# How to put a polynomial through points

#### Ed Bueler

MATH 310 Numerical Analysis

September 2012

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These notes are an online replacement for the 14 September class of Math 310 (Fall 2012), while Bueler is away.

The topics here are also covered in Chapter 8 of the text (Greenbaum & Chartier). The emphasis here is on **how** to put a polynomial through points. When we get to Chapter 8 we will address the "how good" question.

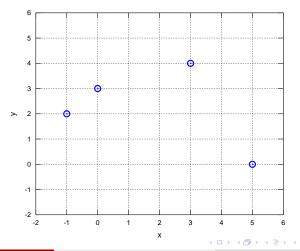
• suppose you have a function y = f(x) which goes through these points:

(-1,2), (0,3), (3,4), (5,0)

- the *x*-coordinates of these points are not equally-spaced!
  - in these notes I will never assume the x-coordinates are equally-spaced
- let us name these points  $(x_i, y_i)$ , for i = 1, 2, 3, 4
- there is a polynomial P(x) of degree 3 which goes through these points
- we will build it concretely
- we will show later that there is only one such polynomial

#### a picture of the problem

- figure below shows the points
- as stated, we suppose that they are values of a function f(x)
- but we don't see that function



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## how to find P(x)

*P*(*x*) is the degree 3 polynomial through the 4 points
a standard way to write it is:

$$P(x) = c_0 + c_1 x + c_2 x^2 + c_3 x^3$$

• note: there are 4 unknown coefficients and 4 points

- degree n 1 polynomials have the right length for n points
- the facts "P(x) = y" for the given points gives 4 equations:

$$\begin{split} c_0 + c_1(-1) + c_2(-1)^2 + c_3(-1)^3 &= 2\\ c_0 + c_1(0) + c_2(0)^2 + c_3(0)^3 &= 3\\ c_0 + c_1(3) + c_2(3)^2 + c_3(3)^3 &= 4\\ c_0 + c_1(5) + c_2(5)^2 + c_3(5)^3 &= 0 \end{split}$$

 MAKE SURE that you are clear on how I got these equations, and that you can do the same thing in an example with different points or different polynomial degree

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#### a linear system

- you can solve the equations by hand ... that would be tedious
- we want to automate the process
- we have a great matrix-vector tool, namely MATLAB
- I recognize the system has a matrix form "Av = b":

$$\begin{bmatrix} 1 & -1 & (-1)^2 & (-1)^3 \\ 1 & 0 & 0^2 & 0^3 \\ 1 & 3 & 3^2 & 3^3 \\ 1 & 5 & 5^2 & 5^3 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \\ 4 \\ 0 \end{bmatrix}$$

- (a known square matrix A) × (an unknown vector v) = (a known vector b)
- I am not simplifying the numbers in the matrix ... because:
  - o a machine can do that, and
  - the pattern in the matrix entries is clear if they are unsimplified
- MAKE SURE you can convert from the original "fit a polynomial through these points" question into the matrix form "Av = b"

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## how to *easily* find P(x)

- MATLAB is designed to solve linear systems ... easily!
- enter the matrix and the known vector into MATLAB:

• solve the linear system to get  $\mathbf{v} = [c_0 c_1 c_2 c_3]$ :

```
>> v = A \ b
v =
3.000000
0.983333
-0.066667
-0.050000
```

#### • so the polynomial is $P(x) = 3 + 0.983333x - 0.066667x^2 - 0.05x^3$

#### notes on matrices and vectors in MATLAB

- you enter matrices like A by rows
  - spaces separate entries
  - semicolons separate rows
- column vectors like b are just matrices with one column
  - o to quickly enter column vectors use the transpose operation:

```
>> b = [2 3 4 0]'
b =
2
3
4
0
```

- to solve the system  $A\mathbf{v} = \mathbf{b}$  we "divide by" the matrix:  $\mathbf{v} = A^{-1}\mathbf{b}$
- ... but this is *left* division, so MATLAB makes it into a single-character operation, the *backslash* operation:

>> v = A  $\setminus$  b

the forward slash does not work because of the sizes of the matrix and the vector are not right:

>> v = b / A % NOT CORRECT for our A and b; wrong sizes

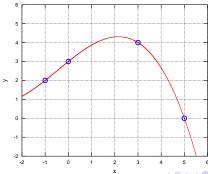
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#### did we solve the problem?

#### • the polynomial we found had better go through the points:

```
>> 3.000000 + 0.983333*(-1) - 0.066667*(-1)^2 -0.050000*(-1)^3
ans = 2
>> 3.000000 + 0.983333*(0) - 0.066667*(0)^2 -0.050000*(0)^3
ans = 3
>> 3.000000 + 0.983333*(3) - 0.066667*(3)^2 -0.050000*(3)^3
ans = 4.0000
>> 3.000000 + 0.983333*(5) - 0.066667*(5)^2 -0.050000*(5)^3
ans = -1.0000e-05
```

• a graph is convincing, too:



#### the general case

• suppose we have *n* points (*x<sub>i</sub>*, *y<sub>i</sub>*) with distinct *x*-coordinates

• for example, if n = 4 we have points  $(x_1, y_1), (x_2, y_2), (x_3, y_3), (x_4, y_4)$ 

- then the polynomial has degree one less: the polynomial P(x) which goes through the *n* points has degree n 1
- the polynomial has this form:

$$P(x) = c_0 + c_1 x + c_2 x^2 + \dots + c_{n-1} x^{n-1}$$

• the equations which determine *P*(*x*) say that *the polynomial goes through the points*:

$$P(x_i) = y_i$$
 for  $i = 1, 2, ..., n$ 

written out there are n equations of this form:

$$c_0 + c_1 x_i + c_2 x_i^2 + \dots + c_{n-1} x_i^{n-1} = y_i$$
 for  $i = 1, 2, \dots, n$ 

• the *n* coefficients *c<sub>i</sub>* are unknown, while the *x<sub>i</sub>* and *y<sub>i</sub>* are known

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#### the pattern in the matrix, for the general case

as a matrix:

$$A = \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & & \ddots & \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{bmatrix}$$

- and **b** is a column vector with entries  $y_i$ : **b** =  $[y_1 \ y_2 \ \dots \ y_n]'$
- as before, this gives a system of *n* equations,  $A\mathbf{v} = \mathbf{b}$
- the matrix A is called a Vandermonde matrix, from about 1772

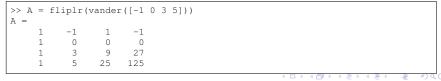
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### Vandermonde matrix, built-in

- actually, Vandermonde matrices are already built-in to MATLAB
- for example, the Vandermonde matrix A for our original four points (-1,2), (0,3), (3,4), (5,0) is

	3 5])	([-1 0	>> vander([-1	
			ans =	
1	-1	1	-1	
1	0	0	0	
1	3	9	27	
1	5	25	125	
	0	2	0 27	

- two comments:
  - o oops! the columns are in reversed order, compared to our choice
  - note that only the x-coordinates are needed to build A, and not the y-coordinates
- we easily fix the column order to agree with our earlier ordering using "fliplr", which stands for "flip left-to-right":



#### Vandermonde matrix method for polynomial interpolation

thus a complete code to solve our 4 point problem earlier is:

```
A = fliplr(vander([-1 0 3 5]));
b = [2 3 4 0]';
v = A \ b
```

• after the coefficients v are computed, they form P(x) this way:

$$P(x) = v(1) + v(2) x + v(3) x^{2} + \dots + v(n) x^{n-1}$$

• thus we can plot the 4 points and the polynomial this way:

```
plot([-1 0 3 5],[2 3 4 0],'o','markersize',12)
x = -2:0.01:6; P = v(1) + v(2) *x + v(3) *x.^2 + v(4) *x.^3;
hold on, plot(x,P,'r'), hold off
xlabel x, ylabel y
```

• this was the "convincing" graph, shown earlier

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#### on the cost of solving Vandermonde matrix problems

- now, often we want to do a polynomial fit problem like this many times for different data
- so, is it quick? here are some facts to know about solving these systems:
  - if there are *n* points then the matrix *A* has *n* rows and *n* columns
  - *internally in* MATLAB, the linear system  $A\mathbf{v} = \mathbf{b}$  is solved by Gaussian elimination
  - Gaussian elimination does about  $\frac{2}{3}n^3$  arithmetic operations (i.e. additions, subtractions, multiplications, divisions) to solve such a linear system
- so finding the coefficients of the polynomial P(x) through n points takes about n<sup>3</sup> operations
- but then you need more operations to *evaluate* that polynomial, which is what you usually do with it

- before Vandermonde there was already a good, practical idea
   o an old idea of Newton, perhaps about 1690
- the idea is to write the polynomial through the data P(x) not using the "monomials"  $1, x, x^2, x^3, \ldots, x^{n-1}$ ,
- ... but instead to use a form of the polynomial which includes the *x*-coordinates of the data points:

$$P(x) = c_0 + c_1(x - x_1) + c_2(x - x_1)(x - x_2) + \cdots + c_{n-1}(x - x_1)(x - x_2) \dots (x - x_{n-1})$$

• do you see why this helps?

#### Newton's form example: 4 points

● with the n = 4 points (-1,2), (0,3), (3,4), (5,0) we can write

$$P(x) = c_0 + c_1(x+1) + c_2(x+1)(x) + c_3(x+1)(x)(x-3)$$

• this polynomial must go through the four points, so:

$$c_0 = 2$$

$$c_0 + c_1(0+1) = 3$$

$$c_0 + c_1(3+1) + c_2(3+1)(3) = 4$$

$$c_0 + c_1(5+1) + c_2(5+1)(5) + c_3(5+1)(5)(5-3) = 0$$

- note that lots of terms are zero!
- the system of equations has the form

$$M\mathbf{w} = \mathbf{b}$$

where M is a triangular matrix, **b** is the same as in the Vandermonde form, and **w** has the unknown coefficients:

$$\mathbf{w} = [c_0 \ c_1 \ c_2 \ c_3]'$$

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- can you solve this by hand?
- yes: find  $c_0$  from first equation, then  $c_1$  from second equation, etc.
- I get  $c_0 = 2, c_1 = 1, c_2 = -1/6, c_3 = -1/20$ , so

$$P(x) = 2 + (x+1) - \frac{1}{6}(x+1)(x) - \frac{1}{20}(x+1)(x)(x-3)$$

MAKE SURE you can do this yourself, on a similar example

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• so we have a concrete polynomial, but not in standard form:

$$P(x) = 2 + (x+1) - \frac{1}{6}(x+1)(x) - \frac{1}{20}(x+1)(x)(x-3)$$

• an uninteresting calculation puts it in standard form:

$$P(x) = 3 + \frac{59}{60}x - \frac{1}{15}x^2 - \frac{1}{20}x^3$$
  
= 3 + 0.983333x - 0.066667x^2 - 0.05x^3

which is exactly the same polynomial we found earlier

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### Newton's form for polynomial interpolation: example code

• the advantage of the Newton form is that a triangular matrix M is created

- o which makes it easier to solve the system by hand
- only  $O(n^2)$  operations are needed to solve the system
- the polynomial comes out in a non-standard form but it is just as easy to evaluate at a point
- for now here is a short code to solve the 4 point problem:

```
newt4.m
% NEWT4 Compute P(x) using the Newton form, for 4 points.
n = 4; x = [-1 0 3 5]'; y = [2 3 4 0]'; % the points
M = zeros(n,n); % makes M the right size
% form M by columns
M(:,1) = ones(n,1);
for j=2:n
M(j:n,j) = M(j:n,j-1) .* (x(j:n) - x(j-1));
end
b = y;
w = M \ b % w has the coefficients of the polynomial:
% P(x) = w1 + w2 (x-x1) + w3 (x-x1) (x-x2) + w4 (x-x1) (x-x2) (x-x3)
```

### Newton's form shows there is a unique interpolating polynomial

- for both Vandermonde and Newton matrix approximations we build an invertible matrix, so in each case there is exactly one solution
- this is easiest to see from the general Newton form matrix:

$$M = \begin{bmatrix} 1 & (x_2 - x_1) \\ 1 & (x_3 - x_1) & (x_3 - x_1)(x_3 - x_2) \\ \vdots & \vdots & \vdots & \ddots \\ 1 & (x_n - x_1) & (x_n - x_1)(x_n - x_2) & \dots & (x_n - x_1)(x_n - x_2) \dots (x_n - x_{n-1}) \end{bmatrix}$$

- the diagonal entries are all nonzero as long as the x-coordinates are distinct
- we can calculate the determinant of this Newton form matrix M
- because the matrix is triangular, the determinant is the product of the diagonal:

$$\det M = \prod_{i>j} (x_i - x_j) \neq 0$$

so the polynomial P(x) always exists and is unique

## Lagrange's idea: no systems at all!

- another new idea
- given the same n = 4 points (-1, 2), (0, 3), (3, 4), (5, 0)
- Lagrange and others, by about 1800, knew how to write down four polynomials, now called the *Lagrange polynomials*, corresponding to the *x*-coordinates *x*<sub>1</sub>,..., *x*<sub>4</sub>:

$$\ell_{1}(x) = \frac{(x - x_{2})(x - x_{3})(x - x_{4})}{(x_{1} - x_{2})(x_{1} - x_{3})(x_{1} - x_{4})} = \frac{x(x - 3)(x - 5)}{(-1)(-4)(-6)}$$
  

$$\ell_{2}(x) = \frac{(x - x_{1})(x - x_{3})(x - x_{4})}{(x_{2} - x_{1})(x_{2} - x_{3})(x_{2} - x_{4})} = \frac{(x + 1)(x - 3)(x - 5)}{(1)(-3)(-5)}$$
  

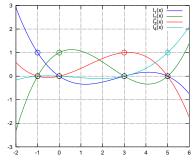
$$\ell_{3}(x) = \frac{(x - x_{1})(x - x_{2})(x - x_{4})}{(x_{3} - x_{1})(x_{3} - x_{2})(x_{3} - x_{4})} = \frac{(x + 1)(x)(x - 5)}{(4)(3)(-2)}$$
  

$$\ell_{4}(x) = \frac{(x - x_{1})(x - x_{2})(x - x_{3})}{(x_{4} - x_{1})(x_{4} - x_{2})(x_{4} - x_{3})} = \frac{(x + 1)(x)(x - 3)}{(6)(5)(2)}$$

• the *pattern* needs attention: **I.** the numerator and denominator have the same pattern, but the denominator is a constant with no variable "*x*"; **II.**  $\ell_i(x)$  has no " $(x - x_i)$ " factor in the numerator, nor " $(x_i - x_i)$ " factor in the denominator; **III.** as long as the  $x_i$  are distinct, we never divide by zero

## Lagrange's idea: polynomials which "hit one point"

- why is this helpful?
- consider a plot of  $\ell_1(x)$ ,  $\ell_2(x)$ ,  $\ell_3(x)$ ,  $\ell_4(x)$ :



• a crucial pattern emerges:

the polynomial  $\ell_i(x)$  has value 0 at all of the x-values of the points, except that it is 1 at  $x_i$ 

• MAKE SURE make sure you can find the Lagrange polynomials if I give you the *x*-values of *n* points

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#### Lagrange's idea, cont.

 the picture on the last page illustrates what is generally true of the Lagrange polynomials:

$$\ell_i(x_j) = \begin{cases} 1, & j = i, \\ 0, & \text{otherwise.} \end{cases}$$

- also, the Lagrange polynomials for the 4 points are each of degree 3
- so why does this help find P(x)?
- recall that we have values y<sub>i</sub> which we want the polynomial P(x) to "hit"
- that is, we want this to be true for each i:

$$P(x_i) = y_i$$

thus the answer is:

$$P(x) = y_1 \ell_1(x) + y_2 \ell_2(x) + y_3 \ell_3(x) + y_4 \ell_4(x)$$

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## Lagrange's idea, cont.<sup>2</sup>

• wait, why is this the answer?:

$$P(x) \stackrel{*}{=} y_1 \ell_1(x) + y_2 \ell_2(x) + y_3 \ell_3(x) + y_4 \ell_4(x)$$

because P(x) is of degree three, as a linear combination of degree 3 polynomials, and

because:

$$P(x_1) = y_1\ell_1(x_1) + y_2\ell_2(x_1) + y_3\ell_3(x_1) + y_4\ell_4(x_1)$$
  
=  $y_1 \cdot 1 + y_2 \cdot 0 + y_3 \cdot 0 + y_4 \cdot 0$   
=  $y_1$ ,

and

$$P(x_2) = y_1 \ell_1(x_2) + y_2 \ell_2(x_2) + y_3 \ell_3(x_2) + y_4 \ell_4(x_2)$$
  
=  $y_1 \cdot 0 + y_2 \cdot 1 + y_3 \cdot 0 + y_4 \cdot 0$   
=  $y_2$ ,

and so on

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## Lagrange's idea, cont.<sup>3</sup>

- on the last slide we saw that P(x<sub>i</sub>) = y<sub>i</sub> because the polynomials l<sub>i</sub>(x) help "pick out" the point x<sub>i</sub> in the general expression \* on the last slide
- we can say this more clearly using summation notation:
  - the polynomial is a sum of the Lagrange polynomials with coefficients y<sub>i</sub>:

$$P(x) = \sum_{i=1}^{4} y_i \ell_i(x)$$

 when we plug in one of the x-coordinates of the points, we get only one "surviving" term in the sum:

$$P(x_j) = \sum_{i=1}^{4} y_i \ell_i(x_j) = y_j \cdot 1 + \sum_{i \neq j} y_i \cdot 0 = y_j$$

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#### returning to our 4-point example

 for our 4 concrete points (-1,2), (0,3), (3,4), (5,0), we can slightly-simplify the Lagrange polynomials we have computed already:

$$\ell_1(x) = -\frac{1}{24}x(x-3)(x-5)$$
  

$$\ell_2(x) = +\frac{1}{15}(x+1)(x-3)(x-5)$$
  

$$\ell_3(x) = -\frac{1}{24}(x+1)(x)(x-5)$$
  

$$\ell_4(x) = +\frac{1}{60}(x+1)(x)(x-3)$$

so the polynomial which goes through our points is

$$P(x) = -(2)\frac{1}{24}x(x-3)(x-5) + (3)\frac{1}{15}(x+1)(x-3)(x-5)$$
$$-(4)\frac{1}{24}(x+1)(x)(x-5) + (0)\frac{1}{60}(x+1)(x)(x-3)$$

a tedious calculation simplifies this to

$$P(x) = 3 + \frac{59}{60}x - \frac{1}{15}x^2 - \frac{1}{20}x^3,$$

which is exactly what we found earlier

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### so, is the Lagrange scheme a good idea?

 for n points { (x<sub>i</sub>, y<sub>i</sub>) } we have the following nice formulas which "completely answer" the polynomial interpolation problem:

$$\ell_i(x) = \prod_{j \neq i} \frac{x - x_j}{x_i - x_j}$$

$$P(x) = \sum_{i=1}^{n} y_i \ell_i(x)$$

- note " $\prod$ " is a symbol for a product, just like " $\sum$ " is a symbol for sum
- we solve no linear systems and we just write down the answer! yeah!
- is this scheme a good idea in practice? NOT REALLY!

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## so, is the Lagrange scheme a good idea? cont.

- we have seen that actually using the formulas to find a familiar form for P(x) is ... awkward
- the problem with the Lagrange form is that even when we write down the correct linear combination of Lagrange polynomials ℓ<sub>i</sub>(x) to give P(x), we do not have quick ways of getting:
  - either the coefficients  $a_i$  in the standard form,

$$P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1}$$

• or the values of the polynomial P(x) at locations  $\bar{x}$  in between the  $x_i$ :

$$P(\bar{x}) = \bar{y}$$

- generally-speaking, the output values of a polynomial are the desired numbers; this is the purpose of polynomial *interpolation*
- **moral**: sometimes a *formula* for the answer is less useful than an algorithm that leads to the numbers you actually want
- ...and we'll get back to that!

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### conclusion: how to do polynomial interpolation

- the problem is to find the degree *n* − 1 polynomial *P*(*x*) which goes through *n* given points (*x<sub>i</sub>*, *y<sub>i</sub>*)
- we have three methods, all of which do the job:
  - o the Vandermonde matrix method,
  - o the Newton polynomial form, and its triangular matrix method,
  - and Lagrange's direct formula for the polynomial
- the first two require solving linear systems, while the last does not
  - Lagrange's direct formula requires us to simplify like crazy
  - Newton gives easier linear systems (triangular) than does Vandermonde
  - MATLAB makes solving linear systems easy anyway
- Iater in the course:
  - o we will address how accurate polynomial interpolation is
  - we will get to one more algorithm, Neville's algorithm, which gets the polynomial values but skips finding any coefficients

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