

$$D = \{(x_1, y_1), \dots, (x_n, y_n)\}$$

$$g(x) = w^T x + w_0 \quad J(w) = SSE = \sum_{i=1}^n (\underbrace{w^T x_i + w_0}_{g(x_i)} - y_i)^2$$

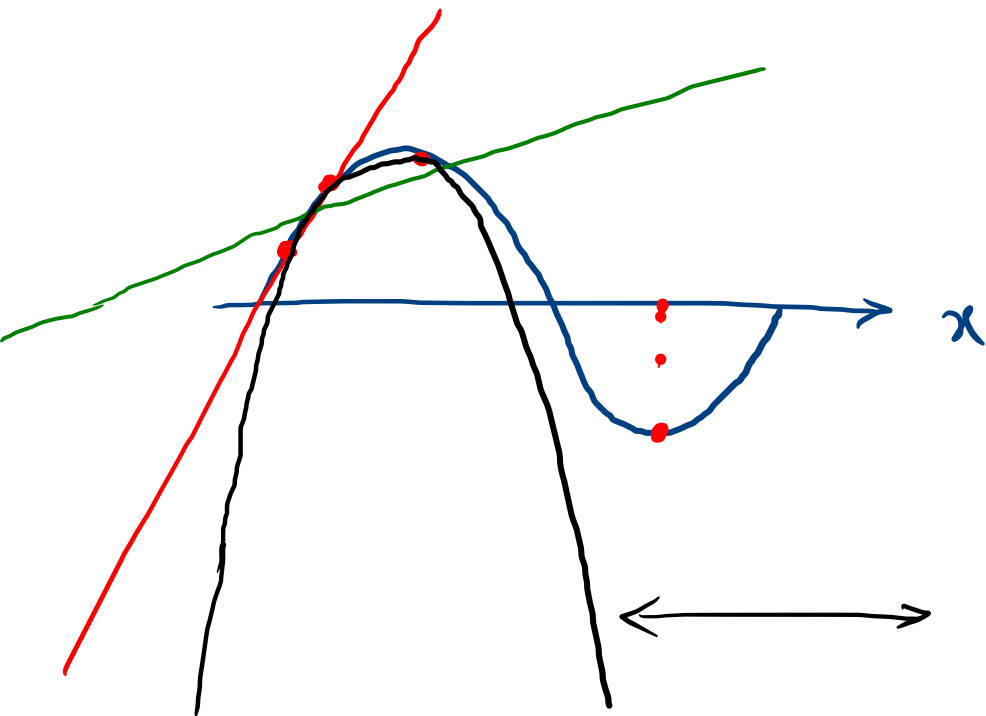
$$Err = \text{Bias} + \text{Variance}$$

↓

$$g(x) = \underline{\underline{w_2}} x + \underline{\underline{w_0}}$$

$$g(x) = \underline{\underline{w_2}} x^2 + \underline{\underline{w_1}} x + \underline{\underline{w_0}}$$

—

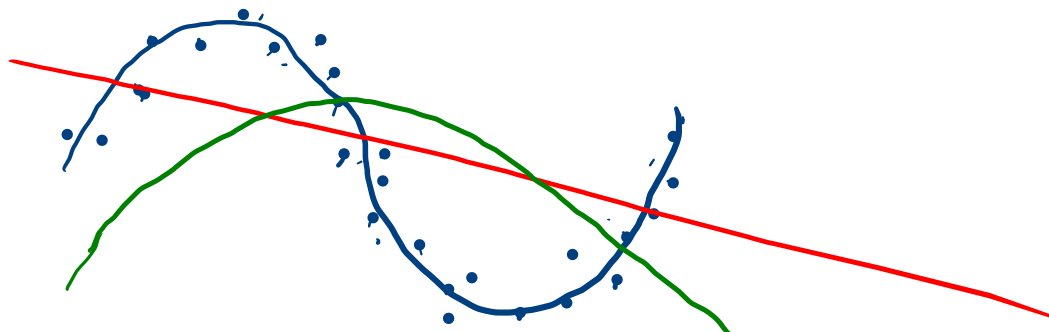


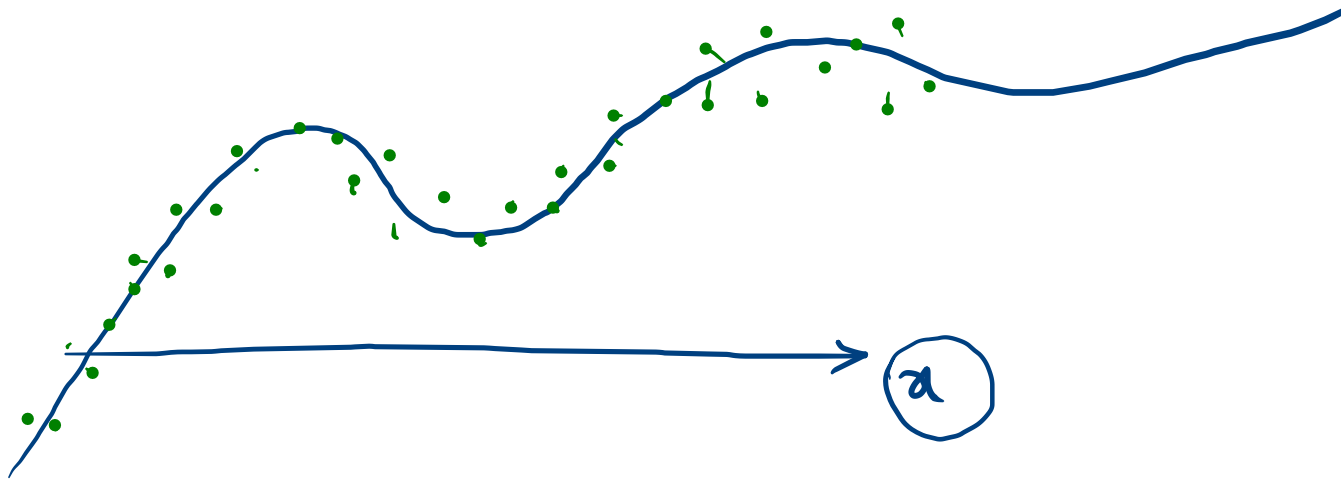
$$f(x) = \sin \pi x$$

$$g(x) = \omega_1 x + \omega_0$$

$$|\omega_1| < \theta$$

$$g(x) = \omega_2 x^2 + \omega_1 x + \omega_0$$





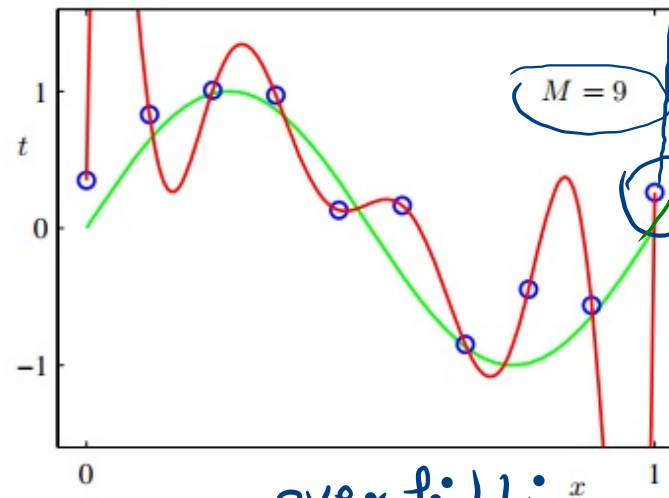
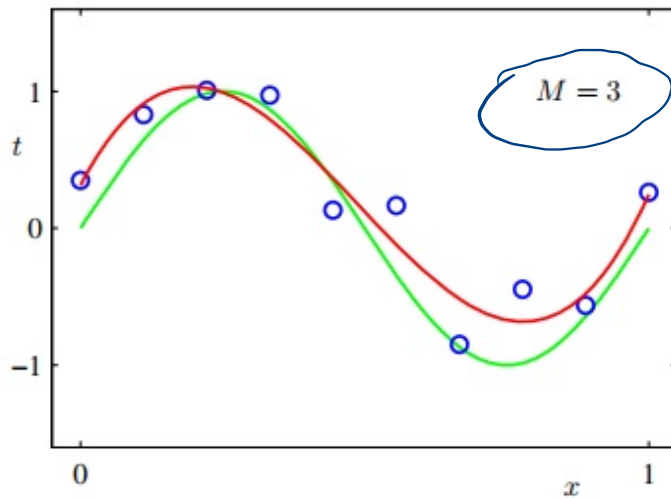
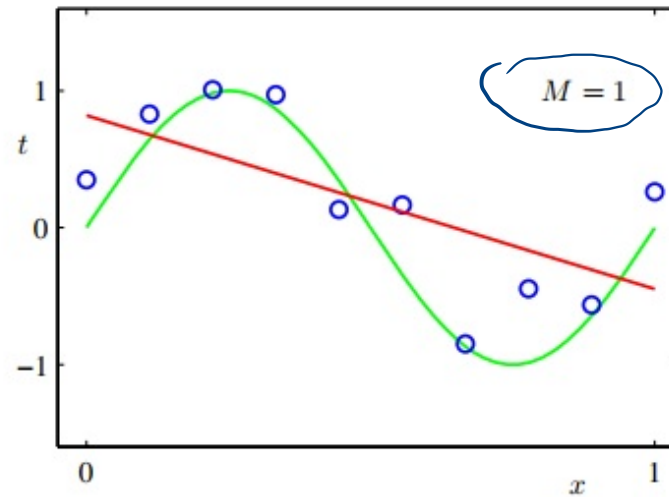
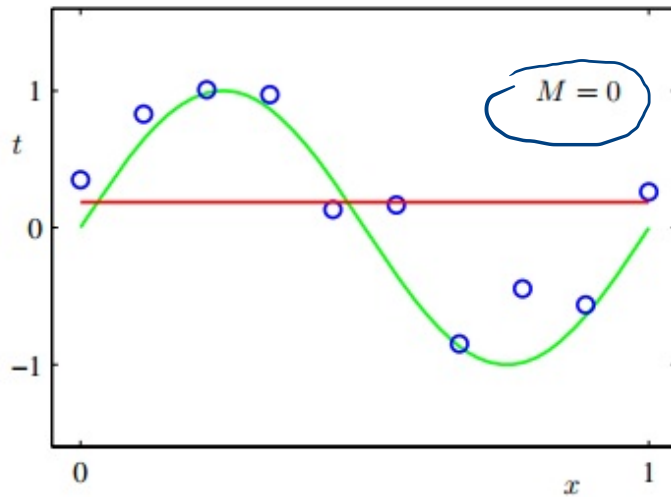
$$g(x) = w_4 x^4 + w_3 x^3 + w_2 x^2 + w_1 x + w_0$$

$$\underline{\underline{\phi(x) = \begin{bmatrix} 1 \\ x \\ x^2 \\ x^3 \\ x^4 \end{bmatrix}}}$$

$$g(x) = \omega^T \phi(x)$$
$$\omega = \begin{bmatrix} w_0 \\ w_1 \\ w_2 \\ w_3 \\ w_4 \end{bmatrix}$$

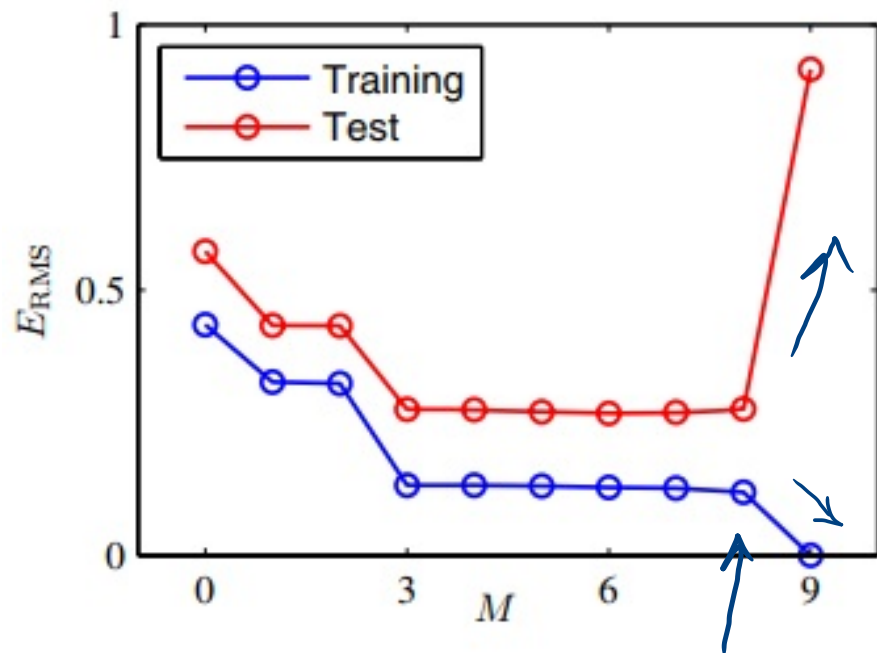
$$y(x, \mathbf{w}) = w_0 + w_1x + w_2x^2 + \dots + w_Mx^M = \sum_{j=0}^M w_jx^j$$

\mathcal{M}



over fitting

\mathcal{M}



	$M = 0$	$M = 1$	$M = 3$	$M = 9$
w_0^*	<u>0.19</u>	0.82	0.31	0.35
w_1^*		<u>-1.27</u>	7.99	232.37
w_2^*			-25.43	-5321.83
w_3^*			<u>17.37</u>	48568.31
w_4^*				-231639.30
w_5^*				640042.26
w_6^*				-1061800.52
w_7^*				1042400.18
w_8^*				-557682.99
w_9^*				<u>125201.43</u>

$$|w_i| < \theta_i$$

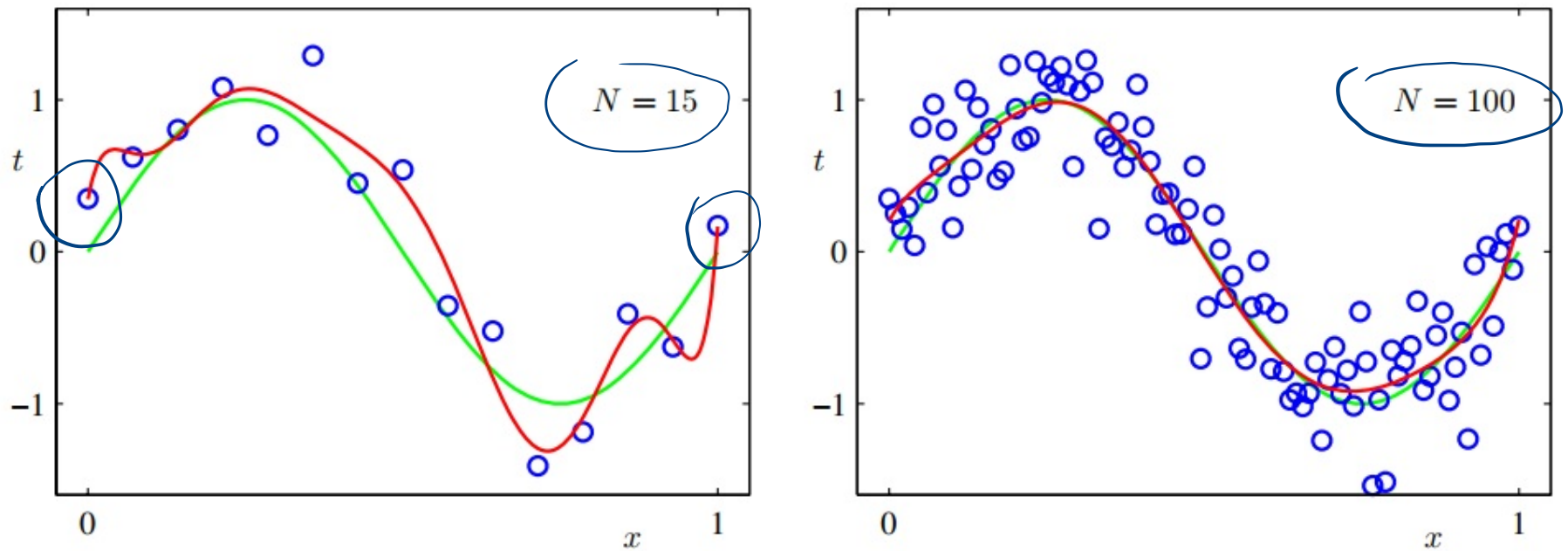


Figure 1.6 Plots of the solutions obtained by minimizing the sum-of-squares error function using the $M = 9$ polynomial for $N = 15$ data points (left plot) and $N = 100$ data points (right plot). We see that increasing the size of the data set reduces the over-fitting problem.

$$g(x) = \sum_{i=1}^M w_i x^i$$

$$\underline{J(w)} = \sum_{j=1}^n (g(x_j) - y_j)^2$$

$$\|w\|_2^2 \leq C$$

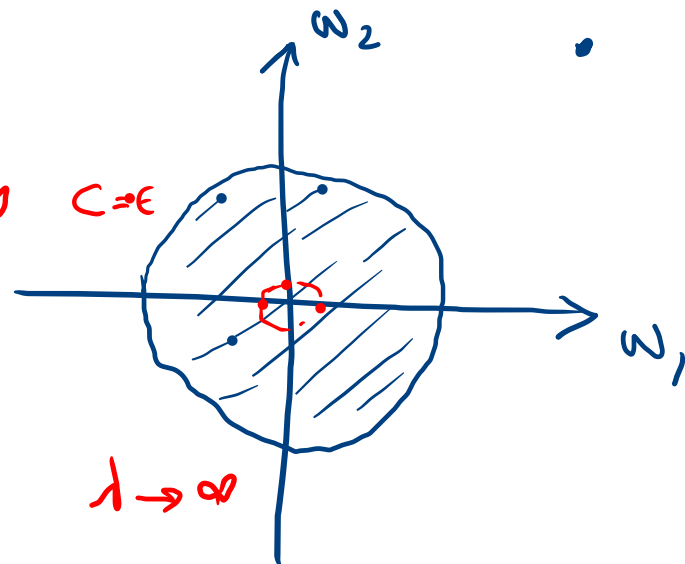
$$\min_w J(w)$$

subject to:

$$\|w\|_2^2 \leq \underline{C}$$

$$\min_w \underline{J(w) + \lambda \|w\|_2^2}$$

$$g(x) = w_1 x_1 + w_2 x_2$$



$$\omega^* = \arg \min_{\omega} \underbrace{J(\omega)} + \underbrace{\lambda \|\omega\|_2^2}$$

ridge regression

regularization term

$$\|\omega\|_2^2 = \sum_{i=1}^d \omega_i^2$$

$$\omega_i = 0$$

$$\begin{array}{l} \omega^* = \arg \min_{\omega} J(\omega) \\ \text{s.t. } \boxed{\|\omega\|_0 \leq \underline{C}} \end{array}$$

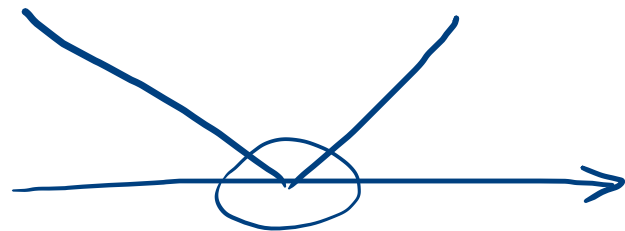
\Rightarrow

$$\omega^* = \arg \min_{\omega} \underbrace{J(\omega) + \lambda \|\omega\|_0}_{\downarrow}$$

$$\omega^* = \arg \min_{\omega} J(\omega) + \lambda \underline{\underline{\|\omega\|_1}}$$

$$\|\omega\|_1 = \sum_{i=1}^d |\omega_i|$$

$$\begin{aligned} & a\omega^2 + b\omega + c \\ & = \\ & a > 0 \end{aligned}$$



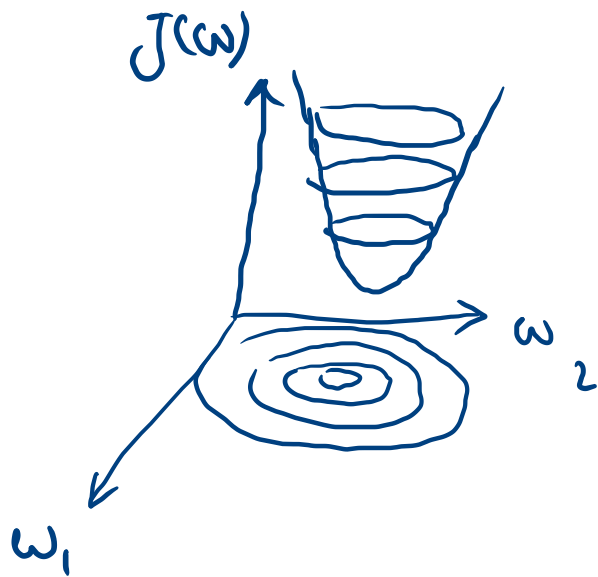
$$g(x) = \omega_1 x_1 + \omega_2 x_2$$

$$J(\omega) = \underline{\underline{\sum_{i=1}^n (\omega^T x_i - y_i)^2}}$$

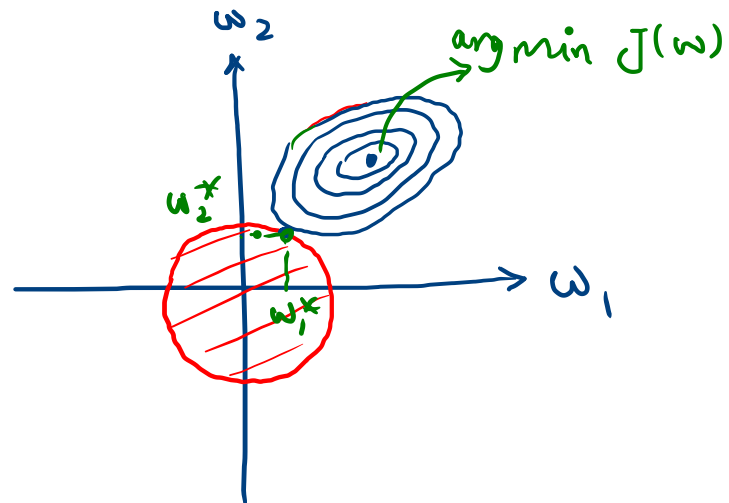
$$J(\omega) = \omega^T A \omega + b^T \omega + c$$

quadratic form

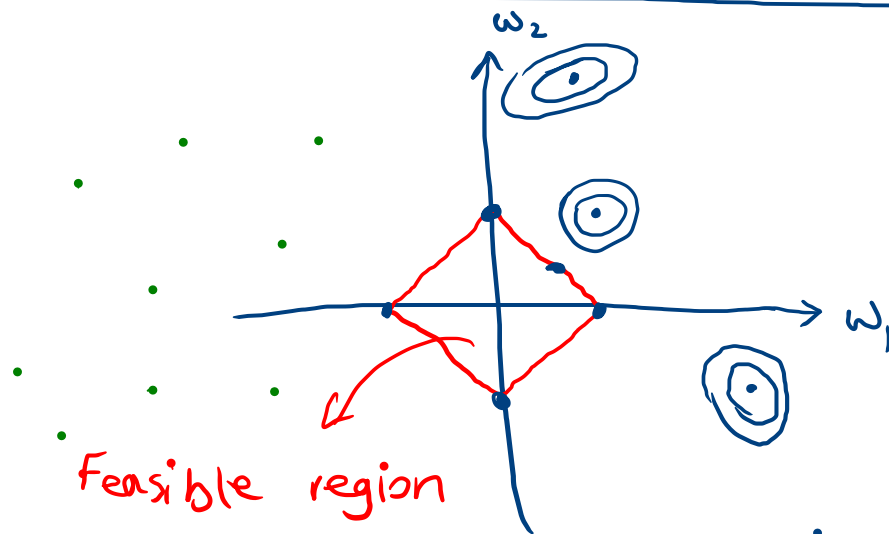
$$A \succ 0$$



$$\begin{aligned} \min_{\mathbf{w}} \quad & J(\mathbf{w}) \\ \text{s.t.} \quad & \underline{\underline{\|\mathbf{w}\|_2^2 \leq C}} \end{aligned}$$



$$\begin{aligned} \min_{\mathbf{w}} \quad & J(\mathbf{w}) \\ \text{s.t.} \quad & \|\mathbf{w}\|_1 \leq C \\ & \downarrow \\ & |w_1| + |w_2| \leq C \end{aligned}$$



$$\min_{\omega} \underbrace{\sum_{i=1}^n (\omega^T x_i - y_i)^2}_{J(\omega)} + \overset{\uparrow}{\lambda} \|\omega\|_1$$

LASSO

$$J(w) + \lambda \|w\|_2^2$$

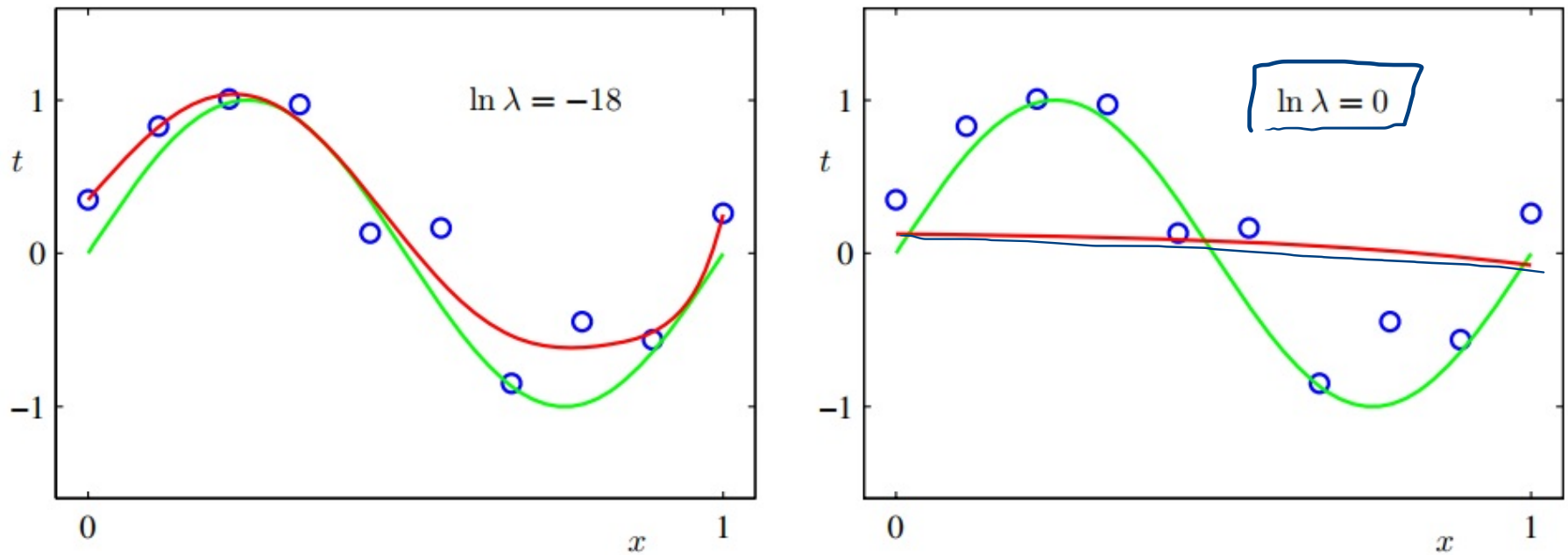


Figure 1.7 Plots of $M = 9$ polynomials fitted to the data set shown in Figure 1.2 using the regularized error function (1.4) for two values of the regularization parameter λ corresponding to $\ln \lambda = -18$ and $\ln \lambda = 0$. The case of no regularizer, i.e., $\lambda = 0$, corresponding to $\ln \lambda = -\infty$, is shown at the bottom right of Figure 1.4.

Table 1.2 Table of the coefficients w^* for $M = 9$ polynomials with various values for the regularization parameter λ . Note that $\ln \lambda = -\infty$ corresponds to a model with no regularization, i.e., to the graph at the bottom right in Figure 1.4. We see that, as the value of λ increases, the typical magnitude of the coefficients gets smaller.

	$\ln \lambda = -\infty$	$\ln \lambda = -18$	$\ln \lambda = 0$
w_0^*	0.35	0.35	0.13
w_1^*	232.37	4.74	-0.05
w_2^*	-5321.83	-0.77	-0.06
w_3^*	48568.31	-31.97	-0.05
w_4^*	-231639.30	-3.89	-0.03
w_5^*	640042.26	55.28	-0.02
w_6^*	-1061800.52	41.32	-0.01
w_7^*	1042400.18	-45.95	-0.00
w_8^*	-557682.99	-91.53	0.00
w_9^*	125201.43	72.68	0.01

بدون نرم منظم ساز

Figure 1.8 Graph of the root-mean-square error (1.3) versus $\ln \lambda$ for the $M = 9$ polynomial.

train | val | (test)
Data

