

For the source code for the coding problems, see the attached Jupyter notebook, Matlab live script. Alternatively, the entire git repository is attached as a zip archive, and is available [on GitHub](#). The comments in the code have been omitted here for brevity. They are present in the Jupyter notebook.

- (1) **Problem Statement:** Compute the interpolating polynomial $p_2(x)$ that interpolates $f = \sqrt{2}x\cos(x)$ at points $x_0 = 0$, $x_1 = \frac{\pi}{4}$, and $x_2 = \frac{\pi}{2}$ in the interval $[0, \frac{\pi}{2}]$.

$$a(x_0)^2 + b(x_0) + c = 0$$

$$a(x_1)^2 + b(x_1) + c = \frac{\pi}{4}$$

$$a(x_2)^2 + b(x_2) + c = 0$$

We can solve this as a matrix-vector equation.

$$\begin{bmatrix} 0 & 0 & 1 \\ \frac{\pi^2}{16} & \frac{\pi}{4} & 1 \\ \frac{\pi^2}{4} & \frac{\pi}{2} & 1 \end{bmatrix} \times \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ \frac{\pi}{4} \\ 0 \end{bmatrix}$$

$$\left[\begin{array}{ccc|c} 0 & 0 & 1 & 0 \\ \frac{\pi^2}{16} & \frac{\pi}{4} & 1 & \frac{\pi}{4} \\ \frac{\pi^2}{4} & \frac{\pi}{2} & 1 & 0 \end{array} \right] \text{ swap(R1, R3)}$$

$$\left[\begin{array}{ccc|c} \frac{\pi^2}{4} & \frac{\pi}{2} & 1 & 0 \\ \frac{\pi^2}{16} & \frac{\pi}{4} & 1 & \frac{\pi}{4} \\ 0 & 0 & 1 & 0 \end{array} \right] \text{ swap(R1, R2)}$$

$$\left[\begin{array}{ccc|c} \frac{\pi^2}{16} & \frac{\pi}{4} & 1 & \frac{\pi}{4} \\ \frac{\pi^2}{4} & \frac{\pi}{2} & 1 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] \text{ R2} = \text{R2} - 4 \times \text{R1}$$

$$\left[\begin{array}{ccc|c} \frac{\pi^2}{16} & \frac{\pi}{4} & 1 & \frac{\pi}{4} \\ 0 & -\frac{\pi}{2} & -3 & -\pi \\ 0 & 0 & 1 & 0 \end{array} \right]$$

Now we can solve for a , b , and c .

$$1(c) = 0$$

$$c = 0$$

$$-\frac{\pi}{2}b - 3c = -\pi$$

$$-\frac{\pi}{2}b = -\pi$$

$$b = 2$$

$$\begin{aligned}
\frac{\pi^2}{16}a + \frac{\pi}{4}b + 1c &= \frac{\pi}{4} \\
\frac{\pi^2}{16}a + \frac{\pi}{4}2 + 1(0) &= \frac{\pi}{4} \\
\frac{\pi^2}{16}a + \frac{\pi}{2} &= \frac{\pi}{4} \\
\frac{\pi^2}{16}a &= -\frac{\pi}{4} \\
\pi^2 a &= -4\pi \\
a &= \frac{-4}{\pi}
\end{aligned}$$

Now that we have the coefficients, we can put them together to find the Lagrange polynomial $p_2(x) = -\frac{4}{\pi}x^2 + 2x$

- (2) **Problem Statement:** U.S. populations for the years 1990, 2000, and 2010, is given in the table

| x (year) | 1990 | 2000 | 2010 |
|------------------|-------------|-------------|-------------|
| y (population) | 248,709,873 | 281,421,906 | 308,745,538 |

Estimate the population in 2006 using

- (a) Linear Splines

$$\begin{aligned}
y(2006) &\approx \frac{y(2010) - y(2000)}{2010 - 2000}(2006 - 2000) + 281,421,906 \\
&\approx \frac{308,745,538 - 281,421,906}{10}6 + 281,421,906 \\
&\approx \frac{27,323,632}{10}6 + 281,421,906 \\
&\approx 2,732,363.2 \times 6 + 281,421,906 \\
y(2006) &\approx 297,816,085
\end{aligned}$$

- (b) $p_2(x)$ that interpolates the population of the U.S. at the data points $x_0 = 1990$, $x_1 = 2000$, and $x_2 = 2010$.

$$\begin{aligned}
&\left[\begin{array}{ccc|c} 3,960,100 & 1990 & 1 & 248,709,873 \\ 4,000,000 & 2000 & 1 & 281,421,906 \\ 4,040,100 & 2010 & 1 & 308,745,538 \end{array} \right] \begin{array}{l} \text{R2} = \text{R2} - 1.010075503\text{R1} \\ \text{R3} = \text{R3} - 1.020201510\text{R1} \end{array} \\
&\left[\begin{array}{ccc|c} 3,960,100 & 1990 & 1 & 248,709,873 \\ 0 & -10.05025097 & -0.010075503 & 30,206,155.93 \\ 0 & -20.2010049 & -0.020201510 & 55,011,350.01 \end{array} \right] \text{R3} = \text{R3} - 2.010000045\text{R2} \\
&\left[\begin{array}{ccc|c} 3,960,100 & 1990 & 1 & 248,709,873 \\ 0 & -10.05025097 & -0.010075503 & 30,206,155.93 \\ 0 & 0 & 0.00005025148340 & -5,703,024.769 \end{array} \right]
\end{aligned}$$

$$0.00005025148340c = -5703024.769$$

$$c = -113489679968.33304$$

$$\begin{aligned}
 -10.05025097b - 0.010075503c &= 30206155.93 \\
 -10.05025097b + 1143465610.9899793 &= 30206155.93 \\
 -10.05025097b &= -1113259455.0599792 \\
 b &= 110769318.93373197
 \end{aligned}$$

$$\begin{aligned}
 3960100a + 1990b + c &= 248709873 \\
 3960100a + 220430944678.12662 - 113489679968.33304 &= 2487098733960100a &= -106692554836.79358 \\
 a &= -26941.884002119536
 \end{aligned}$$

$$\begin{aligned}
 p_2(x) &= -26941.884x^2 + 110769318.93373x - 113489679968.33304 \\
 p_2(2006) &= 298,462,693
 \end{aligned}$$

(3) **Problem Statement:** Find b , c , and d such that

$$s(x) = \begin{cases} 1 + 2x - x^3, & 0 \leq x \leq 1 \\ 2 + b(x-1) + c(x-1)^2 + d(x-1)^3, & 1 < x \leq 2 \end{cases}$$

is a natural cubic spline.

$$s'(x) = \begin{cases} 2 - 3x^2, & 0 \leq x \leq 1 \\ b + 2c(x-1) + 3d(x-1)^2, & 1 < x \leq 2 \end{cases}$$

$$\lim_{x \rightarrow 1^+} s'(x) = \lim_{x \rightarrow 1^-} s'(x) \implies 2 - 3(1)^2 = b - 2c(1-1) + 3d(1-1)^2$$

$$2 - 3 = b - 2c(0) + 3d(0)^2$$

$$-1 = b$$

$$b = -1$$

$$s''(x) = \begin{cases} -6x, & 0 \leq x \leq 1 \\ 2c + 6d(x-1), & 1 < x \leq 2 \end{cases}$$

$$\lim_{x \rightarrow 1^+} s''(x) = \lim_{x \rightarrow 1^-} s''(x) \implies -6(1) = 2c + 6d(1-1)$$

$$-6 = 2c$$

$$c = -3$$

$$s''(0) = s''(2) = 0 \implies -6(0) = 2(-3) + 6d(2-1) = 0$$

$$-6 + 6d(1) = 0$$

$$6d = 6$$

$$d = 1$$

- (4) **Problem Statement:** The following function s is a cubic spline that interpolates the following values. Find a , b , c , d , y_1 , and y_2 .

$$s(x) = \begin{cases} \frac{5}{4}(x+2)^3 - \frac{9}{2}(x+2)^2 + 8, & -2 \leq x < 0 \\ ax^3 + bx^2 + cx + d, & 0 \leq x < 1 \\ (x-1)^3 - 1, & 1 \leq x < 2 \\ -\frac{5}{4}(x-2)^3 + 3(x-2)^2 + 3(x-2), & 2 \leq x \leq 4 \end{cases}$$

$$s'(x) = \begin{cases} \frac{15}{4}(x+2)^2 - 9(x+2), & -2 \leq x < 0 \\ 3ax^2 + 2bx + c, & 0 \leq x < 1 \\ 3(x-1)^2, & 1 \leq x < 2 \\ -\frac{15}{4}(x-2)^2 + 6(x-2), & 2 \leq x \leq 4 \end{cases}$$

$$s''(x) = \begin{cases} \frac{15}{2}(x+2) - 9, & -2 \leq x < 0 \\ 6ax + 2b, & 0 \leq x < 1 \\ 6(x-1), & 1 \leq x < 2 \\ -\frac{15}{2}(x-2) + 6, & 2 \leq x \leq 4 \end{cases}$$

| | | | | | |
|-----|----|-------|-------|---|---|
| x | -2 | 0 | 1 | 2 | 4 |
| y | 8 | y_1 | y_2 | 0 | 8 |

$$\begin{aligned} \lim_{x \rightarrow 0^+} s''(x) &= \lim_{x \rightarrow 0^-} s''(x) \implies \frac{15}{2}(0+2) - 9 = 6a(0) + 2b \\ 15 - 9 &= 2b \\ 6 &= 2b \\ b &= 3 \end{aligned}$$

$$\begin{aligned} \lim_{x \rightarrow 1^+} s''(x) &= \lim_{x \rightarrow 1^-} s''(x) \implies 6(1-1) = 6a(1) + 2b \\ 0 &= 6a + 2(3) \\ -6 &= 6a \\ a &= -1 \end{aligned}$$

$$\begin{aligned} \lim_{x \rightarrow 1^+} s'(x) &= \lim_{x \rightarrow 1^-} s'(x) \implies 3a(1)^2 + 2b(1) + c = 3(1-1)^2 \\ 3(-1) + 2(3) + c &= 3(0) \\ 6 - 3 + c &= 0 \\ 3 + c &= 0 \\ c &= -3 \end{aligned}$$

$$\begin{aligned} \lim_{x \rightarrow 1^+} s(x) &= \lim_{x \rightarrow 1^-} s(x) \implies a(1)^3 + b(1)^2 + c(1) + d = (1-1)^3 - 1 \\ -1 + 3 - 3 + d &= -1 \\ d &= 0 \end{aligned}$$

$$\begin{aligned}
 y_1 &= s(0) = ax^3 + bx^2 + cx + d \\
 &= -1(0)^3 + 3(0)^2 - 3(0) + 0 \\
 y_1 &= 0
 \end{aligned}$$

$$\begin{aligned}
 y_2 &= s(1) = (x - 1)^3 - 1 \\
 &= (1 - 1)^3 - 1 \\
 y_2 &= -1
 \end{aligned}$$

- (5) **Problem Statement:** Find the line of best fit for the following data. What is the error? What does it mean?

| i | x_i | y_i |
|-----|-------|-------|
| 1 | 2 | 3 |
| 2 | 4 | 7 |
| 3 | 6 | 11 |

$$\begin{aligned}
 &\begin{bmatrix} 2 & 1 \\ 4 & 1 \\ 6 & 1 \end{bmatrix} \times \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 7 \\ 11 \end{bmatrix} \\
 &\begin{bmatrix} 2 & 4 & 6 \\ 1 & 1 & 1 \end{bmatrix} \times \begin{bmatrix} 2 & 1 \\ 4 & 1 \\ 6 & 1 \end{bmatrix} \times \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 2 & 4 & 6 \\ 1 & 1 & 1 \end{bmatrix} \times \begin{bmatrix} 3 \\ 7 \\ 11 \end{bmatrix} \\
 &\begin{bmatrix} 56 & 12 \\ 12 & 3 \end{bmatrix} \times \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 100 \\ 21 \end{bmatrix} \\
 &\left[\begin{array}{cc|c} 56 & 12 & 100 \\ 12 & 3 & 21 \end{array} \right] \text{swap(R1, R2)} \\
 &\left[\begin{array}{cc|c} 12 & 3 & 21 \\ 56 & 12 & 100 \end{array} \right] \text{R2} = \text{R2} - \frac{14}{3}\text{R1} \\
 &\left[\begin{array}{cc|c} 12 & 3 & 21 \\ 0 & -2 & 2 \end{array} \right]
 \end{aligned}$$

$$\begin{aligned}
 -2a_2 &= 2 \\
 a_2 &= -1
 \end{aligned}$$

$$\begin{aligned}
 12a_1 + 3a_2 &= 21 \\
 12a_1 - 3 &= 21 \\
 12a_1 &= 24 \\
 a_1 &= 2
 \end{aligned}$$

The line of best fit is $y = 2x - 1$. The error is $\sum_{i=1}^3 (y_i - P(x_i))^2 = 0$. This means that the line exactly matches the data points. The error cannot "cancel out" since it is squared, and all of the terms have the same sign.

- (6) **Problem Statement:** Write a Matlab function, called `Lagrange_poly` that inputs a set of data points $(x, y) = (\text{datx}, \text{daty})$, a set x of numbers at which to interpolate, and outputs the polynomial interpolant, y , evaluated at x using Lagrange polynomial interpolation.

(a) Use the code to interpolate the following functions:

(i) $f_1(x) = e^{-x^2}$

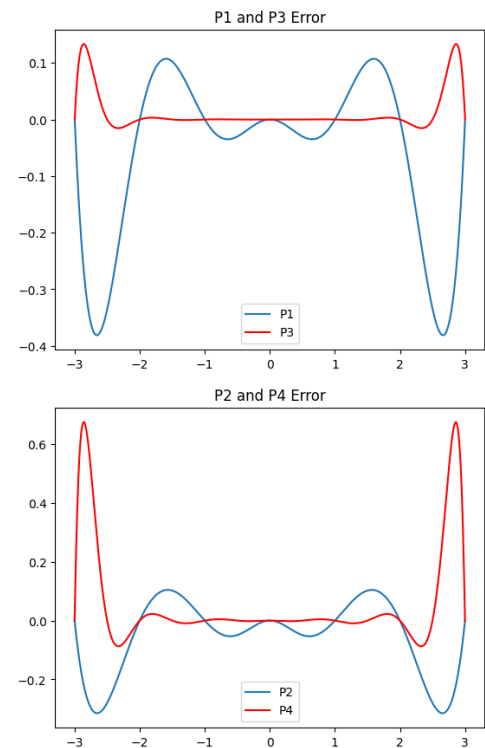
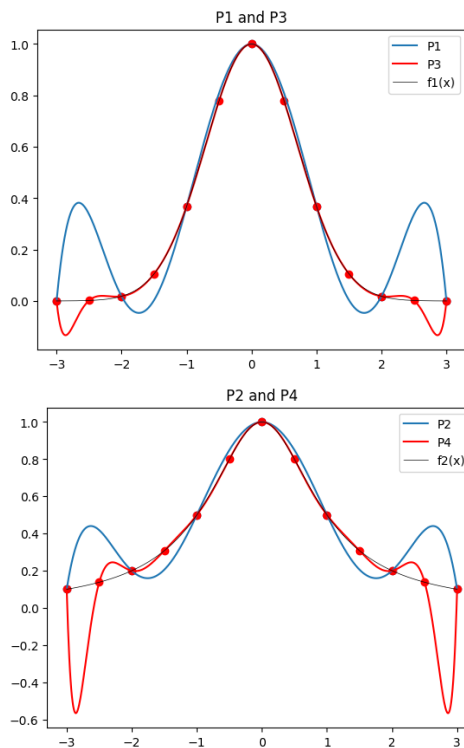
(ii) $f_2(x) = \frac{1}{1+x^2}$

using the data points `datx=-3:1:3`. Interpolate at the points `x=-3:0.01:3`. Call P_1 the Lagrange interpolant of f_1 and P_2 the Lagrange interpolant of f_2 . Repeat the experiment except using the data `datx1=-3:0.5:3`. Call in that case P_3 and P_4 the new interpolants. Compare the results.

For each interpolation problem, plot on the same graph the function, the two interpolants, and the data set. On a separate plot, plot the error between each interpolant y and $f(x)$.

```
def Lagrange_interp(x_data, y_data, x_interp):
    n = len(x_data)
    m = len(y_data)
    xs = len(x_interp)
    if(n != m):
        return
    p_interp = x_interp * 0
    for i in range(n):
        L_i = np.array([1.] * xs)
        for j in range(n):
            if(i == j):
                continue
            L_i *= (x_interp - x_data[j]) / (x_data[i] - x_data[j])
        p_interp += y_data[i] * L_i
    return(p_interp)
```

LISTING 1. Python



- (7) **Problem Statement:** Write a Matlab function called `linear_spline` which inputs a set of data points $(x, y) = (\text{datx}, \text{daty})$, a set x of numbers at which to interpolate, and outputs the polynomial interpolant, y , evaluated at x using a linear spline of interpolation.

(a) Use the code to interpolate the following functions

(i) $f_1(x) = e^{-x^2}$

(ii) $f_2(x) = \frac{1}{1+x^2}$

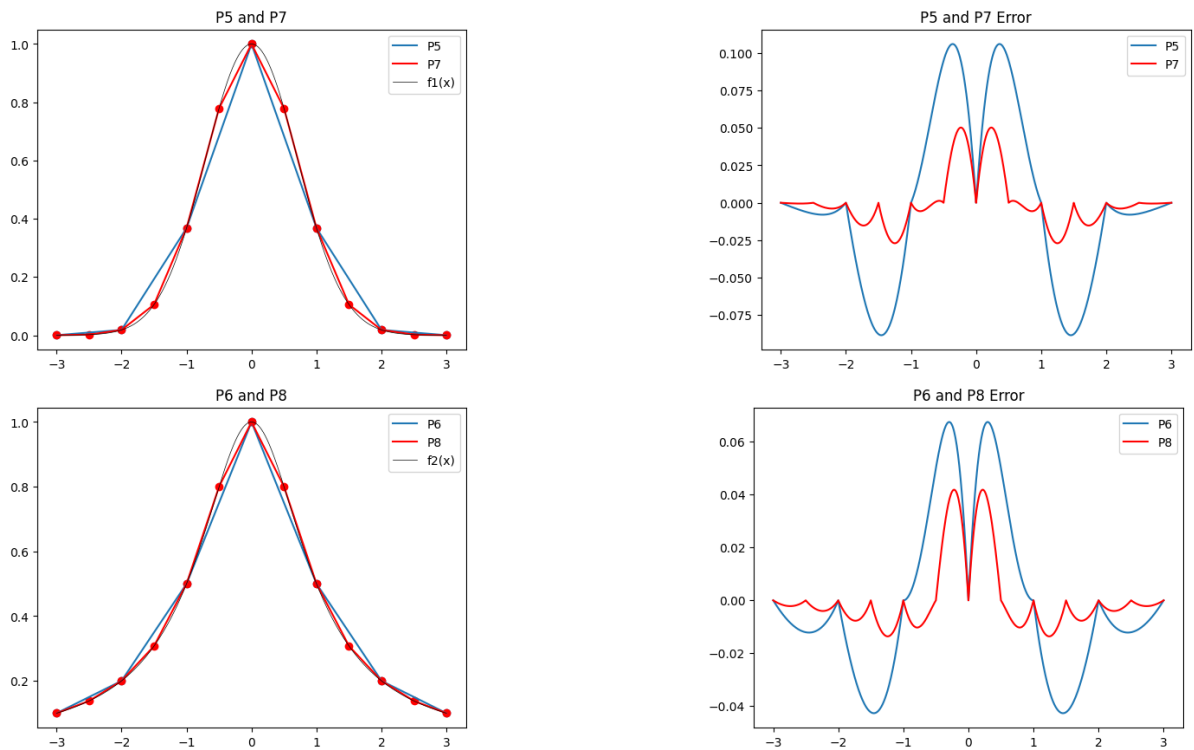
using the data points `datx=-3:1:3`. Interpolate at the points `x=-3:0.01:3`. Call P_5 the Linear Spline interpolant of f_1 and P_6 the Linear Spline interpolant of f_2 . Repeat the experiment except using the data `datx1=-3:0.5:3`. Call in that case P_7 and P_8 the new interpolants. Compare the results.

For each interpolation problem, plot on the same graph the function, the two interpolants, and the data set. On a separate plot, plot the error between each interpolant y and $f(x)$.

```
def linear_spline(x_data, y_data, x_interp):
    if(len(x_data) != len(y_data)):
        return
    p_interp = x_interp.copy()
    for i, x in enumerate(x_interp):
        dist = [(xd, yd) for xd, yd in zip(x_data, y_data)]
        dist.sort(key=lambda k: abs(k[0] - x))
        important = dist[:2]
        lowerx, lowery = min(important)
        higherx, highery = max(important)
        p_interp[i] = (highery - lowery) / (higherx - lowerx) * (x -
lowerx) + lowery

    return(p_interp)
```

LISTING 2. Python



- (8) **Problem Statement:** Write a Matlab function called `least_squares` which inputs a set of data points $(x, y) = (\text{datx}, \text{daty})$, the degree of polynomial n , and outputs the coefficients a_i as a vector.

Write a second function called `exp_least_squares` which computes the exponential components for least squares as $p_2(x) = a_0 e^{a_1 x}$ and outputs $a = (a_0, a_1)$

Write also at the top of the script an anonymous function `N(d)` which uses a normal distribution.

To create `daty` in each case, add `N(0.5)` to each data point.

- (a) Use the code to approximate the functions
 - (i) $f_1(x) = 2x + 4$ on $[0, 4]$ using $n = 1$.
 - (ii) $f_2(x) = x^2 - 3x + 1$ on $[1, 5]$ using $n = 2$.
- (b) Use the code to approximate
 - (i) $f_3(x) = 3e^{0.25x}$ on $[0, 10]$

```
def least_squares(datx, daty, n):
    X = np.vander(datx, n+1)
    coefficients = np.linalg.solve(X.T @ X, X.T @ daty)
    return(coefficients)

def exp_least_squares(datx, daty):
    log_daty = np.log(daty)
    X = np.vstack([np.ones_like(datx), datx]).T
    b, _, _, _ = np.linalg.lstsq(X, log_daty, rcond=None)
    b0 = b[0]
    b1 = b[1]
    a0 = np.exp(b0)
    a1 = b1
    return(a0, a1)
```

LISTING 3. Python