

# Nonlinear Basis Functions

a) Monomials in  $\mathbb{R}^N$  of highest degree  $k$ :

$$\phi_i(\underline{x}) \in \{1, x_1, x_2, \dots, x_N, x_1^2, x_1 x_2, x_1 x_3, \dots, x_1 x_N, \dots, x_N^2, \dots, x_1 x_N^{k-1}, x_1^k, \dots, x_N^k\}$$

$$= \left\{ \prod_{j=1}^N x_j^{a_j} : a_j \in \mathbb{N}_0 \text{ with } 0 \leq \sum_{j=1}^N a_j \leq k \right\}, i=1, \dots, d$$

Homework:  $k=9 (\Rightarrow d=55)$ ,  $N=2$

b) Sphering:

$\phi_i(\underline{x})$  can grow fast for large  $\|\underline{x}\| \Rightarrow$  whitening (sphering) which includes standardization

$\Rightarrow$  scale and rotate centered data s.t. resulting ("sphered") data is uncorrelated with standard deviation 1 in each variable

data pts.  $\alpha$ :

$$\underline{x}_{\text{sphered}}^{(\alpha)} = \underline{\Lambda}^{-\frac{1}{2}} \underline{E}^T \underline{x}_{\text{centered}}^{(\alpha)}$$

where  $\underline{x}_{\text{centered}}^{(\alpha)} = \underline{x}^{(\alpha)} - \bar{\underline{x}}$  ( $\bar{\underline{x}} = \frac{1}{P} \sum_{\alpha=1}^P \underline{x}^{(\alpha)}$ )

$\underline{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_N)$  eigenvalue matrix,  $\underline{E} = (\underline{e}_1, \dots, \underline{e}_N)$  eigenvector matrix of the covariance matrix (of the data)  $\underline{C}$ , i.e.,  $\underline{C} \underline{e}_i = \lambda_i \underline{e}_i$ .

$\underline{C}$  is given by  $C_{ij} = \frac{1}{P} \sum_{\alpha=1}^P \underline{x}_{\text{centered},i}^{(\alpha)} \underline{x}_{\text{centered},j}^{(\alpha)}$  (i.e.  $\underline{C} = \frac{1}{P} \underline{X}_c \underline{X}_c^T$ )

$$\Rightarrow \frac{1}{P} \sum_{\alpha=1}^P \underline{x}_{\text{sphered}}^{(\alpha)} = \underline{0} \quad \text{mean (zero) centered}$$

for  $\underline{x}_c = \begin{pmatrix} \underline{x}_{\text{centered}}^{(1)} \\ \vdots \\ \underline{x}_{\text{centered}}^{(P)} \end{pmatrix}$

and  $\frac{1}{P} \sum_{\alpha=1}^P \underline{x}_{\text{sphered}}^{(\alpha)T} \underline{x}_{\text{sphered}}^{(\alpha)} = 1$  variance (unit std. dev.)

and  $\frac{1}{P} \sum_{\alpha=1}^P \underline{x}_{\text{sphered},i}^{(\alpha)} \underline{x}_{\text{sphered},j}^{(\alpha)} = 0$  correlation  $\downarrow$  (uncorrelated) variables

c) Analytical solution for linear neuron with basis functions, quadratic cost, regularization (L2)

Data Set  $\underline{\phi}(\underline{x}^{(\alpha)}) := \begin{pmatrix} \phi_1(\underline{x}^{(\alpha)}) \\ \vdots \\ \phi_d(\underline{x}^{(\alpha)}) \end{pmatrix}, \alpha=1, \dots, P$  [+ labels  $y_T^{(\alpha)}$ ]

$\Rightarrow$  w/o regularization, i.e.  $\lambda=0$

$\Rightarrow$  From previous results:  $\underline{w}^* = \left( \underline{\Phi} \underline{\Phi}^T + \lambda \underline{I} \right)^{-1} \underline{\Phi}^T \underline{y}$

where  $\underline{\Phi} = (\underline{\phi}^{(1)}, \dots, \underline{\phi}^{(P)}) \in \mathbb{R}^{d \times P}$  |  $y(\underline{x}; \underline{w}) = \underline{w}^T \underline{\phi}(\underline{x}) = \sum_{i=1}^d w_i \phi_i(\underline{x})$