

Recap

- selection, fitting, and validation
- selection of a model = hypothesis class of function
- fitting - finding the best parameter set for a selected model
- validation - test selected and fitted model on a unseen dataset

Modeling: Gaussian processes

$$y = f(x) + \epsilon$$

$$f(x) = w^T x \quad (\text{linear model})$$

$$= w^T \phi(x) \quad (\text{polynomial } \phi(x) = [1 \ x \ x^2 \dots])$$

$$\epsilon \sim N(0, \sigma^2)$$

$$P(y|x, w) = \prod_{i=1}^n P(y_i | x_i, w) \in$$

$$= \prod_{i=1}^n \frac{1}{\sigma \sqrt{2\pi}} \exp\left(-\frac{(y_i - w^T x_i)^2}{2\sigma^2}\right)$$

Gaussian: sum, multiply, marginalization (over variable) conditional distribution.

↓
Gaussian

$$= N(w^T x, \sigma^2 I)$$

$w \sim N(0, \Sigma_p)$ → a sample would be a $p \times p$ matrix

$$P(w|y, X) = P(y|x, w) \times P(w)$$

$$= N(\bar{w}, \Sigma_w)$$

}

$$\begin{bmatrix} w_{11} & w_{12} \\ w_{21} & w_{22} \end{bmatrix}$$

$$w_{11} = N(\mu, \sigma^2)$$

$$w \sim N(0, \Sigma_p)$$

How are we going to predict at a unknown value x_* given our model $f(x_*)$ given (X, y)

$$P(f(x_*) | X, y, x_*)$$

$$= \int P(f_* | x_*, w) P(w | X, y) dw$$

model *posterior*

$$= N(\bar{w}_*, \Sigma_{w_*})$$

Gaussian processes:

$$f = w^T x$$

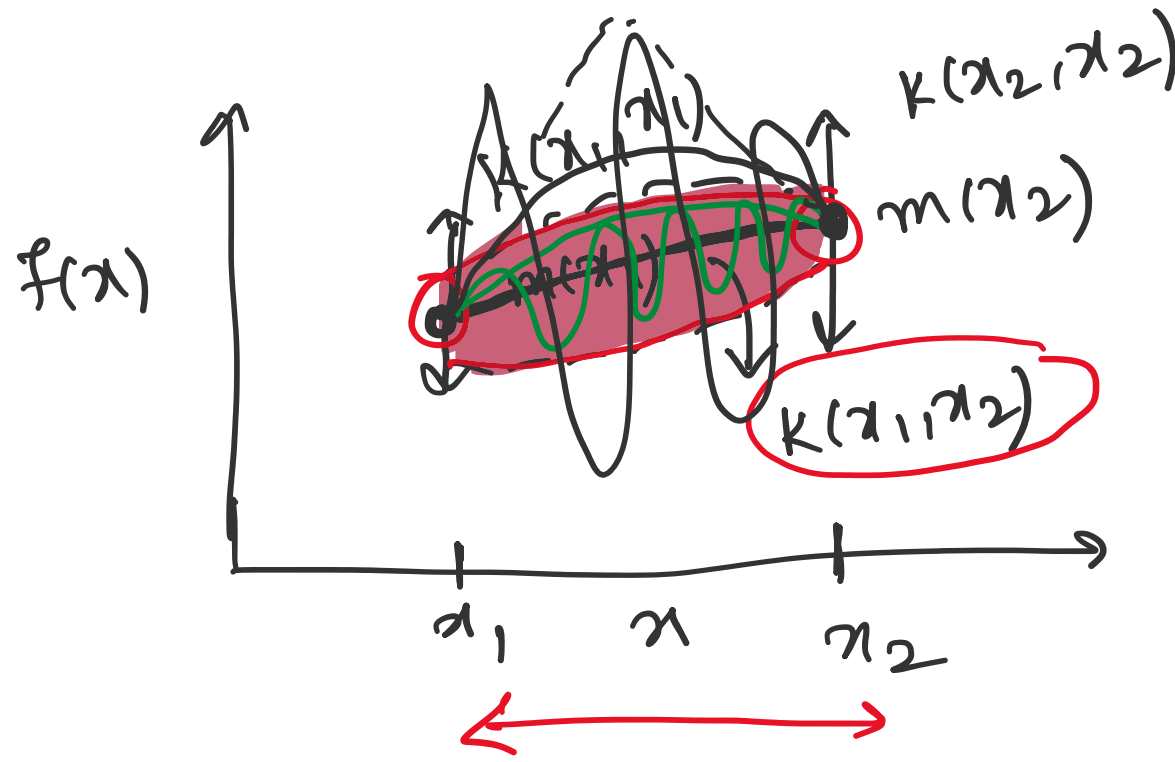
$$f = GP(m(x), k(x, x'))$$

mean covariance
function function

$$m(x) = ax + b$$

$$= a \quad (a=0)$$

$$k(x, x') = \text{smooth}$$



$$f(x) = \phi(x)^T w \Leftarrow$$

$$E[f(x)] = E[\phi(x)^T w]$$

$$= \phi(x)^T E[w] = m(x)$$

$$\Rightarrow E[f(x), f(x')] = \phi(x)^T E[ww^T] \phi(x')$$

$$= k(x, x')$$

$$E[(f(x) - f(x')) (f(x) - f(x'))^T]$$

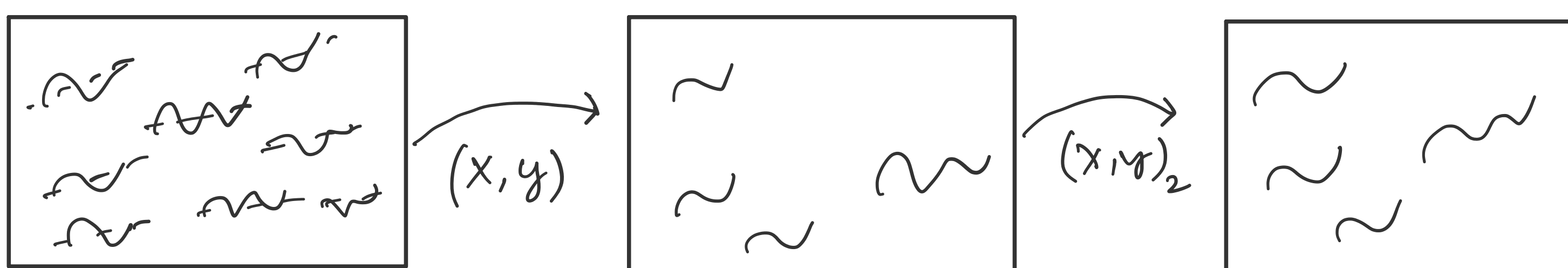
$$m(x) = 0$$

$k(x, x') \leftarrow$ GP choices is the covariance.
(prior knowledge)

- smooth, smoothness defined by length scale, sharp/peaky functions.

$$f(x_1) \leftrightarrow f(x_2)$$

$$f(x) \sim GP(m(x), k(x, x'))$$



$p(f)$ being the model given data

