

NONLINEAR FEATURE MAPS

Sometimes linear methods are not adequate.
One way to create nonlinear estimators or classifiers is to first transform the data via a nonlinear feature map

$$\Phi: \mathbb{R}^d \rightarrow \mathbb{R}^{d'}$$

and apply a linear method to the transformed data $\Phi(x_1), \dots, \Phi(x_n)$

Regression

In regression, the model becomes

$$f(x) = \beta^T \Phi(x) + \beta_0, \quad \beta \in \mathbb{R}^{d'}$$

Exercise | Suppose $f(x)$ is a cubic polynomial.
Determine the least-squares estimate of f
given $(x_1, y_1), \dots, (x_n, y_n)$.

Solution

$$f(x) = \beta_3 x^3 + \beta_2 x^2 + \beta_1 x + \beta_0$$

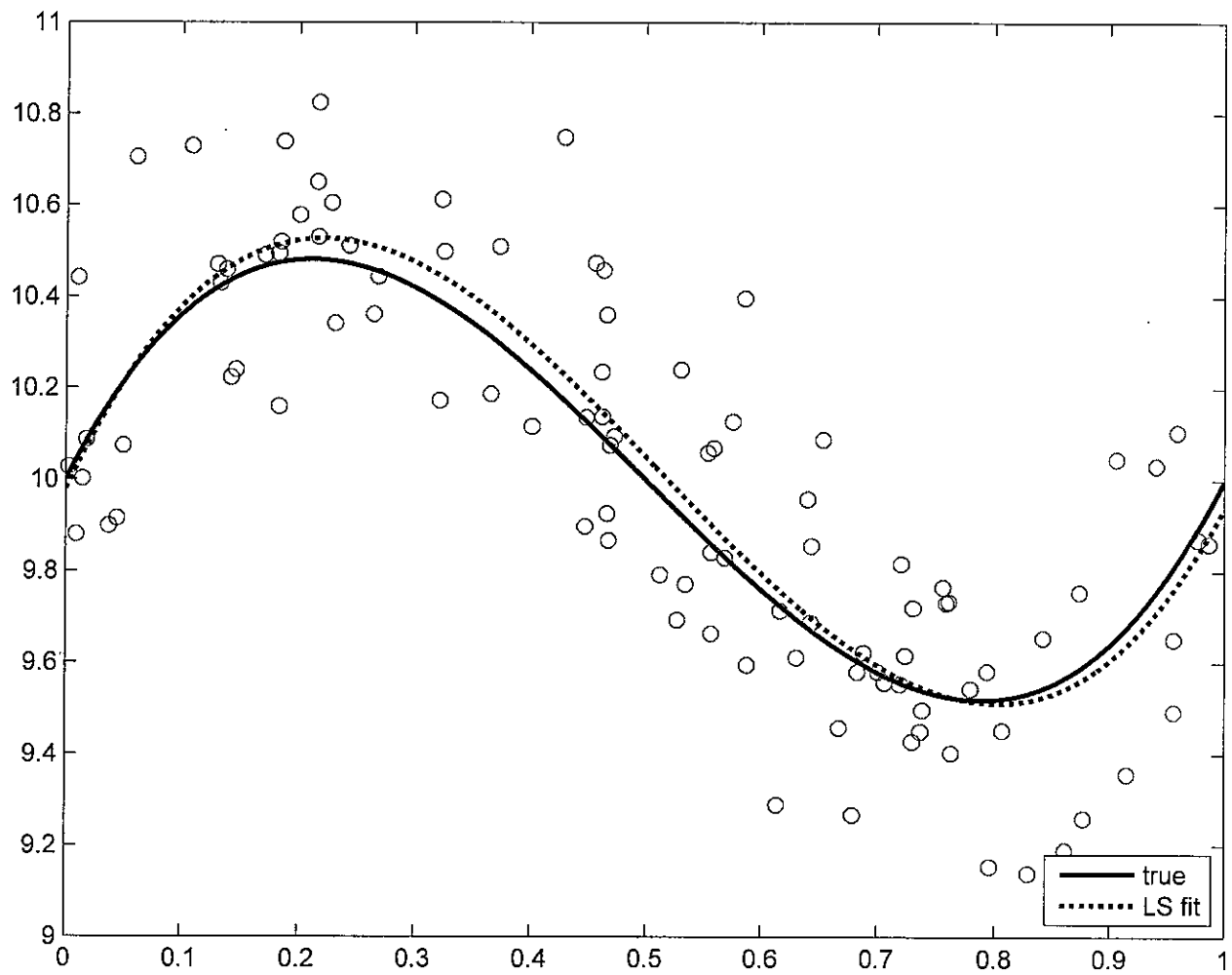
$$\Rightarrow \Phi_k(x) = x^k$$

$$\Rightarrow A = \begin{bmatrix} 1 & x_1 & x_1^2 & x_1^3 \\ 1 & x_2 & x_2^2 & x_2^3 \\ \vdots & \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 & x_n^3 \end{bmatrix}$$

$$\Rightarrow \hat{\theta} = (A^T A)^{-1} A^T \underline{y}$$

gives the LS cubic polynomial fit.

What if a polynomial model is also not appropriate, or the degree is unknown? We'll address these and other issues later in the course.



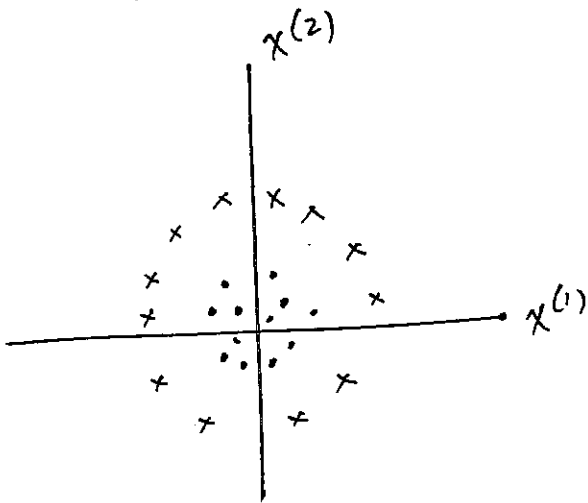
Classification

In classification, the classifier is

$$\text{sign}\{w^T \Phi(x) + b\}$$

where $w \in \mathbb{R}^{d'}$.

Example 1



$$x \mapsto \Phi(x) := \begin{bmatrix} 1 \\ x^{(1)} \\ x^{(2)} \\ x^{(1)}x^{(2)} \\ (x^{(1)})^2 \\ (x^{(2)})^2 \end{bmatrix}$$

Then the data are linearly separable in the new feature space. They are correctly classified by

$$\text{sign}\{w^T \Phi(x)\}$$

where

$$w = \begin{bmatrix} \\ \\ \\ \\ \\ \end{bmatrix}$$