

Question 1 : Part 1 : minimum value of $S(a_0, a_1, a_2)$

$$z(x, y) = a_0 + a_1 x + a_2 y$$

$$S(a_0, a_1, a_2) = \sum_{i=1}^N [z_i - (a_0 + a_1 x_i + a_2 y_i)]^2$$

1.

$$\begin{aligned} \frac{\partial S}{\partial a_0} &= \sum_{i=1}^N 2 [z_i - (a_0 + a_1 x_i + a_2 y_i)] \cdot \frac{\partial}{\partial a_0} [- (a_0 + a_1 x_i + a_2 y_i)] \\ &= -2 \sum_{i=1}^N [z_i - (a_0 + a_1 x_i + a_2 y_i)] \end{aligned}$$

$$\sum_{i=1}^N [z_i - (a_0 + a_1 x_i + a_2 y_i)] = 0$$

$$\sum_i z_i - \sum_i a_0 - \sum_i a_1 x_i - \sum_i a_2 y_i = 0$$

$$\Rightarrow N a_0 + \left(\sum_i x_i \right) a_1 + \left(\sum_i y_i \right) a_2 = \sum_i z_i$$

2. $\frac{\partial S}{\partial a_1} = -2 \sum_{i=1}^N x_i [z_i - (a_0 + a_1 x_i + a_2 y_i)] = 0$

$$\sum_i x_i z_i - a_0 \sum_i x_i - a_1 \sum_i x_i^2 - a_2 \sum_i x_i y_i = 0$$

$$\left(\sum_i x_i \right) a_0 + \left(\sum_i x_i^2 \right) a_1 + \left(\sum_i x_i y_i \right) a_2 = \sum_i x_i z_i$$

$$3. \sum_i y_i z_i - a_0 \sum_i y_i - a_1 \sum_i x_i y_i - a_2 \sum_i y_i^2 = 0$$

$$\left(\sum_i y_i\right) a_0 + \left(\sum_i x_i y_i\right) a_1 + \left(\sum_i y_i^2\right) a_2 = \sum_i y_i z_i$$

abbrev: $S_0 = \sum_i 1 = N$ $S_x = \sum_i x_i$, $S_y = \sum_i y_i$,
 $S_{xx} = \sum_i x_i^2$, $S_{xy} = \sum_i x_i y_i$, $S_{yy} = \sum_i y_i^2$,
 $S_z = \sum_i z_i$ $S_{xz} = \sum_i x_i z_i$, $S_{yz} = \sum_i y_i z_i$

three equations become

$$\begin{cases} S_0 a_0 + S_x a_1 + S_y a_2 = S_z \\ S_x a_0 + S_{xx} a_1 + S_{xy} a_2 = S_{xz} \\ S_y a_0 + S_{xy} a_1 + S_{yy} a_2 = S_{yz} \end{cases}$$

Compact Matrix form:

$$\begin{bmatrix} S_0 & S_x & S_y \\ S_x & S_{xx} & S_{xy} \\ S_y & S_{xy} & S_{yy} \end{bmatrix} \cdot \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} S_z \\ S_{xz} \\ S_{yz} \end{bmatrix}$$

Homework 5

May 9, 2025

Question 1: Part 2, 3, and 4

```
[13]: import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
import pandas as pd

# Read the borehole data
borehole_data = pd.read_csv('borehole-data.csv')
print("Borehole Data:")
print(borehole_data)

# Print column names to debug
print("\nColumn names in the CSV file:")
for col in borehole_data.columns:
    print(f"{col}")

# Extract the data
upper_col = [col for col in borehole_data.columns if 'Upper Surface' in col][0]
lower_col = [col for col in borehole_data.columns if 'Lower Surface' in col][0]

x = borehole_data['X(m)'].values
y = borehole_data['Y(m)'].values
upper_z = borehole_data[upper_col].values
lower_z = borehole_data[lower_col].values
N = len(x)

print(f"\nUsing columns: X(m), Y(m), {upper_col}, {lower_col}")

# Create a 3D plot
fig = plt.figure(figsize=(12, 10))
ax = fig.add_subplot(111, projection='3d')

# Plot the upper and lower surfaces
ax.scatter(x, y, upper_z, c='b', marker='o', label='Upper Surface')
ax.scatter(x, y, lower_z, c='r', marker='^', label='Lower Surface')

# Setting labels and title
```

```

ax.set_xlabel('X (m)')
ax.set_ylabel('Y (m)')
ax.set_zlabel('Depth (m)')
ax.set_title('Mineral Deposits: Upper and Lower Surfaces')
ax.legend()

# Set up matrix equations for upper surface:  $z(x, y) = a_0 + a_1x + a_2y$ 

A = np.zeros((3, 3))
A[0, 0] = N
A[0, 1] = np.sum(x)
A[0, 2] = np.sum(y)
A[1, 0] = np.sum(x)
A[1, 1] = np.sum(x**2)
A[1, 2] = np.sum(x * y)
A[2, 0] = np.sum(y)
A[2, 1] = np.sum(x * y)
A[2, 2] = np.sum(y**2)

# Right hand side for upper surface
b_upper = np.zeros(3)
b_upper[0] = np.sum(upper_z)
b_upper[1] = np.sum(x * upper_z)
b_upper[2] = np.sum(y * upper_z)

# Right hand side for lower surface
b_lower = np.zeros(3)
b_lower[0] = np.sum(lower_z)
b_lower[1] = np.sum(x * lower_z)
b_lower[2] = np.sum(y * lower_z)

# Solve for coefficients
coeffs_upper = np.linalg.solve(A, b_upper)
coeffs_lower = np.linalg.solve(A, b_lower)

print("\nUpper Surface Equation:  $z(x,y) = {:.4f} + {:.4f}x + {:.4f}y$ ".format(
    coeffs_upper[0], coeffs_upper[1], coeffs_upper[2]))
print("Lower Surface Equation:  $z(x,y) = {:.4f} + {:.4f}x + {:.4f}y$ ".format(
    coeffs_lower[0], coeffs_lower[1], coeffs_lower[2]))

# Create a grid for the plane surfaces
grid_x, grid_y = np.meshgrid(np.linspace(min(x), max(x), 50),
                             np.linspace(min(y), max(y), 50))
grid_z_upper = coeffs_upper[0] + coeffs_upper[1] * grid_x + coeffs_upper[2] * grid_y
grid_z_lower = coeffs_lower[0] + coeffs_lower[1] * grid_x + coeffs_lower[2] * grid_y

```

```

# Plot the planes
ax.plot_surface(grid_x, grid_y, grid_z_upper, alpha=0.3, color='blue')
ax.plot_surface(grid_x, grid_y, grid_z_lower, alpha=0.3, color='red')

# Calculate average depth and volume
# Site dimensions are explicitly given as 2 km × 2 km
site_length = 2000 # 2 km = 2000 m
site_width = 2000 # 2 km = 2000 m
site_area = site_length * site_width # 4,000,000 m2

# Sample the planes at a grid of points and calculate average thickness
num_samples = 100
sample_x = np.linspace(min(x), max(x), num_samples)
sample_y = np.linspace(min(y), max(y), num_samples)
xx, yy = np.meshgrid(sample_x, sample_y)
upper_surface = coeffs_upper[0] + coeffs_upper[1] * xx + coeffs_upper[2] * yy
lower_surface = coeffs_lower[0] + coeffs_lower[1] * xx + coeffs_lower[2] * yy

# For raw data points
thickness_raw = np.abs(upper_z - lower_z)
avg_thickness_raw = np.mean(thickness_raw)

# For the fitted planes
# Both should be negative values, with lower surface more negative than upper
thickness_planes = np.abs(lower_surface - upper_surface) # Should be positive_
↪ values
avg_thickness_planes = np.mean(thickness_planes)

# Calculate volume (area × average thickness) - using the fitted planes value
volume = site_area * avg_thickness_planes

print(f"Average thickness from raw data points: {avg_thickness_raw:.2f} m")
print(f"Average thickness from fitted planes: {avg_thickness_planes:.2f} m")

# Print results
print("\nCalculations:")
print(f"Site dimensions: {site_length} m × {site_width} m")
print(f"Site area: {site_area} m2")
print(f"Average mineral deposit thickness: {avg_thickness_planes:.2f} m")

```

POSSIBLE DATA LOSS Some features might be lost if you save this workbook in the comma-delimited (.csv) format. To

[illegible]

```
print(f"Total mineral deposit volume: {volume:.2f} m³")

# Display the plot
plt.tight_layout()
plt.show()
```

Borehole Data:

	Borehole	X(m)	Y(m)	Upper Surface (m)	Lower Surface (m)
0	1	30	30	-18.5	-42.5
1	2	770	30	-17.5	-41.8
2	3	1230	30	-16.0	-41.3
3	4	1970	30	-14.6	-40.5
4	5	30	770	-32.2	-43.4
5	6	770	770	-20.8	-42.6
6	7	1230	770	-19.8	-42.1
7	8	1970	770	-18.3	-41.4
8	9	30	1230	-24.7	-44.0
9	10	770	1230	-23.2	-43.2
10	11	1230	1230	-22.2	-42.7
11	12	1970	1230	-20.8	-42.0
12	13	30	1970	-28.4	-44.8
13	14	770	1970	-27.0	-44.1
14	15	1230	1970	-26.0	-43.6
15	16	1970	1970	-24.5	-43.9

Column names in the CSV file:

```
'Borehole'
'X(m)'
'Y(m)'
'Upper Surface (m)'
' Lower Surface (m)'
```

Using columns: X(m), Y(m), Upper Surface (m), Lower Surface (m)

Upper Surface Equation: $z(x,y) = -20.6207 + 0.0033x + -0.0048y$

Lower Surface Equation: $z(x,y) = -42.3174 + 0.0009x + -0.0013y$

Average thickness from raw data points: 20.59 m

Average thickness from fitted planes: 20.59 m

Calculations:

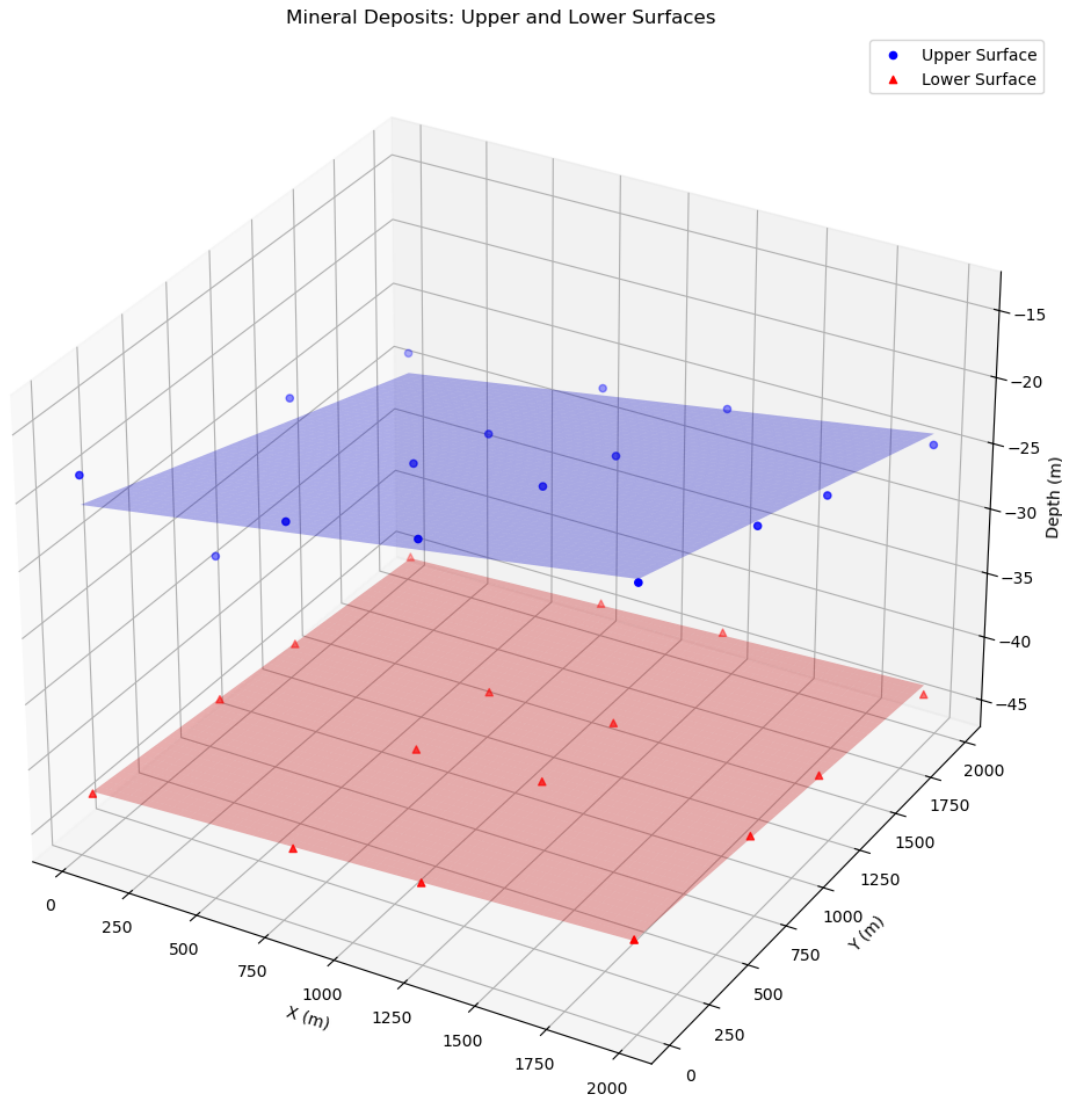
Site dimensions: 2000 m × 2000 m

Site area: 4000000 m²

Average mineral deposit thickness: 20.59 m

Total mineral deposit volume: 82350000.00 m³

Q2: Part 3



Question 4: beginning

```
[7]: import numpy as np

# Define the function to be integrated
def f(x):
    return 3*x**2 + 4*x**3 + 5*x**4

# Define the exact solution (given)
exact_solution = 56

# Integration limits
a = 0
b = 2
```


Question 2: Part 1: Divided differences

i	x_i	$f[x_i]$	$f[x_i, x_{i+1}]$	$f[x_i, x_{i+1}, x_{i+2}]$	$f[x_i, x_{i+1}, x_{i+2}, x_{i+3}]$
0	0.0	36.0	$\frac{32-36}{2-0} = -2.0$	$\frac{-5-(-2)}{3-0} = -1.$	$\frac{-1-(-1)}{5-0} = 0$
1	2.0	32.0	$\frac{27-32}{3-2} = -5.0$	$\frac{-8-(-5)}{5-2} = -1.$	
2	3.0	27.0	$\frac{11-27}{5-3} = -8.0$		
3	5.0	11.0			

first differences Second differences Third differences

$$\begin{aligned} f[x_0, x_1] &= -2 & f[x_0, x_1, x_2] &= -1 & f[x_0, x_1, x_2, x_3] &= 0 \\ f[x_1, x_2] &= -5 & f[x_1, x_2, x_3] &= -1 & & \\ f[x_2, x_3] &= -8 & & & & \end{aligned}$$

$$P(x) = f[x_0] + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1)$$

→ Substituting:

$$P(x) = 36 + (-2)x + (-1)x(x-2) = 36 - 2x - x^2 + 2x = \boxed{36 - x^2}$$

$$P(0) = 36, \quad P(2) = 32, \quad P(3) = 27, \quad P(5) = 11$$

Part 2: Lagrange Interpolation

$$P(x) = \sum_{i=0}^3 f(x_i) L_i(x), \quad L_i(x) = \prod_{\substack{j=0 \\ j \neq i}}^3 \frac{x - x_j}{x_i - x_j}$$

$$L_0(x) = \frac{(x-2)(x-3)(x-5)}{(0-2)(0-3)(0-5)} = \frac{(x-2)(x-3)(x-5)}{-30}$$

$$L_1(x) = \frac{(x-0)(x-3)(x-5)}{(2-0)(2-3)(2-5)} = \frac{(x)(x-3)(x-5)}{6}$$

$$L_2(x) = \frac{(x-0)(x-2)(x-5)}{(3-0)(3-2)(3-5)} = \frac{(x)(x-2)(x-5)}{-6}$$

$$L_3(x) = \frac{(x-0)(x-2)(x-3)}{(5-0)(5-2)(5-3)} = \frac{(x)(x-2)(x-3)}{30}$$

$$P(x) = 36 L_0(x) + 32 L_1(x) + 27 L_2(x) + 11 L_3(x)$$

$$P(x) = 36 \frac{(x-2)(x-3)(x-5)}{-30} + 32 \frac{x(x-3)(x-5)}{6} + 27 \frac{x(x-2)(x-5)}{-6} - 11 \frac{x(x-2)(x-3)}{30}$$

Numerators:

$$1. \quad \overbrace{(x-2)(x-3)(x-5)} = (x^2 - 5x + 6)(x-5) = x^3 - 10x^2 + 31x - 30$$

$$2. \cancel{x}(x-3)(\cancel{x}-5) = (x^2 - 8x + 15)x = x^3 - 8x^2 + 15x$$

$$3. x(x-2)(x-5) = (x^2 - 7x + 10)x = x^3 - 7x^2 + 10x$$

$$4. x(x-2)(x-3) = (x^2 - 5x + 6)x = x^3 - 5x^2 + 6x$$

Term 1

$$36 \cdot \frac{x^3 - 10x^2 + 31x - 30}{-30} = -\frac{36}{30} (x^3 - 10x^2 + 31x - 30) = \boxed{-\frac{6}{5}x^3} + \boxed{12x^2} - \boxed{\frac{186}{5}x} + \boxed{36}$$

Term 2

$$32 \cdot \frac{x^3 - 8x^2 + 15x}{6} = \frac{32}{6} (x^3 - 8x^2 + 15x) = \boxed{\frac{16}{3}x^3} - \boxed{\frac{128}{3}x^2} + \boxed{80x}$$

Term 3

$$27 \cdot \frac{-(x^3 - 7x^2 + 10x)}{6} = -\frac{27}{6} (x^3 - 7x^2 + 10x) = \boxed{-\frac{9}{2}x^3} + \boxed{\frac{63}{2}x^2} - \boxed{45x}$$

Term 4:

$$11 \cdot \frac{x^3 - 5x^2 + 6x}{30} = \boxed{\frac{11}{30}x^3} - \boxed{\frac{11}{6}x^2} + \boxed{\frac{11}{5}x}$$

$$\rightarrow \frac{-36 + 160 - 136 + 11}{30} = 0(x^3)$$

$$\rightarrow 12 - \frac{128}{3} + \frac{63}{2} - \frac{11}{6} = -1(x^2)$$

$$\rightarrow -\frac{186}{5} + 80 - 45 + \frac{11}{5} = 0(x)$$

$$\rightarrow + 36$$

$$\Rightarrow P(x) = 36 + (-1)x^2 = \boxed{36 - x^2}$$

Question 3

Part 1: Intersection equation

$$f(x) = x^2 - 16$$

$$> f(x) = g(x)$$

$$g(x) = \frac{x^3}{8} - 2x + 5$$

$$\Rightarrow x^2 - 16 = \frac{x^3}{8} - 2x + 5$$

$$8 \cdot (x^2 - 16) = \cancel{8} \left(\frac{\cancel{x^3}}{\cancel{8}} - 2x + 5 \right)$$

$$8x^2 - 128 = x^3 - 16x + 40$$

$$\rightarrow 8x^2 - 16x + 168 = 0$$

$$\text{long division: } (x-6)(x^2 - 2x - 28) = 0$$

$$\begin{array}{l} / \\ x = 6 \end{array}$$

$$\backslash \quad x_{1,2} = \frac{-(-2) \pm \sqrt{(-2)^2 - 4 \cdot 1 \cdot (-28)}}{2 \cdot 1}$$

$$x_{1,2} = \frac{2 \pm \sqrt{4 + 112}}{2} = \frac{2 \pm 2\sqrt{29}}{2}$$

$$x_{1,2} = 1 \pm \sqrt{29}$$

The two intersection labeled A and E in the figure lie at :

$$A: x = 1 - \sqrt{29} \quad E: x = 1 + \sqrt{29}$$

Part 2: Area

$x = A$ and $x = E$, since $g(x) \geq f(x)$ for x from A up to the center crossing at $x = 6$, and $f(x) \geq g(x)$ for $6 \leq x \leq E$

$$\text{Area} = \int_A^6 [g(x) - f(x)] dx + \int_6^E [f(x) - g(x)] dx$$

$$h(x) = g(x) - f(x) = \frac{x^3}{8} - x^2 - 2x + 21$$

$$H(x) = \int h(x) dx = \frac{x^4}{32} - \frac{x^3}{3} - x^2 + 21x$$

$$\begin{aligned} \text{Area} &= [H(6) - H(A)] + [H(6) - H(E)] \\ &= 2H(6) - H(A) - H(E) \end{aligned}$$

$$\begin{aligned} &\left[2 \left(\frac{(6)^4}{32} - \frac{(6)^3}{3} - (6)^2 + 21(6) \right) \right] - \left[\left(\frac{(1-\sqrt{29})^4}{32} - \frac{(1-\sqrt{29})^3}{3} - (1-\sqrt{29})^2 + 21(1-\sqrt{29}) \right) \right] \\ &- \left[\left(\frac{(1+\sqrt{29})^4}{32} - \frac{(1+\sqrt{29})^3}{3} - (1+\sqrt{29})^2 + 21(1+\sqrt{29}) \right) \right] \end{aligned}$$

$$= \frac{781 + 145\sqrt{29}}{12} = 130.1540748$$

Part 3: Simpson's Rule

Single Simpson's step over $[A, E]$ with $M = \frac{A+E}{2} = 1$

then:

$$\text{Area}_s \approx \frac{E-A}{6} [h(A) + 4h(1) + h(E)]$$

But $h(A) = g(A) - f(A) = 0$ and $h(E) = 0$, while

$$h(1) = \frac{1^3}{8} - 1^2 - 2(1) + 21 = 0.125 - 1 - 2 + 21 = 18.125$$

\therefore

$$\begin{aligned} \text{Area}_s &= \frac{(1+\sqrt{29}) - (1-\sqrt{29})}{6} (0 + 4(18.125) + 0) \\ &= \frac{2\sqrt{29}}{6} \times 72.5 = \frac{145\sqrt{29}}{6} \approx 130.141 \end{aligned}$$

Part 4: Error estimation and Comparison.

Theoretical Error bound for Simpson's Rule over $[A, E]$ is

$$E_s = - \frac{(E-A)h^4}{180} h^{(4)}\left(\frac{E}{2}\right)$$

Since $h(x) = g(x) - f(x)$ is a cubic polynomial, $h^{(4)} \equiv 0$

so Simpson's Rule is exact for the net integral $g-f$.

Comparison: Calculus area = 130.154

Simpson approx. = 130.141

$$|\text{Error}| : \text{difference} = \underline{130.154 - 130.141} = 0.013$$

$$\frac{|\text{Error}|}{\text{actual}} \times 100 = \% \text{ error} \rightarrow \frac{0.013}{130.154} \times 100 = 0.01 \% \quad \underbrace{\hspace{2cm}}_{\text{excellent agreement}}$$

Because the true area requires taking the absolute value of $g-f$ (which breaks the integrand into two cubic pieces), Simpson's rule no longer integrates that exactly, but even so the single-step estimate errors by ~ 0.013 units.

Question 4: continued

```
# --- 1. Trapezoidal Rule ---
# Using 3 data ordinates means 2 intervals
n_trapezoid = 2
x_trapezoid = np.linspace(a, b, n_trapezoid + 1)
y_trapezoid = f(x_trapezoid)
h_trapezoid = (b - a) / n_trapezoid

# Trapezoidal rule formula
trapezoidal_approximation = (h_trapezoid / 2) * (y_trapezoid[0] + 2*np.
    ↪sum(y_trapezoid[1:-1]) + y_trapezoid[-1])

# --- 2. Simpson's Rule ---
# Using 3 data ordinates means 2 intervals
n_simpson = 2
x_simpson = np.linspace(a, b, n_simpson + 1)
y_simpson = f(x_simpson)
h_simpson = (b - a) / n_simpson

# Simpson's rule formula
simpson_approximation = (h_simpson / 3) * (y_simpson[0] + 4*np.sum(y_simpson[1:
    ↪-1:2]) + 2*np.sum(y_simpson[2:-2:2]) + y_simpson[-1])

# --- 3. Two-Point Gauss Quadrature ---

gauss_points = np.array([-1/np.sqrt(3), 1/np.sqrt(3)])
gauss_weights = np.array([1, 1])

# Transformation from [-1, 1] to [a, b]
x_gauss = ((b - a) / 2) * gauss_points + ((a + b) / 2)
y_gauss = f(x_gauss)

# Gauss Quadrature formula
gauss_approximation = ((b - a) / 2) * np.sum(gauss_weights * y_gauss)

# --- Error Calculations ---

# Absolute Error = |Approximate Value - Exact Value|
abs_error_trapezoid = np.abs(trapezoidal_approximation - exact_solution)
abs_error_simpson = np.abs(simpson_approximation - exact_solution)
abs_error_gauss = np.abs(gauss_approximation - exact_solution)

# Relative Error = |Absolute Error / Exact Value|
rel_error_trapezoid = abs_error_trapezoid / np.abs(exact_solution)
rel_error_simpson = abs_error_simpson / np.abs(exact_solution)
rel_error_gauss = abs_error_gauss / np.abs(exact_solution)
```



```
# --- Output ---
```

```
print("Numerical Integration Results:\n")
print(f"Trapezoidal Rule Approximation: {trapezoidal_approximation}")
print(f"Simpson's Rule Approximation: {simpson_approximation}")
print(f"Gauss Quadrature Approximation: {gauss_approximation}\n")

print("Error Analysis:\n")
print(f"Trapezoidal Rule Absolute Error: {abs_error_trapezoid}")
print(f"Trapezoidal Rule Relative Error: {rel_error_trapezoid}")
print(f"Simpson's Rule Absolute Error: {abs_error_simpson}")
print(f"Simpson's Rule Relative Error: {rel_error_simpson}")
print(f"Gauss Quadrature Absolute Error: {abs_error_gauss}")
print(f"Gauss Quadrature Relative Error: {rel_error_gauss}")
```

Numerical Integration Results:

Trapezoidal Rule Approximation: 74.0

Simpson's Rule Approximation: 57.33333333333333

Gauss Quadrature Approximation: 55.11111111111111

Error Analysis:

Trapezoidal Rule Absolute Error: 18.0

Trapezoidal Rule Relative Error: 0.32142857142857145

Simpson's Rule Absolute Error: 1.3333333333333286

Simpson's Rule Relative Error: 0.023809523809523725

Gauss Quadrature Absolute Error: 0.8888888888888928

Gauss Quadrature Relative Error: 0.015873015873015945

Question 5

```
[5]: import numpy as np
import pandas as pd

def f(x):
    """The function to integrate: 4/(1+x^2)"""
    return 4 / (1 + x**2)

def trapezoid(f, a, b, n):
    """
    Computes the trapezoid rule approximation with n subintervals
    """
    h = (b - a) / n
    x = np.linspace(a, b, n+1)
    y = [f(xi) for xi in x]
```

```

    # Trapezoid rule formula:  $h/2 * (f(a) + 2f(x) + 2f(x) + \dots + 2f(x) + f(b))$ 
    ↪ f(b))
    return h/2 * (y[0] + 2*sum(y[1:-1]) + y[-1])

def romberg(f, a, b, max_level=4):
    """
    Performs Romberg integration
    Returns R matrix and errors
    """
    R = np.zeros((max_level+1, max_level+1))
    h = b - a

    # First column (R(i,0)) - trapezoid rule with h, h/2, h/4, ...
    for i in range(max_level+1):
        n = 2**i
        R[i, 0] = trapezoid(f, a, b, n)

    # Fill the Romberg table using the extrapolation formula
    for j in range(1, max_level+1):
        for i in range(j, max_level+1):
            # Romberg extrapolation formula
            R[i, j] = R[i, j-1] + (R[i, j-1] - R[i-1, j-1]) / (4**j - 1)

    # True value of the integral is
    true_value = np.pi
    abs_errors = np.abs(R - true_value)
    rel_errors = abs_errors / true_value

    return R, abs_errors, rel_errors

def print_results(R, abs_errors, rel_errors, max_level):
    """Prints Romberg table and errors in nicely formatted tables"""
    print("Romberg Integration Table:")
    rows = []

    for i in range(max_level+1):
        row = [f"R({i},{j})" for j in range(i+1)]
        rows.append(row)

    df_labels = pd.DataFrame(rows)

    # Create table of values
    rows_values = []
    for i in range(max_level+1):
        row = [f"{R[i,j]:.10f}" for j in range(i+1)]
        rows_values.append(row)

```

```

df_values = pd.DataFrame(rows_values)

# Print label and value
for i in range(max_level+1):
    line = ""
    for j in range(i+1):
        line += f"{df_labels.iloc[i,j]}: {df_values.iloc[i,j]} "
    print(line)

print("\nAbsolute Errors:")
rows_abs_err = []
for i in range(max_level+1):
    row = [f"{abs_errors[i,j]:.10f}" for j in range(i+1)]
    rows_abs_err.append(row)

df_abs_err = pd.DataFrame(rows_abs_err)

for i in range(max_level+1):
    line = ""
    for j in range(i+1):
        line += f"E({i},{j}): {df_abs_err.iloc[i,j]} "
    print(line)

print("\nRelative Errors (%)")
rows_rel_err = []
for i in range(max_level+1):
    row = [f"{rel_errors[i,j]*100:.8f}%" for j in range(i+1)]
    rows_rel_err.append(row)

df_rel_err = pd.DataFrame(rows_rel_err)

for i in range(max_level+1):
    line = ""
    for j in range(i+1):
        line += f"RE({i},{j}): {df_rel_err.iloc[i,j]} "
    print(line)

# Main execution
max_level = 4 # Number of levels in Romberg table
a, b = 0, 1 # Integration limits
R, abs_errors, rel_errors = romberg(f, a, b, max_level)

print(f"True value = {np.pi:.10f}")
print_results(R, abs_errors, rel_errors, max_level)

# Specifically evaluate at the points mentioned in the problem
print("\nEvaluating f(x) at specified points:")

```

```

points = [0, 1/4, 1/2, 3/4, 1]
for x in points:
    print(f"f({x}) = {f(x):.10f}")

# To show how starting with those specific points works in Romberg integration:
print("\nRomberg integration starting with points [0, 1/4, 1/2, 3/4, 1]:")
# This means we start with n=4 intervals
h = 1/4
x_values = np.array([0, 1/4, 1/2, 3/4, 1])
y_values = f(x_values)

# Calculate initial trapezoid rule approximation with these points
R0 = h/2 * (y_values[0] + 2*sum(y_values[1:-1]) + y_values[-1])
print(f"Initial trapezoid approximation (h=1/4): {R0:.10f}")
print(f"Absolute error: {abs(R0 - np.pi):.10f}")
print(f"Relative error: {abs(R0 - np.pi)/np.pi*100:.8f}%")

```

True value = 3.1415926536

Romberg Integration Table:

```

R(0,0): 3.0000000000
R(1,0): 3.1000000000  R(1,1): 3.1333333333
R(2,0): 3.1311764706  R(2,1): 3.1415686275  R(2,2): 3.1421176471
R(3,0): 3.1389884945  R(3,1): 3.1415925025  R(3,2): 3.1415940941  R(3,3):
3.1415857838
R(4,0): 3.1409416120  R(4,1): 3.1415926512  R(4,2): 3.1415926611  R(4,3):
3.1415926384  R(4,4): 3.1415926653

```

Absolute Errors:

```

E(0,0): 0.1415926536
E(1,0): 0.0415926536  E(1,1): 0.0082593203
E(2,0): 0.0104161830  E(2,1): 0.0000240261  E(2,2): 0.0005249935
E(3,0): 0.0026041591  E(3,1): 0.0000001511  E(3,2): 0.0000014405  E(3,3):
0.0000068698
E(4,0): 0.0006510415  E(4,1): 0.0000000024  E(4,2): 0.0000000076  E(4,3):
0.0000000152  E(4,4): 0.0000000117

```

Relative Errors (%):

```

RE(0,0): 4.50703414%
RE(1,0): 1.32393528%  RE(1,1): 0.26290233%
RE(2,0): 0.33155740%  RE(2,1): 0.00076478%  RE(2,2): 0.01671106%
RE(3,0): 0.08289296%  RE(3,1): 0.00000481%  RE(3,2): 0.00004585%  RE(3,3):
0.00021867%
RE(4,0): 0.02072330%  RE(4,1): 0.00000008%  RE(4,2): 0.00000024%  RE(4,3):
0.00000048%  RE(4,4): 0.00000037%

```

Evaluating f(x) at specified points:

```

f(0) = 4.0000000000
f(0.25) = 3.7647058824

```

```
f(0.5) = 3.2000000000  
f(0.75) = 2.5600000000  
f(1) = 2.0000000000
```

```
Romberg integration starting with points [0, 1/4, 1/2, 3/4, 1]:  
Initial trapezoid approximation (h=1/4): 3.1311764706  
Absolute error: 0.0104161830  
Relative error: 0.33155740%
```

[]:

Question 6: Part 1: Romberg integration to $O(h^6)$

$$R_{i,0} = T_{2^i} \quad (\text{composite trapezoid with } 2^i \text{ panels})$$

$$R_{i,j} = R_{i,j-1} + \frac{R_{i,j-1} - R_{i-1,j-1}}{4^j - 1}$$

$$\text{Let } f(x) = x^3(16 - x^2).$$

$$f(0) = 0, \quad f(1) = 1^3(16-1) = 15, \quad f(2) = 8(16-4) = 96, \quad f(3) = 27(16-9) = 189$$

$$f(4) = 64(16-16) = 0$$

$$a) \quad R_{0,0} = T_1 \quad (\text{one panel, } h=4)$$

$$T_1 = \frac{h}{2} [f(0) + f(4)] = \frac{4}{2} (0+0) = 0$$

$$b) \quad R_{1,0} = T_2 \quad (\text{two panels, } h=2) \quad x=0, 2, 4:$$

$$T_2 = \frac{2}{2} [f(0) + 2f(2) + f(4)] = 1 \cdot (0 + 2 \cdot 96 + 0) = 192$$

$$c) \quad R_{2,0} = T_4 \quad (\text{four panels, } h=1) \quad x=0, 1, 2, 3, 4:$$

$$T_4 = \frac{1}{2} [f(0) + 2[f(1) + f(2) + f(3) + f(4)]]$$

$$= \frac{1}{2} (0 + 2(15 + 96 + 189 + 0)) = \frac{1}{2} \cdot 600 = 300$$

Richardson extrapolation

1. Level 1, column 1

$$R_{1,1} = R_{1,0} + \frac{R_{1,0} - R_{0,0}}{4^1 - 1} = 192 + \frac{192 - 0}{3} = 192 + 64 = 256$$

2. Level 2, column 1:

$$R_{2,1} = 300 + \frac{300 - 192}{3} = 300 + 36 = 336$$

3. Level 2, (Column 2 of $O(h^6)$):

$$R_{2,2} = R_{2,1} + \frac{R_{2,1} - R_{1,1}}{4^2 - 1} = 336 + \frac{336 - 256}{15} = 336 + \frac{18}{15}$$

$$= 336 + 5.\overline{33}$$

$$= 341.\overline{33}$$

Romberg $O(h^6)$ estimate:

$$R_{2,2} = 341.\overline{3333} = \frac{1024}{3}$$

Part 2: Three-point Gauss quadrature.

a) Change of variable

Let $x = 2t + 2, \quad dx = 2dt$

so
$$\int_0^4 f(x) dx = \int_{-1}^1 f(2t+2) 2dt$$

b) nodes and weights

i	t_i	w_i	$x_i = 2t_i + 2$
1	0	8/9	2
2	$-\sqrt{\frac{3}{5}}$	5/9	$2 - 2\sqrt{\frac{3}{5}}$
3	$+\sqrt{\frac{3}{5}}$	5/9	$2 + 2\sqrt{\frac{3}{5}}$

Numerically $\sqrt{3/5} \approx 0.7746$ so

$$x_2 \approx 2 - 1.54919 = 0.4508067$$

$$x_3 \approx 2 + 1.54919 = 3.5491933$$

c) Evaluate f at x_i :

$$f(2) = 2^3(16 - 4) = 8 \cdot 12 = 96$$

$$f(0.4508) \approx (0.4508)^3(16 - (0.4508)^2) \approx 1.447236$$

$$f(3.5492) \approx (3.5492)^3(16 - (3.5492)^2) \approx 152.152764$$

d)

$$I \approx 2 \sum_{i=1}^3 w_i f(x_i)$$

$$= 2 \left(\frac{8}{9} \cdot 96 + \frac{5}{9} (f(0.4508) + f(3.5492)) \right)$$

$$= 2 \left(\frac{768}{9} + \frac{5}{9} (1.447236 + 152.152764) \right)$$

$$= 2 \left(85.3333 + \frac{5}{9} \cdot 153.60000 \right)$$

$$= 2 (85.3333 + 85.33333)$$

$$= 341.33333$$

$$= \frac{1024}{3}$$

Both methods we estimate:

→ Romberg ($O(n^6)$) : $R_{2,2} = 341.3333333$

→ Three-Point Gauss : 341.3333333