

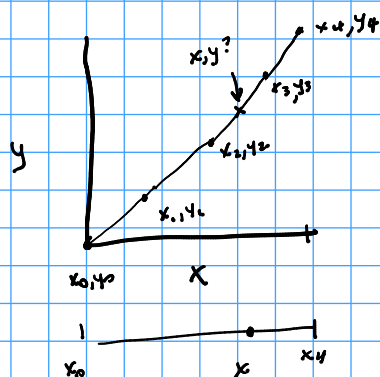
Lecture 18 - Interpolation

I. Concept of Interpolation

II. Linear interpolation by hand / Excel

III. Interpolation in Python

I. Concept of Interpolation

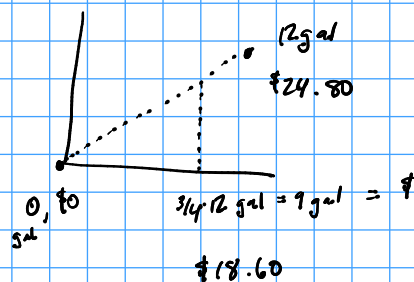


$\{x_i, y_i\}$ data
 $y(x)$ not in our set.

interpolation, but $x \in [\min, \max]$
 extrapolation $x \notin [\min, \max]$

Example: \$24.80 to fill up car w/ gas
 12 gal tank.

• suppose 1/4 full
 → fill 3/4



In interpolation, you want
 smooth functions that go through
all of the data.

→ piecewise
 functions

Piecewise polynomials

- why polynomials?

• Simple: lines, quadratic, ...

• Polynomial Uniqueness Theorem

2 points → 1 line

3 points → 1 quadratic

4 pts → 1 cubic

- why piecewise?

• wild oscillations - Runge's Phenomena

* A piecewise, polynomial that exactly fits every data point
 is called a spline.

- linear spline

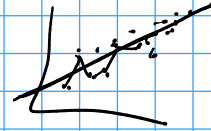


- quadratic spline (2nd, quadratic)
- cubic spline (3rd, cubic)

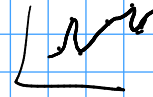
f, f', f'' continuous

Fitting vs. Interpolation

single Eq/ model	piecewise polynomials
Physically informed	purely empirical
Noisy Data	Smooth data
Approx. fit (only some/no pts)	Exact fit (every point)

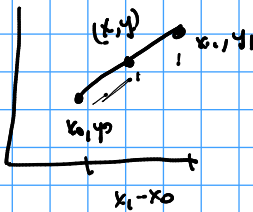


$x = vt$



II. Interpolation by hand & Excel

* Linear Interpolation (Excel has interpolation functions :))



$x_0 \ y_0$
 $x_1 \ y_1$
 $x \rightarrow y$

$$\frac{y - y_0}{x - x_0} = \frac{y_1 - y_0}{x_1 - x_0} \quad (\text{equal slopes})$$

$$y - y_0 = (x - x_0) \frac{y_1 - y_0}{x_1 - x_0}$$

$$y = y_0 + (x - x_0) \frac{(y_1 - y_0)}{x_1 - x_0} \quad \checkmark$$

Example from page 1:

$$x_0 = 0 \quad x_1 = 12 \text{ gal}$$

$$y_0 = \$0 \quad y_1 = \$24.80$$

$$x = \frac{3}{4} \cdot 12 \text{ gal} = 9 \text{ gal}$$

$$y = \$0 \cdot \frac{12 - 9}{12} + \$24.80 \cdot \frac{9 - 0}{12}$$

$$+ \frac{3}{4} \cdot \$24.80 = \$18.60$$

$$= y_0 \frac{x_1 - x}{x_1 - x_0} + y_1 \frac{x - x_0}{x_1 - x_0} - y_0 \frac{x - x_0}{x_1 - x_0}$$

$$y = y_0 \frac{x_1 - x}{x_1 - x_0} + y_1 \frac{x - x_0}{x_1 - x_0}$$

lever rule (use again in thermo)

III. Interpolation in Python

scipy.optimize $\rightarrow$ root
 $\rightarrow$ minimize
 $\rightarrow$ curve-fit

scipy.interpolate