

TRANSCENDENTAL EQUATIONS

1. BISECTION METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTARTE BISECTION METHOD

PROGRAM:

```

import math

# Evaluate the user-defined function safely

def f(x, func_str):

    try:

        return eval(func_str, {"x": x, "math": math, "_builtins_": None})

    except Exception as e:

        print("Error evaluating function:", e)

        return None


# Bisection Method

def bisection(func_str, a, b, tol):

    if f(a, func_str) * f(b, func_str) >= 0:

        print("Invalid interval. f(a) and f(b) must have opposite signs.")

        return

    print("Iter\t a\t b\t Xr\t f(Xr)")

    iter = 1

    while (b - a) / 2 > tol:

        Xr = (a + b) / 2

        fx = f(Xr, func_str)

        print(f"\t{iter}\t{a:.3f}\t{b:.3f}\t{Xr:.3f}\t{fx:.3f}")

        if abs(fx) < tol:

            break

        if f(a, func_str) * fx < 0:

            b = Xr

        else:
    
```

```

a = Xr

iter += 1

print(f"\nApproximate root = {Xr:.3f} (correct to 3 decimal places)")

# === Main Program ===

print("== Bisection Method ==")

func_str = input("Enter the function f(x): ") # Example: x**3 - 4*x + 1
a = float(input("Enter the starting value a: ")) # Example: 0
b = float(input("Enter the ending value b: ")) # Example: 1
tol = 0.00003 # 3 decimal place accuracy

bisection(func_str, a, b, tol)

```

OUTPUT:

```

== Bisection Method ==
Enter the function f(x): x*x*x -4*x +1
Enter the starting value a: 1
Enter the ending value b: 2

Iter      a          b          Xr        f(x)
 1      1.000    2.000    1.500    -1.625
 2      1.500    2.000    1.750    -0.641
 3      1.750    2.000    1.875     0.092
 4      1.750    1.875    1.812    -0.296
 5      1.812    1.875    1.844    -0.107
 6      1.844    1.875    1.859    -0.009
 7      1.859    1.875    1.867     0.041
 8      1.859    1.867    1.863     0.016
 9      1.859    1.863    1.861     0.003
10      1.859    1.861    1.860    -0.003
11      1.860    1.861    1.861     0.000
12      1.860    1.861    1.861    -0.001
13      1.861    1.861    1.861    -0.001
14      1.861    1.861    1.861    -0.000
15      1.861    1.861    1.861     0.000

Approximate root = 1.861 (correct to 3 decimal places)

```

CONCLUSION: The above program has been executed successfully.

2. REGULAR FALSI

Q. WRITE A PROGRAM IN PYTHON TO DEMONSTRATE REGULAR FALSI METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON RAPHSON METHOD

PROGRAM:

```
import math

# Evaluate the user-defined function safely

def safe_eval(expr, x):

    try:

        return eval(expr.strip(), {"x": x, "math": math, "m": math, "_builtins_": None})

    except (NameError, TypeError, ZeroDivisionError, SyntaxError) as e:

        print(f"Error evaluating function: {e}")

        return None


def Regula_Falsi(Func_str, a, b, tol):

    Fa = safe_eval(Func_str, a)

    Fb = safe_eval(Func_str, b)

    if Fa is None or Fb is None:

        return None

    if Fa * Fb >= 0:

        print("Invalid interval. F(a) and F(b) must have opposite signs.")

        return None

    print("\nIter.\t a\t b\t F(a)\t F(b)\t Xr\t F(Xr)")

    Xr_old = a # Initial guess to calculate error if needed

    for i in range(1, 101):

        # Regula Falsi Formula
```

```

Xr = (a * Fb - b * Fa) / (Fb - Fa)

FXr = safe_eval(Func_str, Xr)

print(f"{{i:<6}|t {a:.4f}|t {b:.4f}|t {Fa:.4f}|t {Fb:.4f}|t {Xr:.4f}|t {FXr:.4f}}")

if abs(FXr) < tol:
    return Xr

if Fa * FXr < 0:
    b = Xr
    Fb = FXr
else:
    a = Xr
    Fa = FXr

print(f"\nRoot not found within 100 iterations (Current error: {abs(FXr):.6f})")
return Xr

print("## Regula Falsi Method ##")

# Example: "x*x - 4*x - 4"
# Example: "m.cos(x) - x"
# Example: "x**3 - x - 1"
# Example: "x*x*x - 4*x - 4"

Func_str = input("Enter the function f(x): ")
a = float(input("Enter the starting value a: "))
b = float(input("Enter the starting value b: "))
tol = float(input("Enter the tolerance value: "))
root = Regula_Falsi(Func_str, a, b, tol)

if root is not None:
    print(f"\nApproximate root = {root:.3f} (correct to 3 decimal places)")

```

OUTPUT:

```
==== Regula Falsi Method ====
Enter the function f(x): x*x*x -4*x +1
Enter the starting value a: 1
Enter the ending value b: 2

Iter      a          b          f(a)      f(b)      Xr        f(Xr)
 1      1.0000    2.0000    -2.0000    1.0000    1.6667    -1.0370
 2      1.6667    2.0000    -1.0370    1.0000    1.8364    -0.1528
 3      1.8364    2.0000    -0.1528    1.0000    1.8581    -0.0175
 4      1.8581    2.0000    -0.0175    1.0000    1.8605    -0.0020
 5      1.8605    2.0000    -0.0020    1.0000    1.8608    -0.0002
 6      1.8608    2.0000    -0.0002    1.0000    1.8608    -0.0000

Approximate root = 1.8608 (correct to 3 decimal places)
```

CONCLUSION: The above program has been executed successfully.

18001

3. NEWTON'S RAPHSON METHOD

Q. WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON RAPHSON METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON RAPHSON METHOD

PROGRAM:

```
Import math

# Safely evaluate the user-defined function

def safe_eval(expr, x):

    try:

        return eval(expr.strip(), {"x": x, "math": math, "__builtins__": None})

    except (NameError, TypeError, ZeroDivisionError, SyntaxError) as e:

        print(f"Error evaluating function: {e}")

    return None
```

```
# Safely evaluate the derivative of the function

def df(x, deriv_str):
    try:
        return eval(deriv_str.strip(), {"x": x, "math": math, "__builtins__": None})
    except Exception as e:
        print(f"Error evaluating derivative: {e}")
    return None

def Newton_Raphson_Method(func_str, deriv_str, x0, tol, max_iter=100):
    ai = x0

    print("\nIter.\tai\tf(ai)\t\tdf(ai)\tai+1")
    for i in range(1, max_iter + 1):
        fai = safe_eval(func_str, ai)
        dfai = df(ai, deriv_str)
```

```

if dfai == 0:
    print("Derivative is zero. Method fails.")
    return None

# Newton-Raphson Formula
ai_p1 = ai - fai / dfai

print(f"\n{i:<6}\t{ai:.4f}\t{fai:.4f}\t{dfai:.4f}\t{ai_p1:.4f}")

if abs(ai_p1 - ai) < tol:
    print(f"\nApproximate root = {ai_p1:.3f} (correct to 3 decimal places)")
    return ai_p1

ai = ai_p1

print("\nMaximum iterations reached without convergence.")
return ai_p1

```

import math

```

print("## Newton-Raphson Method ##")

# Example 1: "x*x - 4*x - 4"
# Example 2: "m.cos(x) - x"
# Example 3: "x**3 - x - 1"
# Example of derivative: "3*x*x - 1" for f(x)=x**3 - x - 1

func_str = input("Enter the function f(x): ")
deriv_str = input("Enter the derivative df(x): ")
x0 = float(input("Enter the initial guess x0: "))
tol = float(input("Enter the tolerance for X decimal place accuracy: "))
newton_raphson(func_str, deriv_str, x0, tol)

```

OUTPUT:

```
==== Newton-Raphson Method ====
Enter the function f(x): x*x*x -2*x -5
Enter the derivative f'(x): 3*x*x -2
Enter the initial guess x0: 2

===== Newton-Raphson Iteration Table =====
Iter  x0          f(x0)        f'(x0)       x1
-----
1    2.000000    -1.000000    10.000000   2.100000
2    2.100000     0.061000    11.230000   2.094568
3    2.094568     0.000186    11.161647   2.094551
=====
Approximate root = 2.0946 (correct to 3 decimal places)
```

CONCLUSION:

The above program has been executed successfully.

18001

INTERPOLATION

Q. WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON FORWARD INTERPOLATION.

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON FORWARD INTERPOLATION.

PROGRAM:

```
def forward_difference_table(x, y):
    n = len(y)
    diff_table = [y.copy()] # First row is just y values

    # Generate the forward difference table
    for i in range(1, n):
        row = []
        for j in range(n - i):
            # Calculate the i-th difference: diff(j) = diff(j+1) - diff(j)
            value = diff_table[i-1][j+1] - diff_table[i-1][j]
            row.append(value)
        diff_table.append(row)

    return diff_table

def display_table(x, diff_table):
    n = len(x)
    print("\nForward Difference Table:")
    header = "i\t x\t t\t y" + "\t\t Δy" * (n - 1)
    print(header)
    print("-" * len(header) * 2) # For visual separation

    for i in range(n):
        row = [str(i), f"{x[i]:.2f}", f"{diff_table[0][i]:.2f}"]
        # Add the differences
        for j in range(1, n - i):
            row.append(f"{diff_table[j][i]:.2f}")
        print("\t".join(row))
```

```

row.append(f"\t{diff_table[j][i]:.2f}")
print("\t".join(row))

def main():
    n = int(input("Enter the number of data points: "))
    x = []
    y = []
    print("Enter x values (equally spaced):")
    for i in range(n):
        x.append(float(input(f"x[{i}] = ")))
    print("Enter corresponding y values:")
    for i in range(n):
        y.append(float(input(f"y[{i}] = ")))
    # Check equal spacing
    h_values = []
    for i in range(n - 1):
        h_values.append(x[i+1] - x[i])
    # Check if all h values are approximately equal
    if not all(abs(h_values[i] - h_values[0]) < 1e-5 for i in range(n - 1)):
        print("\nError: X values are not equally spaced.")
    return
    diff_table = forward_difference_table(x, y)
    display_table(x, diff_table)
if __name__ == "__main__":
    main()

```

OUTPUT:

Forward Difference Table:									
x	Δ^0y	Δ^1y	Δ^2y	Δ^3y	Δ^4y	Δ^5y	Δ^6y	Δ^7y	Δ^8y
-1.00	-13.00	6.00	0.00	6.00	0.00	0.00	0.00	0.00	0.00
0.00	-7.00	6.00	6.00	6.00	0.00	0.00	0.00	0.00	
1.00	-1.00	12.00	12.00	6.00	0.00	0.00	0.00		
2.00	11.00	24.00	18.00	6.00	0.00	0.00			
3.00	35.00	42.00	24.00	6.00	0.00				
4.00	77.00	66.00	30.00	6.00					
5.00	143.00	96.00	36.00						
6.00	239.00	132.00							
7.00	371.00								

CONCLUSION:

THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY

Q. WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON BACKWARD INTERPOLATION.

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE NEWTON BACKWARD INTERPOLATION.

PROGRAM:

```
import math
```

x = [0, 30, 60, 90]

$$\mathbf{y} = [1, 0.85, 0.5, 0]$$

$$xp = 70$$

$$h = x[1] - x[0]$$

$$p = (xp - x[-1]) / h$$

$$dy1_3 = y[3] - y[2]$$

$$dy1_2 = y[2] - y[1]$$

$$dy1_1 = y[1] - y[0]$$

$$d2y2 = dy1_3 - dy1_2$$

$$d2y1 = dy1_2 - dy1_1$$

$$d3y1 = d2y2 - d2y1$$

```

yp = (y[-1]
    + p * dy1_3
    + (p * (p + 1) / math.factorial(2)) * d2y2
    + (p * (p + 1) * (p + 2) / math.factorial(3)) * d3y1)

```

```
print(f"\nEstimated cos(70°) using Backward formula = {yp:.5f}")
```

OUTPUT:

x	y	∇y	$\nabla^2 y$	$\nabla^3 y$
0	1			
30	0.85	-0.1500		
60	0.5	-0.3500	-0.2000	
90	0	-0.5000	-0.1500	0.0500

Estimated cos(70°) using Backward formula = 0.34753

CONCLUSION:

THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

CURVE FITTING

1. STRAIGHT LINE

PROGRAM:

```
# Curve_fit_no_numpy.py
# Least-squares straight-line fit (no numpy, no pandas)

from typing import List, Optional, Tuple

def fit_line(x_values: List[float], y_values: List[float]) -> Tuple[float, float]:
    """Return (a0, a1) for best fit line y = a0 + a1*x using least squares."""
    if len(x_values) != len(y_values) or len(x_values) == 0:
        raise ValueError("x_values and y_values must have same non-zero length.")
    n = len(x_values)
    sum_x = sum(x_values)
    sum_y = sum(y_values)
    sum_x2 = sum(x * x for x in x_values)
    sum_xy = sum(x * y for x, y in zip(x_values, y_values))

    denom = n * sum_x2 - sum_x * sum_x
    if abs(denom) < 1e-12:
        raise ValueError("Denominator nearly zero: can't compute unique fit (collinear x?).")

    a1 = (n * sum_xy - sum_x * sum_y) / denom
    a0 = (sum_y - a1 * sum_x) / n
    return a0, a1

def print_table(x_values: List[float], y_values: List[float]) -> None:
    """Print table of x, y, x^2, x*y and the sums."""
    n = len(x_values)
    rows = []
    for x, y in zip(x_values, y_values):
        rows.append((x, y, x*x, x*y))

    # Column widths
    w = [8, 8, 10, 10]
    header = f"{'i':<{w[0]}} {'y':<{w[1]}} {'x^2':<{w[2]}} {'x*y':<{w[3]}}"
    print(header)
    print("*" * (sum(w) + 3))

    for r in rows:
        print(f"{r[0]:<{w[0]}.4g} {r[1]:<{w[1]}.4g} {r[2]:<{w[2]}.4g} {r[3]:<{w[3]}.4g}")



```

```

sum_x = sum(r[0] for r in rows)
sum_y = sum(r[1] for r in rows)
sum_x2 = sum(r[2] for r in rows)
sum_xy = sum(r[3] for r in rows)

# print the sums
print("-" * (sum(w) + 3))
print(f"'SUM':<{w[0]}>{sum_y}>{w[1]}.4g {sum_x2:>{w[2]}.4g} {sum_xy:>{w[3]}.4g}'")
print()

# The image shows extra print statements for the sums:
print(f"'SUM_X':<{w[0]}>{sum_y}>{w[1]}.4g {sum_x2:>{w[2]}.4g} {sum_xy:>{w[3]}.4g}'")
print() # extra line break from image 1000040410.jpg

print(f"\Sigma x = {sum_x:.4g}, \Sigma y = {sum_y:.4g}, \Sigma x^2 = {sum_x2:.4g}, \Sigma xy = {sum_xy:.4g}'")
print()

def predict(a0: float, a1: float, x: float) -> float:
    return a0 + a1 * x
def interactive():
    print("Curve fitting (straight line) - enter data points.")
    n = int(input("How many points? "))
    x_values = []
    y_values = []

    for i in range(n):
        raw = input(f"Point {i+1} as 'x y' (e.g. 2 5): ").strip().split()
        if len(raw) < 2:
            print("Invalid input, try again.")
            return
        x_values.append(float(raw[0]))
        y_values.append(float(raw[1]))

    print()
    print_table(x_values, y_values)
    a0, a1 = fit_line(x_values, y_values)

    print(f"Best fit line: y = ({a0:.6f}) + ({a1:.6f}) x")

    choice = input("Predict y for some x? (y/n): ").strip().lower()
    if choice and choice[0] == 'y':

```

```

xv = float(input("Enter x: "))

print(f"Predicted y = {predict(a0, a1, xv):.6f}")
if __name__ == "__main__":
    # Example usage (change values directly if you prefer):
    x_values = [0, 2, 5, 7]
    y_values = [-1, 5, 12, 20]

    # Print table and compute
    print_table(x_values, y_values)
    a0, a1 = fit_line(x_values, y_values)

    print(f"Best fit line: y = ({a0:.6f}) + ({a1:.6f}) x")
    print(f"For x=0, predicted y = {predict(a0, a1, 0):.6f}")

```

OUTPUT:

x	y	x^2	$x*y$
0	-1	0	0
2	5	4	10
5	12	25	60
7	20	49	140
Σ	36	78	210

$\Sigma x = 14$, $\Sigma y = 36$, $\Sigma x^2 = 78$, $\Sigma xy = 210$

Best fit line: $y = -1.137931 + 2.896552 x$
For $x=8$, predicted $y = 22.034483$

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

2. DEGREE POLYNOMIAL

PROGRAM:

```
# quad_fit_no_numpy.py
# Fit quadratic y = a0 + a1*x + a2*x^2 using normal equations (no numpy, no pandas)
```

```
from typing import List, Tuple
```

```
def build_sums(x_values: List[float], y_values: List[float]) -> dict:
```

```
    """Calculates the necessary sums for the normal equations."""
    s = {
```

```
        'n': 0.0,
        'sx': 0.0,
        'sx2': 0.0,
        'sx3': 0.0,
        'sx4': 0.0,
        'sy': 0.0,
        'sxy': 0.0,
        'sx2y': 0.0
    }
```

```
for x, y in zip(x_values, y_values):
```

```
    s['n'] += 1
    s['sx'] += x
    s['sx2'] += x**2
    s['sx3'] += x**3
    s['sx4'] += x**4
    s['sy'] += y
    s['sxy'] += x * y
```

```

s['sx2y'] += (x**2) * y
return s

def print_table_and_sums(x_values: List[float], y_values: List[float]) -> None:
    """Prints the data points and the calculated sums in a formatted table."""

# Header
print(f"{'x':>8}{'y':>10}{'x^2':>12}{'x^3':>12}{'x^4':>12}{'x*y':>12}{'x^2*y':>12}")
print("-" * 78) # Separator

# Data rows
for x, y in zip(x_values, y_values):
    print(f"{'x':8.4g}{{'y':10.4g}{'x^2':12.4g}{'x3':12.4g}{'x4':12.4g}{'x*y':12.4g}{{'x^2*y':12.4g}}")

# Sums
s = build_sums(x_values, y_values)
print("-" * 78) # Separator
print(f"n = {s['n']:3.4g}, sx = {s['sx']:12.4g}, sx2 = {s['sx2']:12.4g}, sx3 = {s['sx3']:12.4g}, sx4 = {s['sx4']:12.4g}")
print(f"sy = {s['sy']:12.4g}, sxy = {s['sxy']:12.4g}, sx2y = {s['sx2y']:12.4g}")
print()

def solve_3x3(A: List[List[float]], b: List[float]) -> List[float]:
    """
    Simple Gaussian elimination (in-place) to solve Ax = b for a 3x3 A.
    Returns the solution vector x.
    """

# Make copies

```

```

M = [row[:] for row in A]
rhs = b[:]
n = 3

# Forward elimination
for k in range(n):
    # find pivot
    pivot = M[k][k]

    # Check for singularity/pivot too small (1e-14 is a common threshold)
    if abs(pivot) < 1e-14:
        # try to swap with a lower row
        for i in range(k + 1, n):
            if abs(M[i][k]) > 1e-14:
                M[k], M[i] = M[i], M[k]
                rhs[k], rhs[i] = rhs[i], rhs[k]
                pivot = M[k][k]
                break

    if abs(pivot) < 1e-14:
        raise ValueError("Singular matrix in solve_3x3")

    # normalize row k
    for j in range(k, n):
        M[k][j] /= pivot
        rhs[k] /= pivot

    # eliminate
    for i in range(k + 1, n):
        for j in range(k, n):
            M[i][j] -= M[i][k] * M[k][j]
        rhs[i] -= rhs[k] * M[i][k]

```

```

factor = M[i][k]

for j in range(k, n):
    M[i][j] -= factor * M[k][j]
    rhs[i] -= factor * rhs[k]

# Back substitution
x = [0.0] * n

for i in range(n - 1, -1, -1):
    val = rhs[i]
    for j in range(i + 1, n):
        val -= M[i][j] * x[j]

    # The diagonal element M[i][i] should be 1.0 from normalization,
    # but we check for singularity one last time just in case.
    x[i] = val / M[i][i] if abs(M[i][i]) > 1e-14 else val

return x

```

```

def fit_quadratic(x_values: List[float], y_values: List[float]) -> Tuple[float, float, float]:
    """Calculates the coefficients (a0, a1, a2) for the least-squares quadratic fit."""

    if len(x_values) != len(y_values) or len(x_values) == 0:
        raise ValueError("X-values and Y-values must have same non-zero length.")

    s = build_sums(x_values, y_values)

    # Normal equations matrix for [a0, a1, a2]
    # [ n  Σx  Σx^2 ] [a0] = [ Σy ]
    # [ Σx  Σx^2 Σx^3 ] [a1] = [ Σxy ]

```

```

# [ Σx ^2 Σx^3 Σx^4 ] [a2] = [ Σxy ]
A = [
    [s['n'], s['sx'], s['sx2']],
    [s['sx'], s['sx2'], s['sx3']],
    [s['sx2'], s['sx3'], s['sx4']]]
]

b = [s['sy'], s['sxy'], s['sx2y']]

# Solve for a0, a1, a2
a0, a1, a2 = solve_3x3(A, b)

return a0, a1, a2

def predict(a0: float, a1: float, a2: float, x: float) -> float:
    """Calculates the predicted y value for a given x using the fitted quadratic."""
    return a0 + a1*x + a2*(x**2)

if __name__ == "__main__":
    # Example points from your notebook: (0, 1), (1, 6), (2, 17)
    x_values = [0.0, 1.0, 2.0]
    y_values = [1.0, 6.0, 17.0]

    print("### Input Data and Sums ###")
    print_table_and_sums(x_values, y_values)

    # Fit quadratic
    a0, a1, a2 = fit_quadratic(x_values, y_values)

```

```

print("### Fitting Results ###")

print(f"Fitted quadratic: y = {a0:.6f} + {a1:.6f} x + {a2:.6f} x^2")

# Predictions requested in the notebook

print("\n### Predictions ###")

# Prediction for x=1.6

print(f"y(1.6) = {predict(a0, a1, a2, 1.6):.6f}")

# Prediction for x=3.0

print(f"y(3) = {predict(a0, a1, a2, 3.0):.6f}")

```

OUTPUT:

x	y	x^2	x^3	x^4	x*y	x^2*y
0	1	0	0	0	0	0
1	6	1	1	1	6	6
2	17	4	8	16	34	68
Σ	24	5	9	17	40	74

$\Sigma x = 3.0$, $\Sigma y = 24.0$, $\Sigma x^2 = 5.0$, $\Sigma x^3 = 9.0$, $\Sigma x^4 = 17.0$
 $\Sigma (xy) = 40.0$, $\Sigma (x^2 y) = 74.0$

Fitted quadratic: $y = 1.000000 + 2.000000 x + 3.000000 x^2$
 $y(1.6) = 11.880000$
 $y(3) = 34.000000$

CONCLUSION:

THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

SOLUTION OF SIMULTANEOUS ALGEBRAIC EQUATIONS

GUASSIAN ELIMINATION METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE GUASSIAN ELIMINATION METHOD

PROGRAM:

```
#Gaussian elimination with partial pivoting row operations
```

```
# Solve the system:
```

```
# x1 + 10 x2 - x3 = 3
```

```
# 2x1 + 3 x2 + 20 x3 = 7
```

```
# 10x1 - x2 + 2 x3 = 4
```

```
# Augmented matrix (each row: [a11, a12, a13, b])
```

```
A = [
    [1.0, 10.0, -1.0, 3.0],
    [2.0, 3.0, 20.0, 7.0],
    [10.0, -1.0, 2.0, 4.0]
```

```
]
```

```
n = len(A) # Number of equations/variables (n=3)
```

```
def print_matrix(M: List[List[float]], msg=None) -> None:
```

```
    """Prints the augmented matrix with 10.6f formatting."""

```

```
    if msg:
```

```
        print(msg)
```

```
    for r in M:
```

```
        # Join values with spaces, formatting each to 10 characters with 6 decimal places
```

```
        print("[" + " ".join(f"{val:10.6f}" for val in r) + "]")
```

```
    print()
```

```
def swap_rows(M: List[List[float]], i: int, j: int) -> None:
    """Swaps row i and row j in matrix M and prints the operation."""
    M[i], M[j] = M[j], M[i]
    # Print R(i+1) <-> R(j+1) to use 1-based indexing for output
    print(f"R({i+1}) <-> R({j+1})")
    print_matrix(M)
```

```
def scale_and_add(M: List[List[float]], col: int, factor: float, row: int) -> None:
```

```
"""
```

Performs $R_{dest} = R_{dest} - k * R_{src}$.

In the context of elimination, row is dest (row to eliminate in), col is src.

```
"""
```

```
n_cols = len(M[0])
```

```
# Print R(dest+1) <- R(dest+1) - (k) * R(src+1) to use 1-based indexing
```

```
print(f"R({row+1}) <- R({row+1}) - ({factor:.6f})*R({col+1})")
```

```
# Perform the operation: M[row] = M[row] - factor * M[col]
```

```
for c in range(n_cols):
```

```
    M[row][c] = M[row][c] - factor * M[col][c]
```

```
print_matrix(M)
```

```
# --- Main Solution Logic ---
```

```
# Work on a copy of the augmented matrix
```

```
M = deepcopy(A)
```

```
print_matrix(M, "Initial augmented matrix [A | b]:")
```

```
for col in range(n):
```

```
    # Partial pivot: find row with max abs value in column 'col' from rows col..n-1
```

```
    # max() returns the row index 'r'
```

```
    pivot_row = max(range(col, n), key=lambda r: abs(M[r][col]))
```

```
    if pivot_row != col:
```

```
        swap_rows(M, pivot_row, col)
```

```
    pivot = M[col][col]
```

```
    if abs(pivot) < 1e-12: # Check for near-zero pivot (singularity)
```

```
        raise ValueError("Zero pivot encountered")
```

```
    # Eliminate below
```

```
    for row in range(col + 1, n):
```

```
        factor = M[row][col] / pivot
```

```
        # scale_and_add(Matrix, source_row, factor, destination_row)
```

```
        scale_and_add(M, col, factor, row)
```

```
    print("Upper-triangular matrix after forward elimination:")
```

```
    print_matrix(M)
```

```
# Back substitution
```

```
x = [0.0] * n # Solution vector [x1, x2, x3]
```

```
# Loop backward from the last row (n-1) to the first row (0)
```

```
for i in range(n - 1, -1, -1):
```

```
# s is the RHS (augmented column), which is M[i][n]
```

```
s = M[i][n]
```

```
# Subtract known x[j]'s multiplied by their coefficients M[i][j]
```

```
for j in range(i + 1, n)
```

```
    s -= M[i][j] * x[j]
```

```

# Solve for x[i]
x[i] = s / M[i][i]

# Print Solution vector
print("Solution vector:")

# Enumerate x starting from 1 for x1, x2, x3 display

for i, xi in enumerate(x, 1):
    print(f"x{i} = {xi:.8f}")

```

OUTPUT:

```

Initial augmented matrix [A | b]:
[ 1.000000 10.000000 -1.000000 3.000000]
[ 2.000000 3.000000 20.000000 7.000000]
[ 10.000000 -1.000000 2.000000 4.000000]

R3 <-> R1
[ 10.000000 -1.000000 2.000000 4.000000]
[ 2.000000 3.000000 20.000000 7.000000]
[ 1.000000 10.000000 -1.000000 3.000000]

R2 = R2 - (0.200000)*R1
[ 10.000000 -1.000000 2.000000 4.000000]
[ 0.000000 3.200000 19.600000 6.200000]
[ 1.000000 10.000000 -1.000000 3.000000]

R3 = R3 - (0.100000)*R1
[ 10.000000 -1.000000 2.000000 4.000000]
[ 0.000000 3.200000 19.600000 6.200000]
[ 0.000000 10.100000 -1.200000 2.600000]

R3 <-> R2
[ 10.000000 -1.000000 2.000000 4.000000]
[ 0.000000 10.100000 -1.200000 2.600000]
[ 0.000000 3.200000 19.600000 6.200000]

R3 = R3 - (0.316832)*R2
[ 10.000000 -1.000000 2.000000 4.000000]
[ 0.000000 10.100000 -1.200000 2.600000]
[ 0.000000 0.000000 19.980198 5.376238]

Upper-triangular matrix after forward elimination:
[ 10.000000 -1.000000 2.000000 4.000000]
[ 0.000000 10.100000 -1.200000 2.600000]
[ 0.000000 0.000000 19.980198 5.376238]

Solution vector:
x1 = 0.37512389
x2 = 0.28939544
x3 = 0.26907830

```

CONCLUSION:

THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

NUMERICAL SOLUTIONS OF FIRST AND SECOND ORDER DIFFERENTIAL EQUATIONS

1. TAYLOR SERIES

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE TAYLOR SERIES

PROGRAM:

```
# Taylor_ode_no_sympy.py

# Compute y(n0 + h) using Taylor series for ODE dy/dn = n - y^2, y(n0)=y0

# No external libraries beyond 'math'
```

```
from math import factorial
```

```
def compute_derivatives_at(n0: float, y0: float) -> List[float]:
```

```
    """
```

Compute derivatives y', y'', y''', y^(4), y^(5) at (n0, y0)

using formulas obtained by differentiating dy/dn = n - y^2.

Returns list [y0, y1, y2, y3, y4, y5], where yk is the kth derivative.

```
    """
```

```
# y^(0) = y0
```

```
y_0 = y0
```

```
# y' = n - y^2
```

```
y_1 = n0 - (y_0 ** 2)
```

```
# Second derivative: y'' = d/dn(n - y^2) = 1 - 2*y*(dy/dn) = 1 - 2*y*y'
```

```
y_2 = 1.0 - 2.0 * y_0 * y_1
```

```
# Third derivative: y''' = d/dn(1 - 2*y*y') = 0 - 2 * [ y'*y' + y*y'' ]
```

```
# y''' = -2*y'^2 - 2*y*y''
```

```
y_3 = -2.0 * (y_1 ** 2) - 2.0 * y_0 * y_2
```

```
# Fourth derivative: y^(4) = d/dn(-2*y'^2 - 2*y*y'')
```

```

#  $y^4 = -2*(2*y'y'') - 2[y'*y'' + y*y''']$ 
#  $y^4 = -4*y'*y'' - 2*y'*y'' - 2*y*y''' = -6*y'*y'' - 2*y*y'''$ 
y_4 = -2.0 * y_0 * y_3 - 6.0 * y_1 * y_2

# Fifth derivative:  $y^5 = d/dn(-6*y'*y'' - 2*y*y''')$ 
#  $y^5 = -6*[y''y'' + y'*y'''] - 2[y'*y'''' + y*y^4]$ 
#  $y^5 = -6*y''^2 - 6*y'*y''' - 2*y''*y''' - 2*y*y^4$ 
#  $y^5 = -2*y*y^4 - 8*y'*y''' - 6*y''^2$ 
y_5 = -2.0 * y_0 * y_4 - 8.0 * y_1 * y_3 - 6.0 * (y_2 ** 2)

return [y_0, y_1, y_2, y_3, y_4, y_5]

def taylor_at(n0: float, y0: float, h: float, order: int = 5) -> Tuple[float, List[float]]:
    """
    Evaluate Taylor polynomial of given order (<=5) for y at n0+h.
    Returns (approx_value, derivatives_list).
    """
    if order > 5:
        raise ValueError("This implementation supports up to 5th derivative (order<=5).")

    derivs = compute_derivatives_at(n0, y0)

    # Build Taylor sum:  $y(n0+h)$  approx  $y(n0) + h*y'(n0)/1! + h^2*y''(n0)/2! + ...$ 

    taylor_sum = 0.0
    for k in range(order + 1):
        # Term =  $y^k * h^k / k!$ 
        taylor_sum += derivs[k] * (h ** k) / factorial(k)
    return taylor_sum, derivs

```

```

if __name__ == "__main__":
    # Initial point and step
    n0 = 0.0
    y0 = 1.0
    h = 0.1
    order = 5 # use terms up to y^(5)/5!
    # Compute the approximation and the derivatives
    approx, derivs = taylor_at(n0, y0, h, order)
    print("Derivatives at n0 = {:.4g}, y0 = {:.4g}: ".format(n0, y0))
    print(f"y'(0) = {derivs[1]:.6g}")
    print(f"y''(0) = {derivs[2]:.6g}")
    print(f"y'''(0) = {derivs[3]:.6g}")
    print(f"y^(4)(0) = {derivs[4]:.6g}")
    print(f"y^(5)(0) = {derivs[5]:.6g}")
    print()
    print("Taylor approximation up to order {}:".format(order))
    print(f"y({n0 + h:.4g}) ≈ {approx:.10f}")
    print(f"Rounded to 4 decimal places: {approx:.4f}")

```

OUTPUT:

```

=====
Derivatives at n0 = 0, y0 = 1:
y(0)      = 1
y'(0)     = -1
y''(0)    = 3
y'''(0)   = -8
y^(4)(0)= 34
y^(5)(0)= -186

Taylor approximation up to order 5:
y(0.1) ≈ 0.9137928333
Rounded to 4 decimal places: 0.9138

```

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

2. EULER'S METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTARTE EULER'S METHOD.

PROGRAM:

```
# Euler_method.py
# Solve dy/dx = -y with y(0)=1 using Euler's method
```

```
def f(x, y):
    """The ODE: dy/dx = -y"""
    return -y
```

```
def euler(x0, y0, h, x_target):
    """Euler's method to approximate y(x_target)"""
    steps = int((x_target - x0) / h)
```

```
x = x0
y = y0
```

```
print("Step | x | y ")
print("----|----|----")
print(f" 0 | {x0:.2f} | {y0:.6f}")
```

```
for i in range(1, steps + 1):
    # Euler formula: y(i+1) = y(i) + h * f(x(i), y(i))
    y = y + h * f(x, y)
    x = x + h
    print(f"{i:3d} | {x:.2f} | {y:.6f}")
return y
```

```
if __name__ == "__main__":
    # initial values
    x0 = 0.0
    y0 = 1.0
    h = 0.01
    x_target = 0.04

    result = euler(x0, y0, h, x_target)

    print(f"\nApproximate value at x={x_target:.2f}:", round(result, 6))
```

OUTPUT:

Step	x	y
0	0.00	1.000000
1	0.01	0.990000
2	0.02	0.980100
3	0.03	0.970299
4	0.04	0.960596

Approximate value at x=0.04: 0.960596

CONCLUSION: THE ABOVE PRAOGRAM HAS BEEN EXECUTED SIUCESSFULLY.

3.MODIFIED EULER'S METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE MODIFIED EULER'S METHOD

PROGRAM:

```
# Modified_Euler's_Method.py

# Equation: dy/dx = x^2 + y, y(0) = 1

# Find y(0.2) with step size h = 0.02

def f(x, y):
    """The ODE: dy/dx = x^2 + y"""
    return x**2 + y

# Initial conditions
x0 = 0
y0 = 1
h = 0.02
x_end = 0.2
n = int((x_end - x0) / h) # Number of steps: (0.2 - 0) / 0.02 = 10

# Table header
print("-----")
print("i | x(i) | y(i) | f(x(i),y(i)) (f1) | y'(Pred) | f(x+h, y') (f2) | y(i+1)")
print("-----")

# Print initial condition (Step 0)
print(f'{0:<2d} | {x0:10.6f} | {y0:10.6f} | {"':20} | {"':10} | {"':20} | {"':10}"')
# Iterative Modified Euler Calculation

# 1. Predictor (Standard Euler): y* = y_i + h * f(x_i, y_i)
f1 = f(x0, y0)
y_pred = y0 + h * f1
```

```

# 2. Corrector (Heun's Formula):  $y_{i+1} = y_i + (h / 2) * [ f(x_i, y_i) + f(x_{i+1}, y^*) ]$ 

x_next = x0 + h
f2 = f(x_next, y_pred)
y_next = y0 + (h / 2) * (f1 + f2)

# Print intermediate results for the current step (i+1)
print(f"{{i+1:<2d} | {x_next:10.6f} | {y0:10.6f} | {f1:20.6f} | {y_pred:10.6f} | {f2:20.6f} | {y_next:10.6f}"})

# Update for next iteration
x0 = x_next
y0 = y_next

print("-----")
print(f"h = {h:.2f}, Number of steps = {n}")
print(f"Formula used:")
print(f"y*(i+1) = y_i + h * f(x_i, y_i) (Euler Predictor)")
print(f"y(i+1) = y_i + (h/2) * [ f(x_i, y_i) + f(x_i + h, y*(i+1)) ] (Heun Corrector)")
print(f"Approximate value of y({x_end:.1f}) = {y0:.6f}")
print("-----")

```

OUTPUT:

i	x(i)	y(i)	f(x(i),y(i))	y* (Pred)	f(x+h, y*)	y(i+1)
0	0.0000	1.000000	1.000000	1.020000	1.020400	1.020204
1	0.0200	1.020204	1.020604	1.040616	1.042216	1.040832
2	0.0400	1.040832	1.042432	1.061681	1.065281	1.061909
3	0.0600	1.061909	1.065509	1.083220	1.089620	1.083461
4	0.0800	1.083461	1.089861	1.105258	1.115258	1.105512
5	0.1000	1.105512	1.115512	1.127822	1.142222	1.128089
6	0.1200	1.128089	1.142489	1.150939	1.170539	1.151219
7	0.1400	1.151219	1.170819	1.174636	1.200236	1.174930
8	0.1600	1.174930	1.200530	1.198941	1.231341	1.199249
9	0.1800	1.199249	1.231649	1.223882	1.263882	1.224204

h = 0.02, Number of steps = 10
 Formula used:
 $y_{(i+1)} = y_i + (h/2) * [f(x_i, y_i) + f(x_i + h, y^*)]$
 Approximate value of y(0.2) = 1.224204

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

4. RUNGE-KUTTA 4th ORDER METHOD

AIM: AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE RUNGE-KUTTA 4th ORDER METHOD

PROGRAM:

RK4 step-by-step for $y' = x + y$, $y(0)=1$

```
import math
```

```
def f(x, y):
```

```
    """The ODE:  $y' = x + y$ """

```

```
    return x + y
```

```
def exact_solution(x):
```

```
    """The exact solution of  $y' - y = x$  with  $y(0)=1$  is  $y = 2e^x - x - 1$ """

```

```
    return 2 * math.exp(x) - x - 1
```

```
# Initial values
```

```
x = 0.0
```

```
y = 1.0
```

```
h = 0.1
```

```
steps = int(0.2 / h) # Compute up to x = 0.2 (steps = 2)
```

```
print("Runge-Kutta 4th order (RK4) step-by-step")
```

```
print(f"Equation:  $y' = x + y$ ,  $y(0)={y}$ ")
```

```
print(f"Step | x_n | y_n (before) | k1 | k2 | k3 | k4 | y_{steps * h:.1f}")
```

```
print("-" * 100)
```

```
for n in range(steps):
```

```
    # RK4 coefficients
```

```
    k1 = f(x, y) * h
```

```
    k2 = f(x + h/2.0, y + (k1/2.0)) * h
```

```

k3 = f(x + h/2.0, y + (k2/2.0)) * h
k4 = f(x + h, y + k3) * h

# RK4 Update Formula
increment = (h/6.0) * (k1 + 2.0*k2 + 2.0*k3 + k4)

y_next = y + increment

# Print step details
# Using 'n' for the step count (0 and 1) and 'n+1' for the y_next index (1 and 2)
print(f"{n+1:3d} | {x:6.3f} | {y:16.10f} | "
      f"{k1:7.6f} | {k2:7.6f} | {k3:7.6f} | {k4:7.6f} | {y_next:10.9f}")

# Update
x += h
x = round(x, 10) # avoid floating accumulation
y = y_next

print("-" * 100)

# Final output
exact_y = exact_solution(x)
absolute_error = abs(y - exact_y)

print(f"Final RK4 approximation: y({{x:.3f}}) = {{y:.9f}}")
print(f"Exact value      : y({{x:.3f}}) = {{exact_y:.9f}}")
print(f"Absolute error   : |abs({{absolute_error:.12e}})|")

```

OUTPUT:

Runge-Kutta 4th order (RK4) step-by-step

Equation: $y' = x + y$, $y(0)=1$

Step	x_n	y_n (before)	k_1	k_2	k_3	k_4	$y_{(n+1)}$
1	0.000	1.0000000000	1.000000	1.100000	1.105000	1.210500	1.110341667
2	0.100	1.1103416667	1.210342	1.320859	1.326385	1.442980	1.242805142

Final RK4 approximation: $y(0.200) = 1.242805142$
Exact value : $y(0.200) = 1.242805516$
Absolute error : $3.746189507492e-07$
|

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY,

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NUMERICAL INTEGRATION

1. TRAPEZOIDAL RULE

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE TRAPEZOIDAL RULE.

PROGRAM:

```
# Trapezoidal Rule for I = integral(f(x) dx from 0 to 1) with 2 subintervals
```

```
def f(x):
```

```
    """The integrand: 1 / (1 + x^2)"""

```

```
    return 1 / (1 + x**2)
```

```
# Given limits
```

```
a = 0
```

```
b = 1
```

```
n = 2 # number of subintervals
```

```
# Step size
```

```
h = (b - a) / n
```

```
# Compute x values
```

```
x = [a + i * h for i in range(n + 1)] # [0.0, 0.5, 1.0]
```

```
# Compute f(x) values
```

```
f_values = [f(xi) for xi in x] # [1.0, 0.8, 0.5]
```

```
# Display table header
```

```
print("-----")
```

```
print("i | x(i) | f(x(i)) = 1/(1+x^2)")
```

```
print("-----")
```

```

# Display table values
for i in range(n + 1):
    print(f"\{i}<3\} | {x[i]}<8.4f} | {f_values[i]}<8.4f}")
print("-----")

# Apply Trapezoidal Rule
# The formula in Python: I = (h / 2) * (f[0] + 2 * sum(f[1:-1]) + f[-1])
I = (h / 2) * (f_values[0] + 2 * sum(f_values[1:-1]) + f_values[-1])

# Step-by-step explanation
print(f"h = (b - a) / n = ({b} - {a}) / {n} = {h}")
print("\nUsing Trapezoidal Rule:")
print(f"I = (h / 2) * [f(x0) + 2*f(x1) + f(x2)]")
print(f"I = ({h/2}) * [{f_values[0]}:.4f} + 2*{f_values[1]}:.4f} + {f_values[2]}:.4f}]")

# Final result
print("-----")
print(f"Approximate value of the integral I = {I:.4f}")
print("-----")

```

OUTPUT:

```
-----  
 i      x(i)      f(x(i)) = 1/(1+x^2)  
-----  
 0      0.0000    1.0000  
 1      0.5000    0.8000  
 2      1.0000    0.5000  
-----
```

$h = (b - a) / n = (1 - 0) / 2 = 0.5$

Using Trapezoidal Rule:

```
I = (h/2) * [f(x0) + 2*f(x1) + f(x2)]  
I = (0.5/2) * [1.0000 + 2*0.8000 + 0.5000]
```

Approximate value of the integral I = 0.7750

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

2. SIMPSON'S 1/3 RULE

AIM: AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE SIMPSON'S 1/3 RULE.

PROGRAM:

```
# Simpson's 1/3 Rule for I = ∫(0 to 1) e^(-x^2) dx with n = 4
```

```
import math
```

```
# Define the function
```

```
def f(x):
```

```
    return math.exp(-x**2)
```

```
# Given values
```

```
a = 0 # lower limit
```

```
b = 1 # upper limit
```

```
n = 4 # number of subintervals (must be even)
```

```
# Step size
```

```
h = (b - a) / n
```

```
# Generate x and f(x) values
```

```
x = [a + i * h for i in range(n + 1)]
```

```
f_values = [f(xi) for xi in x]
```

```
# Display table
```

```
print("-----")
```

```
print(" i | x(i) | f(x(i)) = e^(-x^2)")
```

```
print("-----")
```

```
for i in range(n + 1):
```

```
    print(f"{i:<3} | {x[i]:<8.4f} | {f_values[i]:<10.6f}")
```

```
print("-----")
```

```
# Simpson's 1/3 rule computation
```

```
sum_odd = sum(f_values[i] for i in range(1, n, 2))
```

```

sum_even = sum(f_values[i] for i in range(2, n, 2))

I = (h / 3) * (f_values[0] + 4 * sum_odd + 2 * sum_even + f_values[-1])

# Show steps

print(f"\nh = (b - a) / n = ({b} - {a}) / {n} = {h}")

print("\nUsing Simpson's 1/3 Rule:")

print(f"I = (h/3) * [f(x0) + 4*(f(x1) + f(x3) + ...) + 2*(f(x2) + f(x4) + ...) + f(xn)]")

print(f"I = ({h}/3) * [{f_values[0]:.6f} + 4*{{sum_odd:.6f}} + 2*{{sum_even:.6f}} + {f_values[-1]:.6f}]")

```

Display final result

```

print("\n-----")
print(f"Approximate value of the integral I = {I:.6f}")
print("-----")

```

OUTPUT:

i	x(i)	f(x(i)) = e^(-x^2)
0	0.0000	1.000000
1	0.2500	0.939413
2	0.5000	0.778801
3	0.7500	0.569783
4	1.0000	0.367879

h = (b - a) / n = (1 - 0) / 4 = 0.25

Using Simpson's 1/3 Rule:
 $I = (h/3) * [f(x0) + 4*(f(x1) + f(x3) + ...) + 2*(f(x2) + f(x4) + ...) + f(xn)]$
 $I = (0.25/3) * [1.000000 + 4*(1.509196) + 2*(0.778801) + 0.367879]$

Approximate value of the integral I = 0.746855

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

2. SIMPSON'S 3/8 RULE

AIM: AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE SIMPSON'S 3/8 RULE.

PROGRAM:

```
# Simpson's 3/8 Rule for I = ∫(0 to 1) e^(-x^2) dx with n = 3
```

```
import math
```

```
# Define the function
```

```
def f(x):
```

```
    return math.exp(-x**2)
```

```
# Given values
```

```
a = 0 # lower limit
```

```
b = 1 # upper limit
```

```
n = 3 # must be a multiple of 3 for Simpson's 3/8 rule
```

```
# Step size
```

```
h = (b - a) / n
```

```
# Generate x and f(x)
```

```
x = [a + i * h for i in range(n + 1)]
```

```
f_values = [f(xi) for xi in x]
```

```
# Display table
```

```
print("-----")
```

```
print(" i | x(i) | f(x(i)) = e^(-x^2)")
```

```
print("-----")
```

```
    for i in range(n + 1):
```

```
        print(f"{i:<3} | {x[i]:<8.6f} | {f_values[i]:<10.6f}")
```

```
print("-----")
```

```
# Apply Simpson's 3/8 rule formula (for n=3)
```

```
I = (3 * h / 8) * (f_values[0] + 3*f_values[1] + 3*f_values[2] + f_values[3])
```

```
# Step-by-step output
```

```
print(f"\nh = (b - a) / n = ({b}) / ({n}) = {h:.6f}")
```

```
print("\nUsing Simpson's 3/8 Rule:")
```

```
print(f"I = (3h/8) * [f(x0) + 3f(x1) + 3f(x2) + f(x3)]")
```

```
print(f"I = (3*{h:.6f}/8) * [{f_values[0]:.6f} + 3*{f_values[1]:.6f} + 3*{f_values[2]:.6f} + {f_values[3]:.6f}]")
```

```
# Final result
```

```
print("\n-----")
```

```
print(f"Approximate value of the integral I = {I:.6f}")
```

```
print("-----")
```

OUTPUT:

i	x(i)	f(x(i)) = e^(-x^2)
0	0.000000	1.000000
1	0.333333	0.894839
2	0.666667	0.641180
3	1.000000	0.367879

```
h = (b - a)/n = (1 - 0)/3 = 0.333333
```

```
Using Simpson's 3/8 Rule:
```

```
I = (3h/8) * [f(x0) + 3f(x1) + 3f(x2) + f(x3)]
I = (3*0.333333/8) * [1.000000 + 3*0.894839 + 3*0.641180 + 0.367879]
```

```
-----
```

```
Approximate value of the integral I = 0.746992
```

```
-----
```

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

TRANSPORTATION PROBLEM

TRANSPORTATION PROBLEM USING NORTHWEST METHOD

AIM: TO WRITE A PROGRAM IN PYTHON TO DEMONSTRATE TRANSPORTATION PROBLEM USING NORTHWEST METHOD.

PROGRAM:

```
# Northwest Corner Method - step-by-step  
# Problem data (from your sheet)  
# Costs matrix: rows = origins O1,O2 ; cols = destinations D1,D2,D3
```

```
costs = [  
    [8, 6, 10], # O1  
    [10, 4, 9] # O2  
]
```

```
supply = [2000, 2500] # supplies for O1, O2  
demand = [1500, 2000, 1000] # demands for D1, D2, D3
```

```
# Make copies so we don't destroy originals if we want to reuse them
```

```
sup = supply.copy()  
dem = demand.copy()
```

```
# Prepare an allocation matrix initialized to zeros  
alloc = [[0 for _ in range(len(demand))] for _ in range(len(supply))]
```

```
print("Northwest Corner Method - step by step\n")
```

```
print("Initial supply:", supply)  
print("Initial demand:", demand)  
print()  
i = 0 # origin index (row)  
j = 0 # destination index (col)  
step = 0
```

```
# Loop until all supplies and demands are satisfied
```

```
while i < len(sup) and j < len(dem):
```

```
    step += 1
```

```
    qty = min(sup[i], dem[j])
```

```
    alloc[i][j] = qty
```

```
    sup[i] -= qty
```

```
    dem[j] -= qty
```

```
# Print step details
```

```
print(f"Step {step}: Allocate {qty} units to cell O{i+1}, D{j+1}")
```

```
print(f"  cost per unit = {costs[i][j]}")
```

```
print(f"  Remaining supply for O{i+1} = {sup[i]}")
```

```
print(f"  Remaining demand for D{j+1} = {dem[j]}\n")
```

```
# Move to next row or column (if supply exhausted move down, if demand exhausted move right)
```

```
# If both become zero, move one and then the other: standard choice is to advance column (j) after row
```

```
if sup[i] == 0 and dem[j] == 0:
```

```
    # If both exhausted, advance (commonly advance row or column) - advance column then row to avoid skipping
```

```
    # But we must ensure not to go out of bounds: handle carefully:
```

```
# Advance column if possible, otherwise advance row.
```

```
if j + 1 < len(dem):
```

```
    j += 1
```

```
elif i + 1 < len(sup):
```

```
    i += 1
```

```
else:
```

```
break # Finished
```

```
elif sup[i] == 0:
```

```
    i += 1
```

```
elif dem[j] == 0:
```

```
    j += 1
```

```
# This else normally won't happen because qty = min(sup[i], dem[j]) forces one to zero
```

```
else:
```

```
    pass
```

```
# Display final allocation matrix
```

```
print("Final allocation matrix (rows = O1,O2 ; cols = D1,D2,D3):\n")  
header = [" | "] + [f" D{c+1}" for c in range(len(demand))] + [" | Supply"]  
print("".join(header))  
for r in range(len(alloc)):  
    row_str = [f"O{r+1} | "] + [f"{alloc[r][c]:6d}" for c in range(len(alloc[r]))] + [f" | {supply[r]:6d}"]  
    print("".join(row_str))  
print()
```

```
# Compute total cost
```

```
total_cost = 0  
for r in range(len(alloc)):  
  
    for c in range(len(alloc[0])):  
        total_cost += alloc[r][c] * costs[r][c]  
  
# Print non-zero allocations  
print("Allocations (non-zero):")  
for r in range(len(alloc)):  
    for c in range(len(alloc[0])):  
        if alloc[r][c] != 0:  
            print(f"O{r+1}, D{c+1} -> {alloc[r][c]} units at cost {costs[r][c]} => contribution = {alloc[r][c] *  
costs[r][c]}")  
  
print(f"\nTotal transportation cost (initial NW-corner solution) = {total_cost}")
```

OUTPUT:

```
Northwest Corner Method - step by step

Initial supply: [2000, 2500]
Initial demand: [1500, 2000, 1000]

Step 1: Allocate 1500 units to cell (O1, D1)
cost per unit = 8
Remaining supply for O1 = 500
Remaining demand for D1 = 0

Step 2: Allocate 500 units to cell (O1, D2)
cost per unit = 6
Remaining supply for O1 = 0
Remaining demand for D2 = 1500

Step 3: Allocate 1500 units to cell (O2, D2)
cost per unit = 4
Remaining supply for O2 = 1000
Remaining demand for D2 = 0

Step 4: Allocate 1000 units to cell (O2, D3)
cost per unit = 9
Remaining supply for O2 = 0
Remaining demand for D3 = 0

Final allocation matrix (rows = O1,O2 ; cols = D1,D2,D3):
    D1   D2   D3   | Supply
O1   1500     500     0   |   2000
O2       0     1500   1000   |   2500

Allocations (non-zero):
(O1, D1) -> 1500 units at cost 8 => contribution = 12000
(O1, D2) -> 500 units at cost 6 => contribution = 3000
(O2, D2) -> 1500 units at cost 4 => contribution = 6000
(O2, D3) -> 1000 units at cost 9 => contribution = 9000

Total transportation cost (initial NW-corner solution) = 30000
```

CONCLUSION: THE ABOVE PROGRAM HAS BEEN EXECUTED SUCCESSFULLY.

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