Finding the Compact Derivative

Let's first review some key notations, remember how we defined a function?

$$f(x) = w_1 \phi_1(x) + w_2 \phi_2(x) + w_3 \phi_3(x) + \dots = \underbrace{\begin{bmatrix} w_1 & w_2 & w_3 & \dots \end{bmatrix}}_{w^\top} \underbrace{\begin{bmatrix} \phi_1(x) \\ \phi_2(x) \\ \dots \end{bmatrix}}_{\phi(x)} = \underbrace{w^\top \phi(x) = \phi(x)^\top w}_{\text{they are equivalent}}.$$

Next, if we have n samples consists of $\{x_1, x_2, x_3, ...\}$, then our prediction of each sample would be

$$\begin{bmatrix} f(x_1) \\ f(x_2) \\ f(x_3) \\ \dots \end{bmatrix} = \begin{bmatrix} \phi(x_1)^\top w \\ \phi(x_2)^\top w \\ \phi(x_3)^\top w \\ \dots \end{bmatrix} = \underbrace{\begin{bmatrix} \phi(x_1)^\top \\ \phi(x_2)^\top \\ \phi(x_3)^\top \\ \dots \end{bmatrix}}_{\text{Let's call it } \Phi} w = \Phi w.$$

The key takeaway is that everyone of our prediction can be denoted simply as Φw and that

$$\Phi = \begin{bmatrix} \phi(x_1)^\top \\ \phi(x_2)^\top \\ \phi(x_3)^\top \\ \dots \end{bmatrix}$$

Given that we are solving a regression problem, the derivative can be written as

$$\mathcal{L}(w) = \frac{1}{2} \sum_{i}^{N} (w^{\top} \phi(x_{i}) - y_{i})^{2} \tag{4}$$

$$\frac{d\mathcal{L}}{dw} = \sum_{i}^{N} (w^{\top} \phi(x_{i}) - y_{i}) \phi(x_{i}) \tag{5}$$

$$= (w^{\top} \phi(x_{1}) - y_{1}) \phi(x_{1}) + (w^{\top} \phi(x_{2}) - y_{2}) \phi(x_{2}) + (w^{\top} \phi(x_{3}) - y_{3}) \phi(x_{3}) + \dots \tag{6}$$
You can find $\frac{d\mathcal{L}}{dw}$ using this equation. But this is way too slow. Let's see a faster trick. Det
$$a_{i} = (w^{\top} \phi(x_{i}) - y_{i}) = (\phi(x_{i})^{\top} w - y_{i}) \tag{7}$$
then
$$\frac{d\mathcal{L}}{dw} = a_{1} \phi(x_{1}) + a_{2} \phi(x_{2}) + a_{3} \phi(x_{3}) + a_{4} \phi(x_{4}) \tag{8}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} a_{1} \\ a_{2} \\ a_{3} \\ a_{4} \end{bmatrix} \tag{9}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} (\phi(x_{1})^{\top} w - y_{1}) \\ (\phi(x_{2})^{\top} w - y_{2}) \\ (\phi(x_{3})^{\top} w - y_{4}) \end{bmatrix} \tag{10}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} \left[\phi(x_{1})^{\top} w - y_{1}\right] \\ \phi(x_{2})^{\top} \\ \phi(x_{3})^{\top} \\ \phi(x_{3})^{\top} \end{bmatrix} \tag{11}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} \left[\phi(x_{1})^{\top} w - y_{1}\right] \\ \phi(x_{2})^{\top} \\ \phi(x_{3})^{\top} \\ \phi(x_{3})^{\top} \end{bmatrix} \tag{11}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} \left[\phi(x_{1})^{\top} w - y_{1}\right] \\ \phi(x_{2})^{\top} \\ \phi(x_{3})^{\top} \\ \phi(x_{3})^{\top} \end{bmatrix} \tag{12}$$

$$= \left[\phi(x_{1}) \quad \phi(x_{2}) \quad \phi(x_{3}) \quad \phi(x_{4})\right] \begin{bmatrix} \left[\phi(x_{1})^{\top} w - y_{1}\right] \\ \phi(x_{2})^{\top} \\ \phi(x_{3})^{\top} \\ \phi(x_{3})^{\top} \end{bmatrix} \tag{11}$$