

# Regression solution: simple matrix math

$$\mathbf{w}^* = \arg \min_{\mathbf{w}} \underbrace{(\mathbf{H}\mathbf{w} - \mathbf{t})^T (\mathbf{H}\mathbf{w} - \mathbf{t})}_{\text{residual error}}$$

$$\text{solution: } \mathbf{w}^* = \underbrace{(\mathbf{H}^T \mathbf{H})^{-1}}_{\mathbf{A}^{-1}} \underbrace{\mathbf{H}^T \mathbf{t}}_{\mathbf{b}} = \mathbf{A}^{-1} \mathbf{b}$$

$$\text{where } \mathbf{A} = \mathbf{H}^T \mathbf{H} = \underbrace{\begin{bmatrix} \square & \square & \square & \square \\ \square & \square & \square & \square \\ \square & \square & \square & \square \\ \square & \square & \square & \square \end{bmatrix}}_{\substack{\text{k} \times \text{k} \text{ matrix} \\ \text{for k basis functions}}} \quad \mathbf{b} = \mathbf{H}^T \mathbf{t} = \underbrace{\begin{bmatrix} \square \\ \square \\ \square \\ \square \end{bmatrix}}_{\text{k} \times 1 \text{ vector}}$$