

Assume: $\theta^1 = \alpha_1 \cdot \phi(x')$ PS2) 2)

$$\theta^2 = \alpha_2 \cdot \phi(x^2) + \alpha_1 \cdot \phi(x')$$

$$\theta^3 = \alpha_3 \cdot \phi(x^3) + \alpha^2 \phi(x^2) + \alpha_1 \cdot \phi(x')$$

Training: $\Rightarrow \theta = \sum_i \alpha_i \phi(x^i)$ where $\alpha_i \in \{-1, 0, +1\}$

~~the weights~~ $\in \{-1, 0, +1\}$

Prediction: $\theta^\top \cdot \phi(x)$

$$\Rightarrow h_\theta\left(\frac{z}{k}\right) = g\left(\sum_i \alpha_i \phi(x^i)^\top \cdot \phi(z)\right) = g\left(\sum_i \alpha_i K(x^i, z)\right)$$

Insight \Rightarrow only need to learn all α_i ! ~~from~~

\Rightarrow how to train and find α_i ?

Example: α_2 :

$$\alpha_2 = y^{(2)} - g(\alpha_1 \phi(x')^\top \phi(x^2)) = y^{(2)} - g(\alpha_1 K(x', x^2))$$

$$\Rightarrow \alpha_{i+1} = \alpha(y^{(i+1)} - g(\sum_{p=1}^i \alpha_p \phi(x^p)^\top \phi(x^{i+1}))) = \alpha(y^{(i+1)} - g(\sum_{p=1}^i \alpha_p K(x^p, x^{i+1})))$$

Update Rule: $\alpha_{i+1} = \alpha(y^{(i+1)} - g(\sum_{p=1}^i \alpha_p K(x^{(p)}, x^{(i+1)})))$ for $i \geq 1$

Zero State: $\alpha_1 = \alpha(y^{(1)} - g(\vec{0}^\top \phi(x'))) = y^{(1)} - g(K(\vec{0}, x'))$
 $\alpha(y^{(1)} - g(0)) = \alpha(y^{(1)} - 1)$

\Rightarrow a) θ_a^{i+1} is represented by the $x_{1 \dots i}$
and the prev data points leading up
to $i+1$