

Notes

- ◆ Added required reading to web (numerical disasters)

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Interpolation in 1D

- ◆ Linear interpolation
- ◆ Polynomial interpolation
 - Lagrange formula
 - Dangers
- ◆ Splines and more
- ◆ Can be extended to 2d etc with tensor products

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Function Space Interpretation

- ◆ Rather than think of it as a discrete problem:
“How do I estimate function at a given point given data at nearby points”
think in terms of function space:
“Which function from my chosen space fits the data?”
- ◆ This perspective will come up again and again

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1D Function Spaces

- ◆ Piecewise-linear interpolation
- ◆ Polynomial interpolation
- ◆ Splines
- ◆ Choice of basis

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Scattered Data Interpolation

- ◆ Look at something more interesting
- ◆ Arbitrarily scattered data points in multiple dimensions
- ◆ How do we fit a smooth surface through them?
- ◆ Triangulation
- ◆ Radial Basis Functions
- ◆ Moving Least-Squares

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Triangulation

- ◆ First construct a mesh through the data points
- ◆ Then interpolate on each triangle separately
- ◆ In 2D this works fine
- ◆ In 3D (tetrahedralization) it gets tricky
- ◆ Doesn't scale well past 3D
- ◆ We'll talk about mesh generation later

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Radial Basis Functions (RBF)

- ◆ Assume data is in the span of a set of radial basis functions
- ◆ Solve for the right coefficients

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RBF + polynomials

- ◆ Problem: RBF's can't even get constants correct
- ◆ Fix by adding a low order polynomial term
- ◆ Under-constrained system!
- ◆ What extra constraints should we add?

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PDE Interpretation

- ◆ Think of the problem physically: what problem are we really solving? (another perspective which will come up again and again)
- ◆ We want the smoothest surface which goes through the given points
- ◆ Define smoothest, then solve the problem exactly (gives a differential equation...)

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First Try in 1D

- ◆ Try to minimize a measure of how 'wobbly' the function is:

$$\min_{f(x_i)=f_i} \int_{-\infty}^{\infty} \frac{1}{2} (f'(x))^2 dx$$

- ◆ Calculus of variations gives:

$$f''(x) = 0 \quad \text{in } (x_{i-1}, x_i)$$

$$f(x_i) = f_i$$

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Exact Solution

- ◆ Direct method: differential equation has piecewise linear solution
 - But this is too hard to do directly in more dimensions...
- ◆ Method of fundamental solution:
 - First find a function that satisfies
$$\phi''(x) = \delta(x)$$
 - Simplest (symmetric) answer is:

$$\phi(x) = \frac{1}{2}|x|$$

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Fundamental Solution...

- ◆ Then we know the exact solution is of the form

$$f(x) = A + \sum_{i=1}^n \lambda_i \phi(x - x_i)$$

$$= A + \sum_{i=1}^n \frac{1}{2} \lambda_i |x - x_i|$$

- ◆ The constant A doesn't affect f'
- ◆ Also want f' → 0 at infinity
- ◆ Away from all data points, f' is proportional to $\sum_{i=1}^n \lambda_i$

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Simple 1D RBF

- ◆ Putting this together, our equations are then:

$$A + \sum_{i=1}^n \lambda_i \phi(x_j - x_i) = f_i \quad j = 1, \dots, n$$

$$\sum_{i=1}^n \lambda_i = 0$$

- ◆ n+1 linear equations for n+1 unknowns

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Smoother Interpolation

- ◆ Minimize curvature

$$\min_{f(x_i)=f_i} \int_{-\infty}^{\infty} \frac{1}{2} (f''(x))^2 dx$$

- ◆ Calculus of variations gives us:

$$f^{(iv)}(x) = 0 \quad \text{in } (x_{i-1}, x_i)$$

$$f(x_i) = f_i$$

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Smoother Fundamental Solution

- ◆ Our basis function is now

$$\phi(x) = \frac{1}{12} |x|^3$$

and we also include a linear term in polynomial (doesn't change f'')

- ◆ So our solution is of the form:

$$f(x) = A + Bx + \sum_{i=1}^n \lambda_i \phi(x - x_i)$$

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Boundedness

- ◆ Outside of the data points, f'' is:

$$\begin{aligned} f''(x) &= \pm \sum \frac{1}{2} \lambda_i (x - x_i) \\ &= \pm \frac{1}{2} \left(\sum \lambda_i \right) x \mp \frac{1}{2} \left(\sum \lambda_i x_i \right) \end{aligned}$$

- ◆ Setting this to zero gives two conditions on the coefficients...
- ◆ n+2 equations for n+2 unknowns

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More Dimensions

- ◆ Same approach generalizes
- ◆ E.g. "thin-plate spline" comes out of

$$\min_{f(x_i)=f_i} \int \frac{1}{2} (\nabla \cdot \nabla f)^2$$

- ◆ The Laplacian $\nabla \cdot \nabla$ is a measure of mean curvature

- ◆ Calculus of variations gives the "biharmonic equation":

$$\nabla \cdot \nabla \nabla \cdot \nabla f = 0$$

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Thin-Plate Spline

- ◆ In 2D, fundamental solution is proportional to

$$\phi(x) = |x|^2 \log|x|$$

and in 3D: $\phi(x) = |x|$

- ◆ Include a linear polynomial
- ◆ Boundedness conditions are:

$$\sum \lambda_i = 0, \quad \sum \lambda_i x_i = 0, \quad \sum \lambda_i y_i = 0, \quad \sum \lambda_i z_i = 0$$

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Other RBF's

- ◆ Other basis functions can be used of course
- ◆ Usual alternatives:
 - Triharmonic basis function: $\phi(x) = |x|^3$
 - Multiquadric: $\phi(x) = \sqrt{|x|^2 + c^2}$
 - Gaussian: $\phi(x) = \exp\left(-\frac{x^2}{c^2}\right)$
- ◆ With or without low order polynomial

The Equations

- ◆ In all cases, we end up with a linear system to solve
- ◆ How do we solve it?
- ◆ Gaussian Elimination is the usual answer