

Rational Function Interpolation  
Barycentric Rational Interpolation  
Coefficients of Interpolating Polynomial

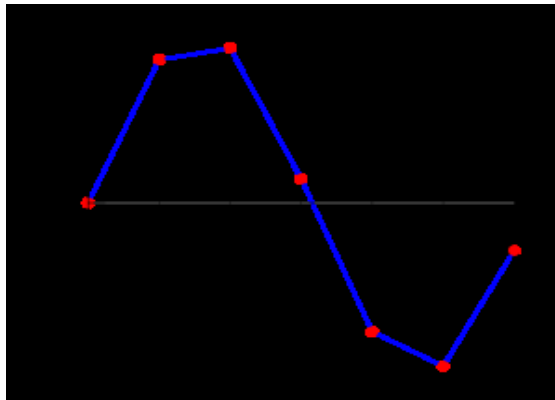
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# Summary of Previous Interpolation Methods

Linear Interpolation  
Polynomial Interpolation  
Cubic Spline Interpolation

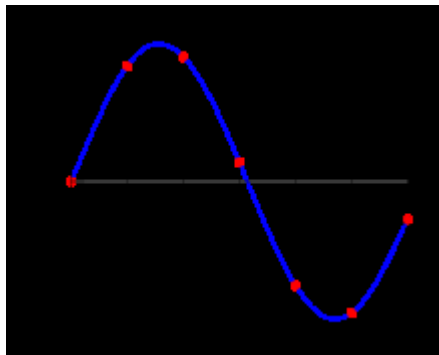
# Linear Interpolation

- Concatenation of linear interpolants between a pair of data points
- Piecewise linear function
- Cheap . . .  $O(N \log(N))$
- Connect data points in a table (e.g. given population in 1990 and 2000, what was population in 1995?)
- Historically used with astronomical data



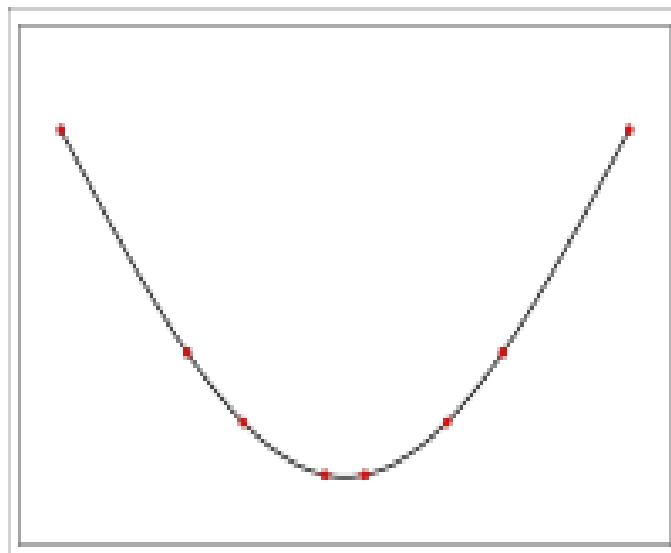
# Polynomial Interpolation

- Given a data set, find a polynomial that goes exactly through each point
- Neville's algorithm:  $O(N^2)$
- Basis for algorithms solving numerical ODEs and numerical integration
- Unstable on equidistant grid



# Cubic Spline

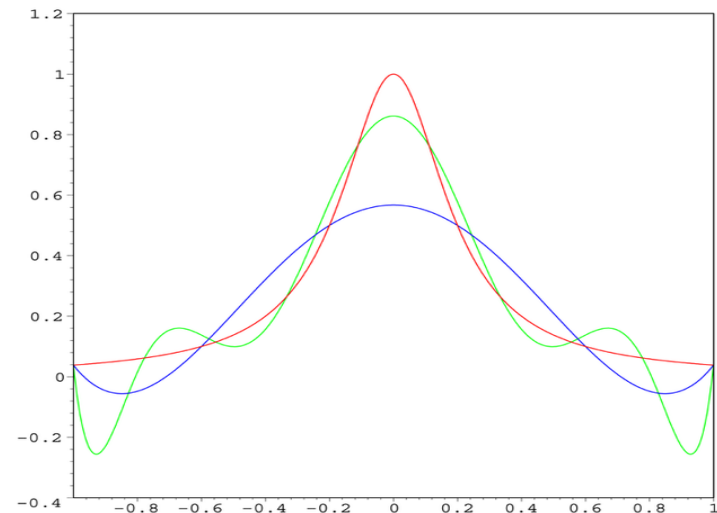
- Interpolation interval divided into subintervals and each subinterval interpolated using 3rd degree polynomial
- Piecewise cubic functions with continuous first and second derivatives
- Requires function continuity and passing through all data points
- $O(N)$  complexity . . . stable and simplistic calculation



# Cubic Spline Continued

- Known boundary first derivatives . . .  $O(h^4)$
- Natural spline (2nd derivatives = 0) . . .  $O(h^2)$
- Preferable to polynomials because the interpolation error can be made small even when using low degree polynomials
- Avoid Runge's Phenomenon in which boundaries oscillate wildly for functions like

$$f(x) = \frac{1}{1 + 25x^2}$$



# Why Rational Function Interpolation?

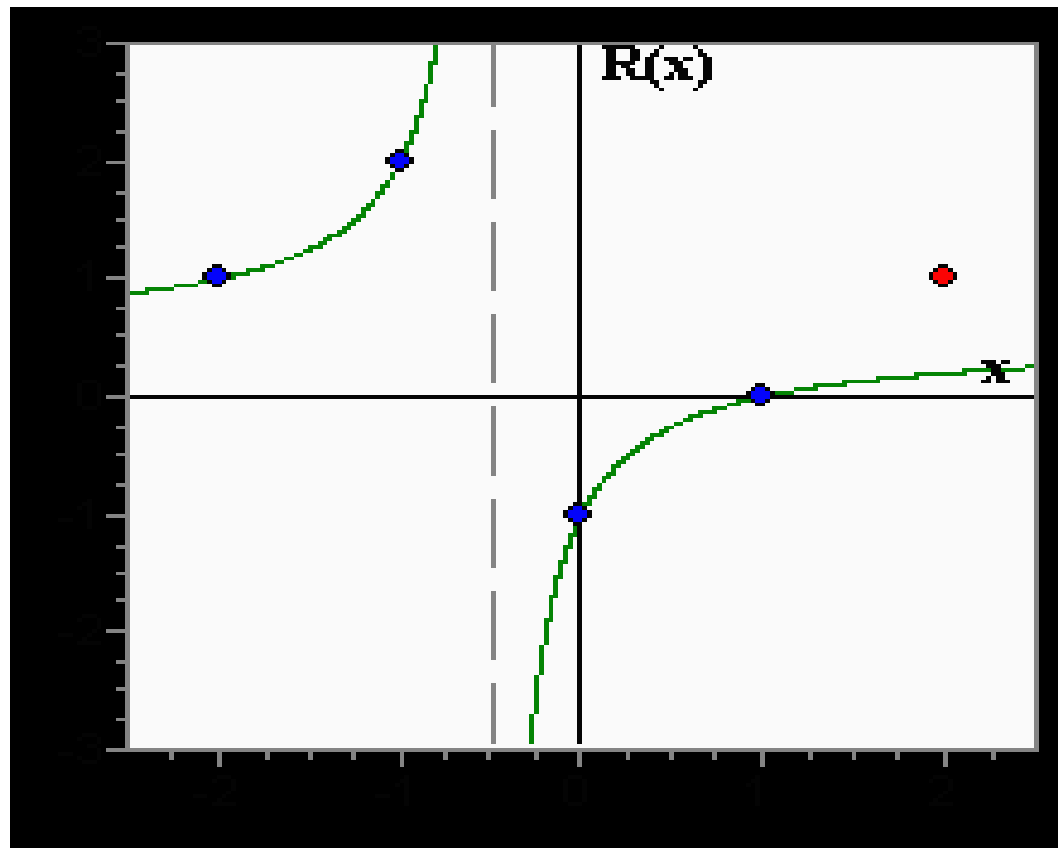
- Express more diverse behavior than polynomials
- Solves disadvantages of polynomial interpolation but polynomial can be found at any point; rational function cannot
- Ability to model equations with poles (while polynomials fail) , . . . Good if modeling a function with poles. Bad if goal is numerical stability.
- Higher orders give higher accuracy

# Example of Unstable Rational Interpolation

Numerator and Denominator of 2nd degree

Pole at  $x = -0.5$

x	y
-2	1
-1	2
0	-1
1	0
2	1





# Rational Functions Continued

- Historically, the rational interpolant was constructed by solving a set of equations with unknown coefficients.
  - However, the larger the data set the larger the error in calculating coefficients
- Neville's algorithm solves this problem by setting degree of numerator and denominator equal to  $N/2$
- Main disadvantage: no mechanism to find poles

# Why Barycentric Rational Interpolation?

- Suppresses all nearby poles
- Experimentation with higher orders encouraged
- Favorable comparison to splines but with smaller error and infinitely smooth curves
- If spacing of points is  $O(h)$ , error is  $O(h^{d+1})$  as  $h \rightarrow 0$ . The complexity is order  $O(Nd)$

# Runge's example with barycentric rational interpolation

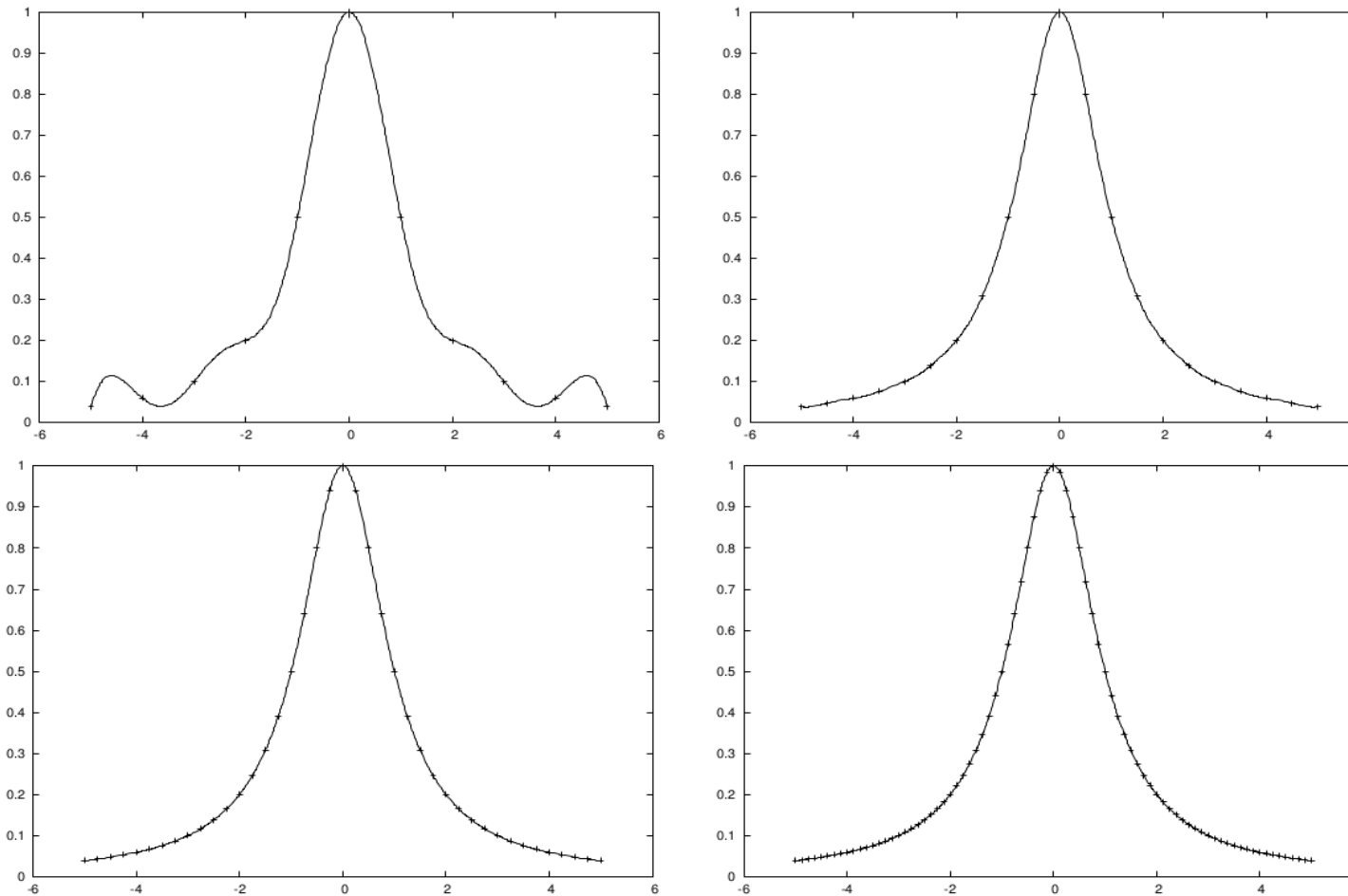


Figure 1: Interpolating Runge's example with  $d = 3$  and  $n = 10, 20, 40, 80$ .

# Sin(x) and errors for Sin(x) and Runge's Eq

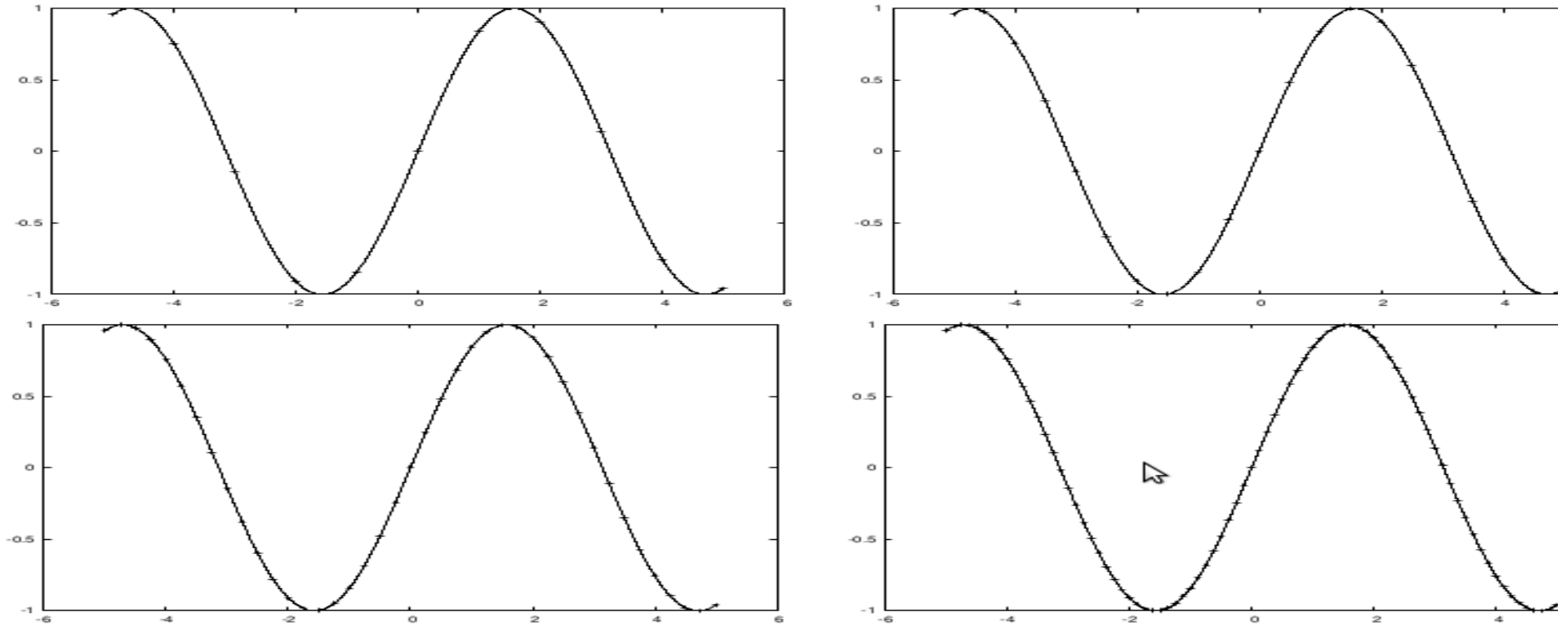


Figure 2: Interpolating the sine function with  $d = 4$  and  $n = 10, 20, 40, 80$ .

$n$	Runge, $d = 3$	order	sine, $d = 4$	order
10	6.9e-02		1.7e-02	
20	2.8e-03	4.6	3.9e-04	5.5
40	4.3e-06	9.4	7.1e-06	5.8
80	5.1e-08	6.4	1.3e-07	5.7
160	3.0e-09	4.1	2.7e-09	5.6
320	1.8e-10	4.0	6.0e-11	5.5
640	1.1e-11	4.0	1.5e-12	5.3

Table 1: Error in rational interpolant.

# Optimal Order for $y = \text{Abs}(x)$

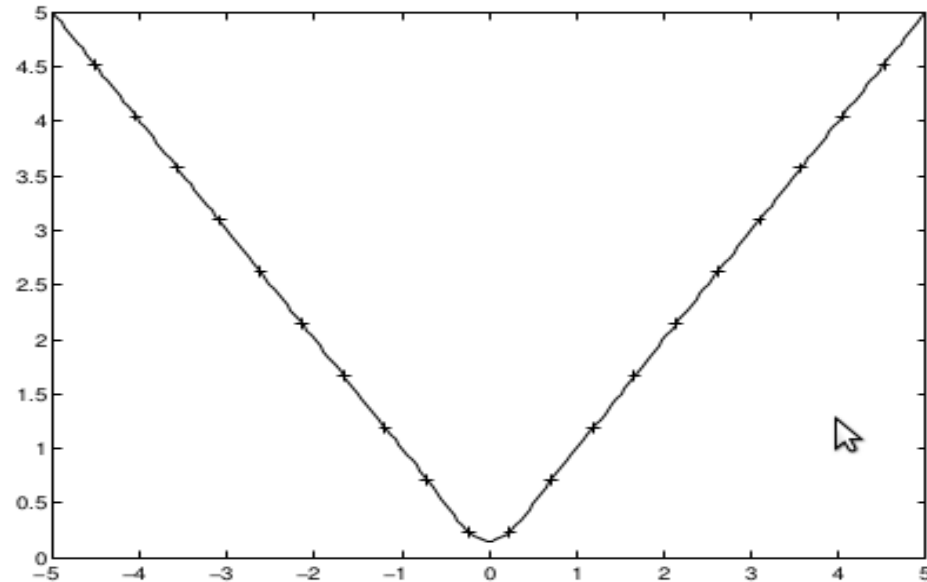


Figure 3: Interpolating  $|x|$  over  $[-5, 5]$  with  $d = 3$  and  $n = 21$ .

$n$	best $d$ value	error
10	$d = 0$	$3.6\text{e-}02$
20	$d = 1$	$1.5\text{e-}03$
40	$d = 3$	$4.3\text{e-}06$
80	$d = 7$	$2.0\text{e-}10$
160	$d = 10$	$1.3\text{e-}15$

Table 2: Error in Runge's example, varying  $d$ .

# Errors in barycentric vs cubic spline interpolation for Runge's equation and the sine function

$n$	rational, $d = 3$	cubic spline	$n$	rational, $d = 3$	cubic spline
10	6.9e-02	2.2e-02	10	1.3e-02	3.3e-03
20	2.8e-03	3.2e-03	20	1.2e-03	1.7e-04
40	4.3e-06	2.8e-04	40	8.4e-05	1.0e-05
80	5.1e-08	1.6e-05	80	5.4e-06	6.4e-07
160	3.0e-09	9.5e-07	160	3.4e-07	4.0e-08
320	1.8e-10	5.9e-08	320	2.1e-08	2.5e-09
640	1.1e-11	3.7e-09	640	1.3e-09	1.6e-10

Table 3: Error in rational and spline interpolation of Runge's (left) and the sine function (right)

# Rational Function Algorithm

$$R_{i(i+1)\dots(i+m)} = \frac{P_\mu(x)}{Q_\nu(x)} = \frac{p_0 + p_1x + \dots + p_\mu x^\mu}{q_0 + q_1x + \dots + q_\nu x^\nu} \quad (3.4.1)$$

Since there are  $\mu + \nu + 1$  unknown  $p$ 's and  $q$ 's ( $q_0$  being arbitrary), we must have

$$m + 1 = \mu + \nu + 1 \quad (3.4.2)$$

# Recurrence Relations

Polynomial Approximations

$$P_{i(i+1)\dots(i+m)} = \frac{(x - x_{i+m})P_{i(i+1)\dots(i+m-1)} + (x_i - x)P_{(i+1)(i+2)\dots(i+m)}}{x_i - x_{i+m}} \quad (3.2.3)$$

Rational Function Approximations

$$R_{i(i+1)\dots(i+m)} = R_{(i+1)\dots(i+m)} + \frac{R_{(i+1)\dots(i+m)} - R_{i\dots(i+m-1)}}{\left(\frac{x-x_i}{x-x_{i+m}}\right) \left(1 - \frac{R_{(i+1)\dots(i+m)} - R_{i\dots(i+m-1)}}{R_{(i+1)\dots(i+m)} - R_{(i+1)\dots(i+m-1)}}\right) - 1} \quad (3.4.3)$$



# Barycentric Algorithm

Barycentric form  
of rational  
interpolant

$$R(x) = \frac{\sum_{i=0}^{N-1} \frac{w_i}{x - x_i} y_i}{\sum_{i=0}^{N-1} \frac{w_i}{x - x_i}} \quad (3.4.9)$$

$$w_k = \sum_{\substack{i=k-d \\ 0 \leq i < N-d}}^k (-1)^k \prod_{\substack{j=i \\ j \neq k}}^{i+d} \frac{1}{x_k - x_j} \quad (3.4.10)$$

Formula for the  
weights

For example,

$$w_k = (-1)^k, \quad d = 0$$

$$w_k = (-1)^{k-1} \left[ \frac{1}{x_k - x_{k-1}} + \frac{1}{x_{k+1} - x_k} \right], \quad d = 1 \quad (3.4.11)$$

# Coefficients of Interpolating Polynomials and Vandermonde Matrix

$$y = c_0 + c_1x + c_2x^2 + \cdots + c_{N-1}x^{N-1} \quad (3.5.1)$$

The  $c_i$ 's are required to satisfy the linear equation

$$\begin{bmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^{N-1} \\ 1 & x_1 & x_1^2 & \cdots & x_1^{N-1} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & x_{N-1} & x_{N-1}^2 & \cdots & x_{N-1}^{N-1} \end{bmatrix} \cdot \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_{N-1} \end{bmatrix} = \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_{N-1} \end{bmatrix} \quad (3.5.2)$$

- Transforms coefficients of a polynomial to the actual values it takes at particular points.
- Vandermonde determinant is non-vanishing for these points proving that the mapping is a one-to-one correspondence between coefficients and values . . . i.e. coefficients in polynomial interpolation have a unique solution

# Coefficients continued

- Problems

- Ill-conditioned as  $N$  increases . . . so technique only practical for small data sets
- If coefficients are used to interpolate functions, the interpolation will not pass through data points
- First algorithm in NR3 has  $O(N^2)$ ; second has  $O(N^3)$
- For high degrees of interpolation, precision of coefficients is essential . . . so interpolation error compounded by inaccuracy of coefficients

# Citations

<http://www.alglib.net/interpolation/rational.php>

<http://www.alglib.net/interpolation/spline3.php>

[http://en.wikipedia.org/wiki/Linear\\_interpolation](http://en.wikipedia.org/wiki/Linear_interpolation)

[http://en.wikipedia.org/wiki/Spline\\_interpolation](http://en.wikipedia.org/wiki/Spline_interpolation)

[http://en.wikipedia.org/wiki/Polynomial\\_interpolation](http://en.wikipedia.org/wiki/Polynomial_interpolation)