Welcome from the Department of Mathematics at the University of Southern Mississippi

# **RBF** Hermite Interpolation

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#### **Outline**

- ▶ RBF Interpolation
- ► RBF Hermite Interpolation
- ► Comparison of results
- ► Conclusion

### **Surface Reconstruction Scheme**

Assume that  $f(\mathbf{x}) \approx s(\mathbf{x})$ 

Data set  $\left\{\mathbf{x}_i\right\}_1^N$  of pairwise distinct centres with the imposed

conditions  $f(\mathbf{x}_i) = s(\mathbf{x}_i), \quad 1 \le i \le N.$ 

The linear system  $s(\mathbf{x}_i) = \sum_{j=1}^{N} a_j \varphi(||\mathbf{x}_i - \mathbf{x}_j||), \quad 1 \le i \le N,$ 

is well-posed if the interpolation matrix is non-singular

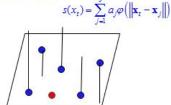
$$A_{\varphi} = \left[ \left. \varphi \left\| \mathbf{x}_i - \mathbf{x}_j \, \right\| \right. \right]_{1 \leq i,j \leq N}$$

$$s(x_i) = \sum_{j=1}^{5} a_j \varphi(\|\mathbf{x}_i - \mathbf{x}_j\|), \quad 1 \le i \le 5,$$

$$\begin{bmatrix} \varphi_{11} & \varphi_{12} & \varphi_{13} & \varphi_{14} & \varphi_{15} \\ \varphi_{21} & \varphi_{22} & \varphi_{23} & \varphi_{24} & \varphi_{25} \\ \varphi_{31} & \varphi_{32} & \varphi_{33} & \varphi_{34} & \varphi_{35} \\ \varphi_{41} & \varphi_{42} & \varphi_{43} & \varphi_{44} & \varphi_{45} \\ \varphi_{51} & \varphi_{52} & \varphi_{53} & \varphi_{54} & \varphi_{55} \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix} = \begin{bmatrix} f(x_1) \\ f(x_2) \\ f(x_3) \\ f(x_4) \\ f(x_5) \end{bmatrix}$$

# Interpolation Matrix

Note: 
$$\varphi_{ij} = \varphi(\|\mathbf{x}_i - \mathbf{x}_j\|)$$



- ► This scheme use the data and the derivative informations of the data.
- ▶ Let us consider the data  $(x_i, \lambda_i f)$ , i = 1, 2, ...N,  $x_i \in \mathbb{R}^s$ .
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► The generalized Hermite interpolant is of the form

$$P_{f}(x) = \sum_{i=1}^{N} c_{j} \lambda_{j}^{\xi} \phi(\|x - \xi\|), \ x \in \mathbb{R}^{s}.$$
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▶ The vector  $f_{\lambda}$  is of the form:

$$f_{\lambda} = [\lambda_1 f, ..., \lambda_N f]^T \tag{5}$$

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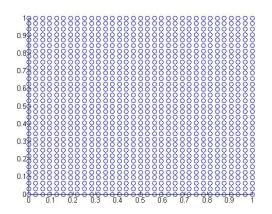
$$RMSE = \sqrt{\frac{1}{n_t} \sum_{i=1}^{n_t} (\hat{u}_i - u_i)^2},$$
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where  $n_t$  is the number of test points in the domain and  $\hat{u}_j$  and  $u_j$  are the approximate and exact solution of the partial differential equation at the  $j^{th}$  computational point respectively.

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