# Interpolation

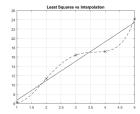
MATH-151: Mathematical Algorithms in Matlab

September 18, 2023



#### FITTING THE DATA

- Sometimes we are given some information and want to find a function that agrees with the data. This is a process called interpolation
  - Some of you have probably seen least-squares fits or regression before, that seeks to match the "trend" of the data, however with interpolation, we want to describe the functional relationship of the data and make sure all of our data points are included



 Most often we use polynomials as our interpolants, but different sets of functions can also be used!

### When is Interpolation Used?

- Interpolation is useful in many purposes when we are interested in representing data with a smooth function that captures changes in slope.
   For example
  - Plotting a smooth curve through data
  - Reading between entries of a data table
  - Differentiation or integration of function data
  - Approximating a complicated function with a simpler one
  - Using trigonometric functions to interpolate uniformly-spaced data points is called the Discrete Fourier Transform (or DFT) and commonly used in signal processing.

#### Lagrange Method

- Lagrange's method is centered around creating terms that equal  $f(x_i)$  when  $x=x_i$  and 0 when  $x=x_j, j\neq i$ .
- Let's start with the second part, how do we make our term be 0, when our x is any of our other points? We multiply by  $(x-x_j)$  for all the  $j \neq i$

We write this 
$$\prod_{j=1, j \neq i}^{N} (x - x_j)$$

• How do we make it correct when  $x=x_i$ ? We take our product above and divide by  $(x_i-x_j)$  for each  $j\neq i$  so we get 1, then multiply by  $f(x_i)$ 

This gives us 
$$f(x_i) \frac{\prod_{j=1, j \neq i}^N (x-x_j)}{\prod_{i=1, j \neq i}^N (x_i-x_j)}$$

• We repeat this for each of our points and add them together

$$p(x) = \sum_{i=1}^{N} f(x_i) \frac{\prod_{j=1, j \neq i}^{N} (x - x_j)}{\prod_{j=1, j \neq i}^{N} (x_i - x_j)}$$

#### LAGRANGE METHOD IN PRACTICE

- Lagrange's method looks a lot more complicated than it is. Let's see an
  example.
- Suppose we have 3 points. (2,5), (7,0), and (8,11).
- We start with  $x_1 = 2$ ,  $f(x_1) = 5$ . Our term to make sure this works is

$$f(x_1)\frac{(x-x_2)(x-x_3)}{(x_1-x_2)(x_1-x_3)} = 5\frac{(x-7)(x-8)}{(2-7)(2-8)}$$

ullet Repeating this process for  $x_2=7$  and  $x_3=8$  we can find our interpolating polynomial

$$p(x) = 5\frac{(x-7)(x-8)}{(2-7)(2-8)} + 0\frac{(x-2)(x-8)}{(7-2)(7-8)} + 11\frac{(x-2)(x-7)}{(8-2)(8-7)}$$

We can simplify this, but we don't have to. The computer loves to grind through arithmetic!

#### Newton Divided Differences Method

- Another way that we can find our interpolating polynomial with less computational cost is to use Newton's divided differences method
- ullet The divided differences are denoted by f[] and defined using a recursive definition

$$f[x_1, x_2, \dots, x_n] = \frac{f[x_2, \dots, x_n] - f[x_1, \dots, x_{n-1}]}{x_n - x_1}$$

where we start with base cases  $f[x_i] = f(x_i)$ 

And we use these to build our polynomial as follows

$$p(x) = f[x_1] + f[x_1, x_2](x - x_1) + \dots + f[x_1, \dots, x_n](x - x_1) \times \dots \times (x - x_{n-1})$$

 Just like Lagrange's method, this looks much more complicated when writing it abstractly than when performing it in practice, so lets see an example again.

## NEWTON DIVIDED DIFFERENCES IN PRACTICE

 Using our points from earlier we can compute our interpolating polynomial using our divided differences.

$$x f(x)$$

$$2 f[x_1] = 5$$

$$f[x_1, x_2] = \frac{0-5}{7-2} = 1$$

$$7 f[x_2] = 0$$

$$f[x_2, x_3] = \frac{11-0}{8-7} = 11$$

$$f[x_1, x_2, x_3] = \frac{11-(-1)}{8-2} = 2$$

$$8 f[x_3] = 11$$

Putting it all together we can find our interpolating polynomial is

$$f(x) = 5 + (-1)(x-2) + 2(x-2)(x-7)$$
 or, if we want to be fancy, 
$$f(x) = 5 + (x-2)\bigg(-1 + (x-7)(2)\bigg)$$

 If we want to do the work, we can see this is the same answer that Lagrange's method gave us!

#### WHY USE EACH METHOD?

- Both of these methods get us to the same answer, so why show both of them?
- Each method has its own strengths and weaknesses.
  - Lagrange's method is easier to understand and more straightfoward.
     But it is less flexible to new data, and often less computationally efficient
  - Newton's divided differences are easier to add new data points to and can be written to be very fast, however it is harder to understand and implement