

Data Science for People in a Hurry

Scattered Data Interpolation

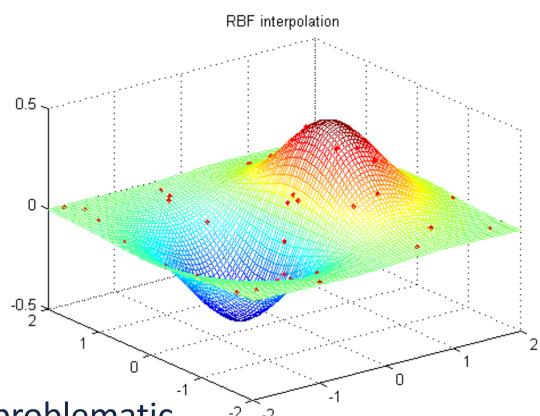
Scientific Visualization Professor Eric Shaffer



Scattered Data

Scattered data is irregularly sampled

No spatial structure

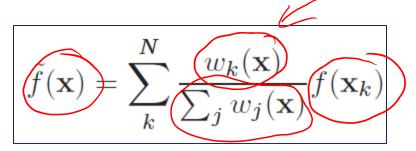


Using bilinear or trilinear interpolation problematic



Shepard's Method

Simplest scattered data interpolation method

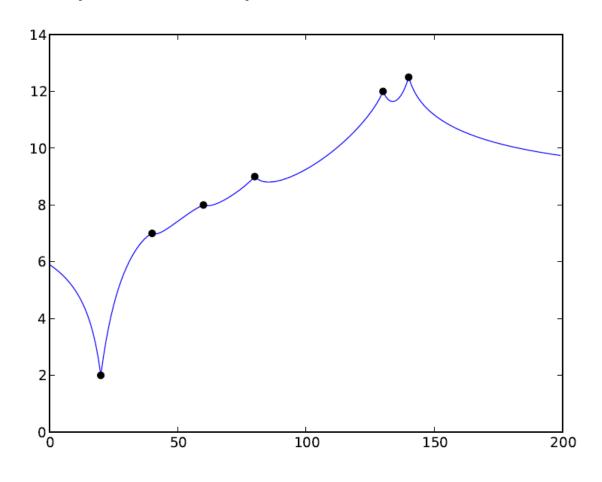


- x_k are the locations in space with known function values
- x is the query point
- w is a weight function inversely dependent on distance to x
- p is a positive real number
 - larger p is, the greater influence points close to x will have



Shepard's Method Issues

For p ≤ 1 the interpolant has peaks...not ideal for smooth interpolation



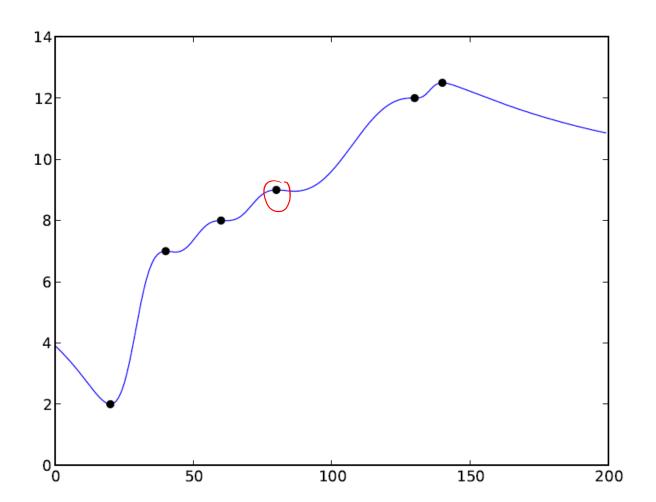
Example with p=1



Shepard's Method Issues

For p > 1 the interpolant is smooth...

But, first derivative is equal to 0 at data points...again not usually a desired behavior



Example with p=2



Modified Shepard's Method

One other issue is lack of scalability...ALL points in a data set used at each query x

Modified Shepard's Method uses only points within a radius of r around x

For those points, the weight function is
$$w_j(\mathbf{x}) = \left[\frac{r - d(\mathbf{x}, \mathbf{x}_i)}{r d(\mathbf{x}, \mathbf{x}_i)}\right]^2$$

Requires use of a spatial data structure such as kd-tree or quadtree/octree



Radial Basis Functions

- Any function dependent on distance from a center is radial
- We can compute an interpolating function as a weighted sum...

r- 11x-P11

$$\phi(x,p) = \phi(||x-p||)$$

$$f(x) \approx \sum_{i=1}^{N} w_i \phi(x, p_i)$$

• Some popular kernel functions

$$\phi(r) = e^{-\lambda r^2}$$
 Gaussian

$$\phi(r) = \frac{1}{1+r^2}$$
 Inverse distance



RBFs Computing Weights

- Need to compute weights
- Constraint is function interpolates data points

- For scalability use a kernel function with width
 - Can lead to non-smooth interpolant

$$f(p_j) = \sum_{i=1}^{N} w_i \phi(p_j, p_i)$$

$$Aw = p$$

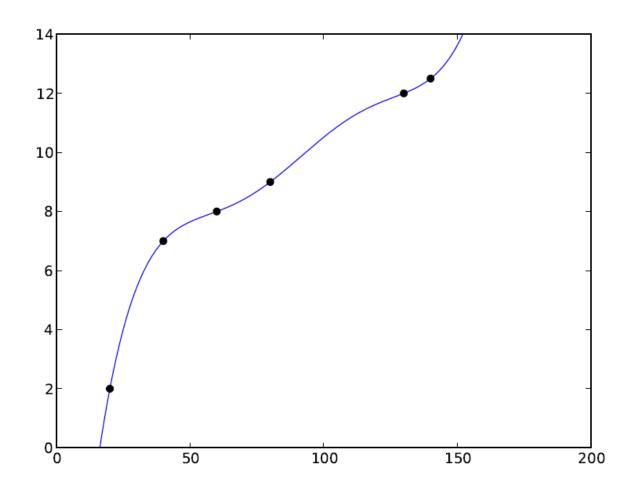
$$A = \begin{bmatrix} \phi(p_{1}, p_{1}) & \dots & \phi(p_{1}, p_{N}) \\ \dots & \dots & \dots \\ \phi(p_{N}, p_{1}) & \dots & \phi(p_{N}, p_{N}) \end{bmatrix}$$

$$w = \begin{bmatrix} w_1 \\ \dots \\ w_N \end{bmatrix}$$

$$p = \begin{bmatrix} f(p_1) \\ \dots \\ f(p_N) \end{bmatrix}$$



RBF Interpolation Example



Example radial basis interpolation with a Gaussian kernel



Kernel Function Choice

Positive definite functions will result in non-singular A for any data points

One definition of positive definite function: matrix A has all positive eigenvalues

Some useful kernels are not positive definite



Other Interpolation Method Options

- Moving Least Squares
- Natural Neighbor Interpolation
- ...many more

