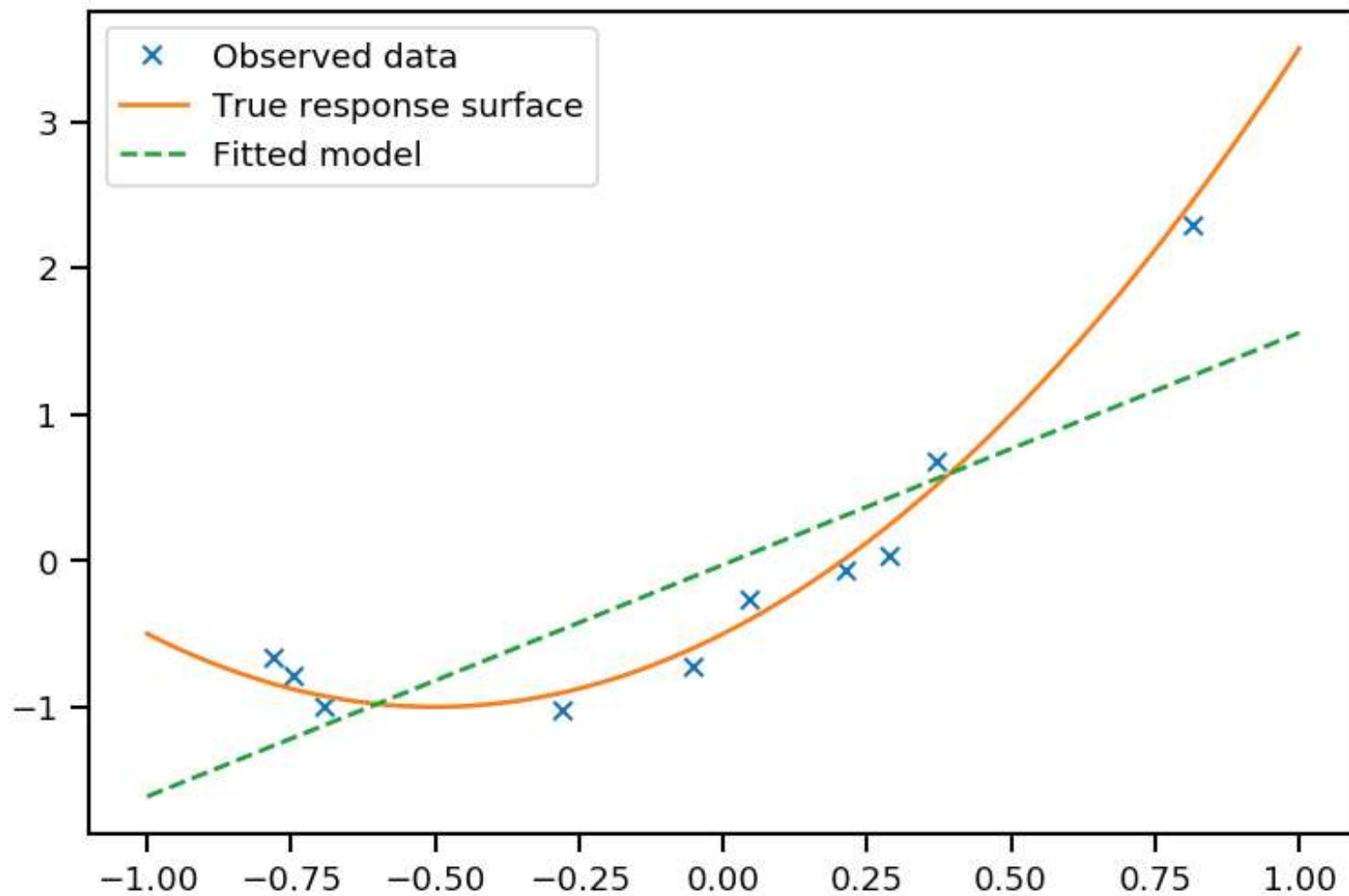


# Lecture 13: Linear Regression via Least Squares

Professor Ilias Bilonis

## Polynomial regression

# An example that doesn't work



# Regression model

$$y = w_0 + w_1 \cdot x + w_2 \cdot x^2$$

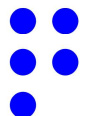
$$\underline{w} = (w_0, w_1, w_2)$$

# Least squares loss function

$$L(\underline{w}) = \sum_{i=1}^n \left( \overset{\text{target}}{y_i} - \underbrace{w_0 - w_1 \cdot x_i - w_2 \cdot x_i^2}_{\text{prediction}} \right)^2$$

$$= \left\| \underline{y} - \left( \underline{X} \right) \underline{w} \right\|_2^2$$

$$\underline{X} = \begin{pmatrix} 1 & x_1 & x_1^2 \\ \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 \end{pmatrix} \quad (n \times \underline{\underline{3}})^{\text{degree} + 1}$$

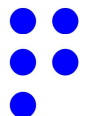


# Minimizing the loss function

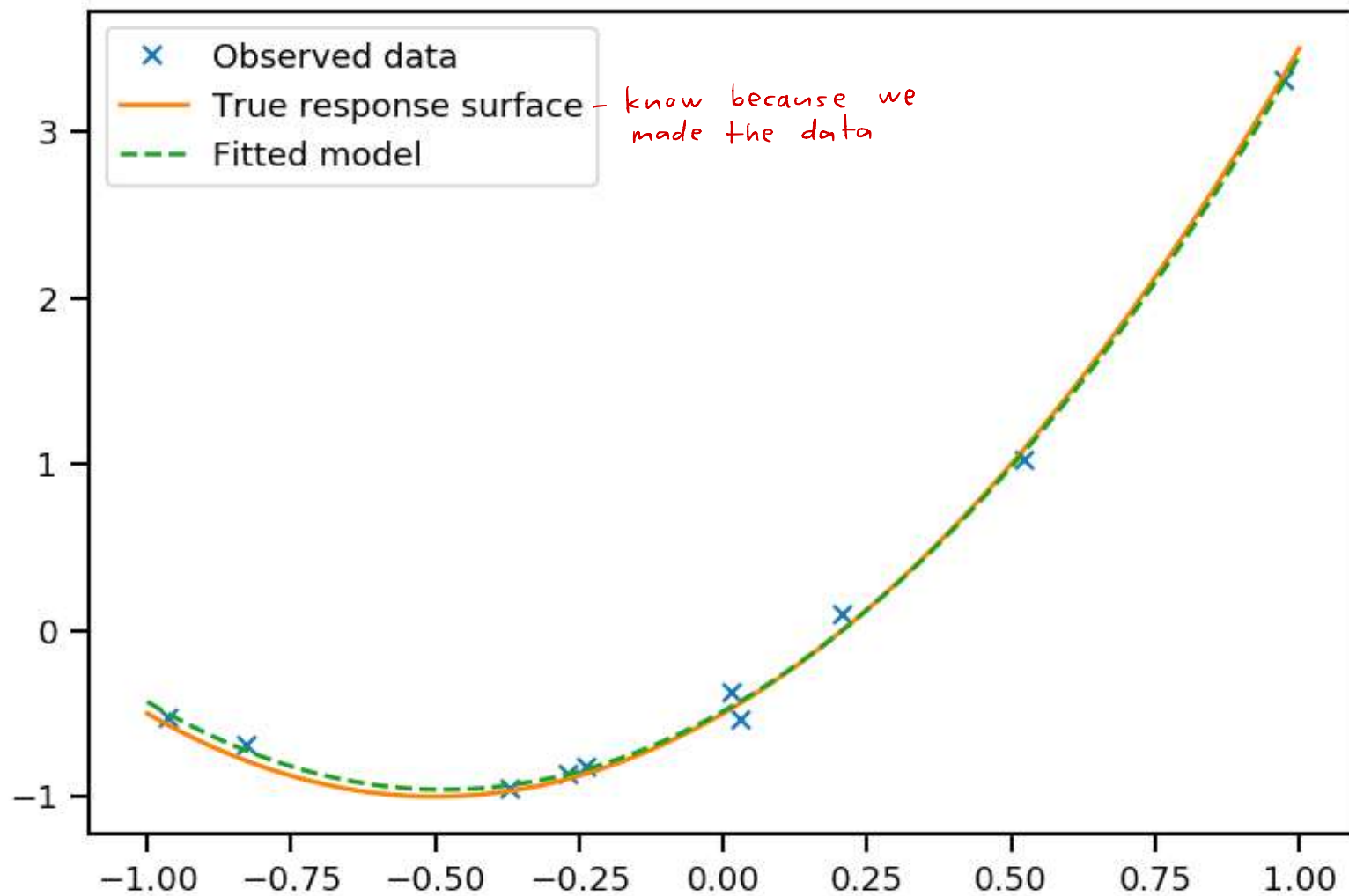
$$L(\underline{w}) = \|\underline{y} - \underline{X} \cdot \underline{w}\|_2^2$$

$$\nabla_{\underline{w}} L(\underline{w}) = 0 \Rightarrow \underbrace{\left( \underline{X}^T \underline{X} \right)}_{\text{Linear System}} \underline{w} = \underline{X}^T \underline{y}$$

/ solve for



# Example



# Higher degree polynomials

column in design matrix for each

$$y = w_0 + w_1 x + w_2 x^2 + \dots + w_p \cdot x^p$$

$$L(\underline{w}) = \sum_{i=1}^n (y_i - \underbrace{w_0 + w_1 x_i + w_2 x_i^2 + \dots + w_p x_i^p}_{\text{prediction}})^2$$

$$= \|\underline{y} - \underline{X} \cdot \underline{w}\|_2^2$$

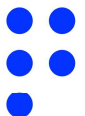
$$\underline{w} = (w_0, w_1, \dots, w_p)$$

$$\underline{X} = \begin{pmatrix} 1 & x_1 & x_1^2 & \dots & x_1^p \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^p \end{pmatrix} \quad (n \times (p+1))$$

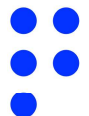
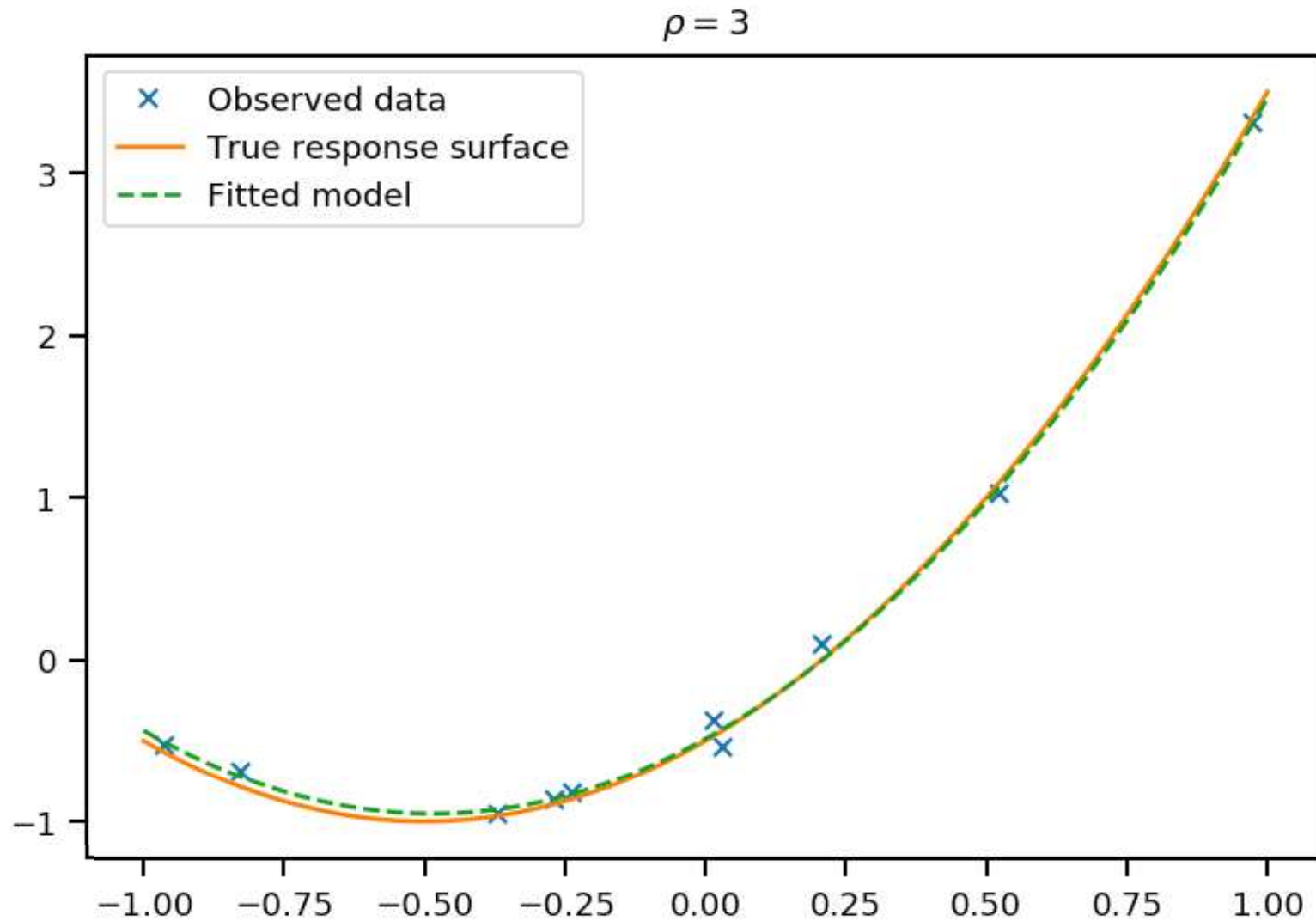
$$\nabla_{\underline{w}} L(\underline{w}) = 0 \Rightarrow$$

$$\underline{X}^T \underline{X} \cdot \underline{w} = \underline{X}^T \cdot \underline{y}$$

same form

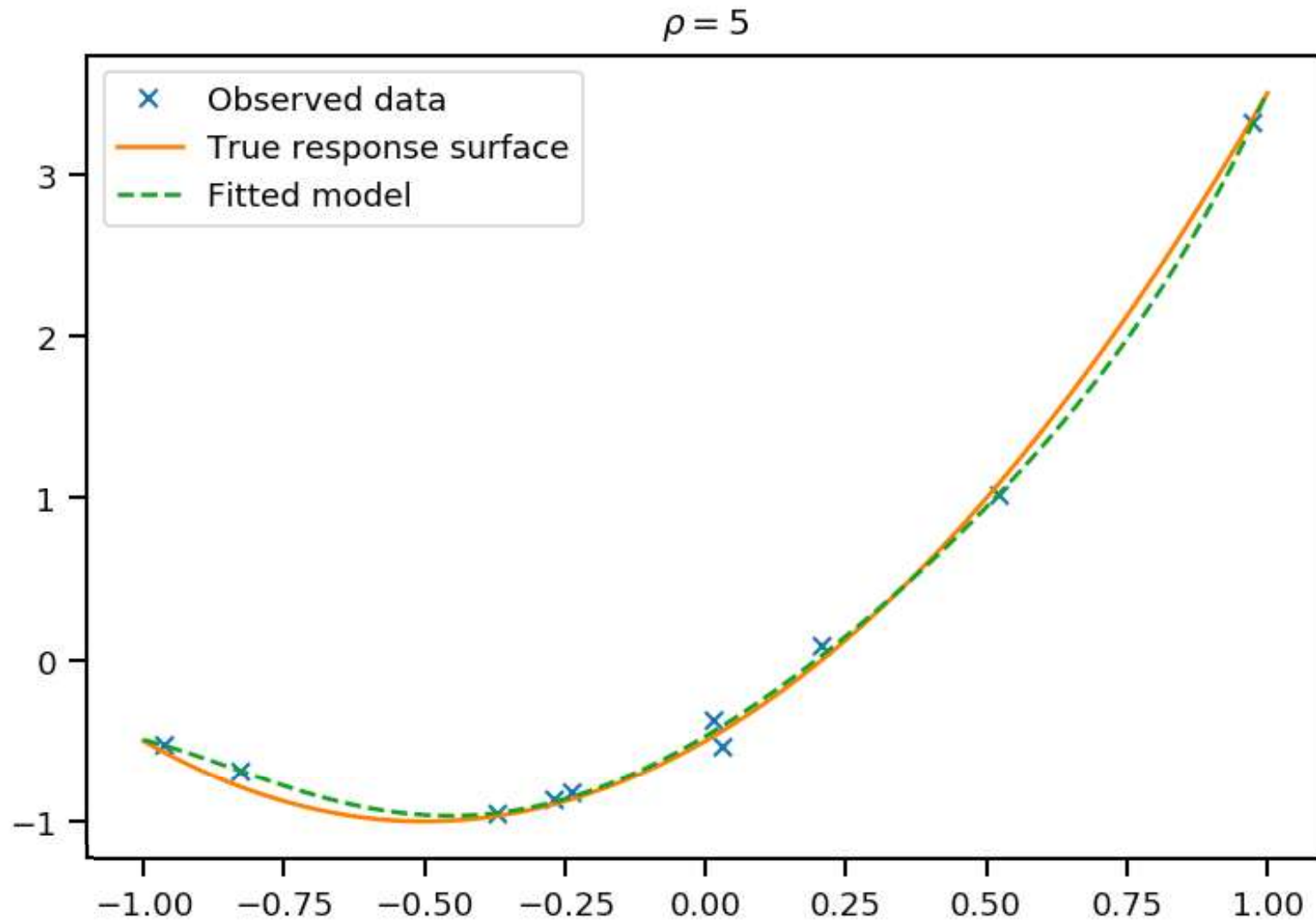


# Example





# Example



# Example

