F\beta),$$



## References

The associated paper contains references used in this poster.  
Plumlee, M and Joseph, V. R. (2012). Tractable functional Response Modeling with Nonseparable and Nonstationary Covariance. *Submitted*