Lab 5

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1.

```
E = 10^-9;
N_MAX = 25;
% Functions
f_a = @(x) x.^2 .* exp(-x.^2);
f_b = @(x) x.^(1/2) .* exp(-x.^2);
```

```
% Plot functions
x = 0:1e-3:2;

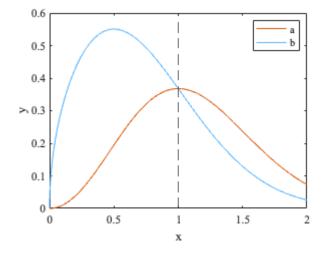
figure

colororder('reef')
set(groot, 'DefaultAxesFontName', 'Times New Roman', 'DefaultAxesFontSize',
12)
set(groot, 'DefaultTextFontName', 'Times New Roman', 'DefaultTextFontSize',
14)

plot(x, f_a(x), x, f_b(x))
xline(1, '--')

legend('a', 'b', 'Location', 'northeast')

xlabel('x')
ylabel('y')
saveas(gcf, 'lab5/functions.png')
```



```
% Function (a) with limits of integration from 0 to 1 fprintf('----\n')
```

```
fprintf('Function (a) with limits of integration from 0 to 1')
```

Function (a) with limits of integration from 0 to 1

```
[R, n_romberg] = romberg(f_a, 0, 1, E, N_MAX);
```

(Romberg) n: 6

(Romberg) Approximation: 0.18947

(Romberg) Number of function evaluations: 134

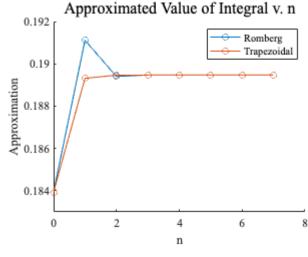
```
[T, n_trapezoidal] = trapezoidal(f_a, 0, 1, E, N_MAX);
```

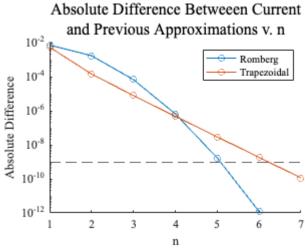
(Trapezoidal) n: 7

(Trapezoidal) Approximation: 0.18947

(Trapezoidal) Number of function evaluations: 129

```
plot_romberg_trapezoidal(R, n_romberg, T, n_trapezoidal, 'lab5/a01')
```





```
R_a_01 = R(n_romberg+1, n_romberg+1);
```

```
% Function (a) with limits of integration from 1 to 2 fprintf('----\n')
```

```
fprintf('Function (a) with limits of integration from 1 to 2')
```

Function (a) with limits of integration from 1 to 2

```
[R, n_romberg] = romberg(f_a, 1, 2, E, N_MAX);
```

(Romberg) n: 5

(Romberg) Approximation: 0.23325

(Romberg) Number of function evaluations: 69

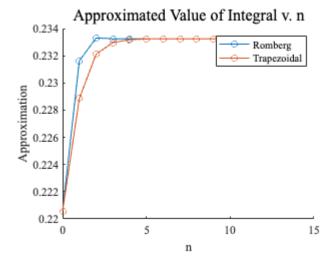
```
[T, n_trapezoidal] = trapezoidal(f_a, 1, 2, E, N_MAX);
```

(Trapezoidal) n: 13

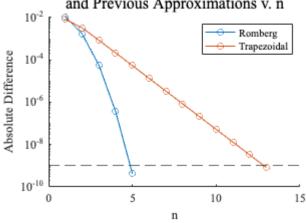
(Trapezoidal) Approximation: 0.23325

(Trapezoidal) Number of function evaluations: 8193

```
plot_romberg_trapezoidal(R, n_romberg, T, n_trapezoidal, 'lab5/a12')
```



Absolute Difference Betweeen Current and Previous Approximations v. n



```
R_a_12 = R(n_romberg+1, n_romberg+1);
```

```
% Function (b) with limits of integration from 0 to 1 fprintf('----\n')
```

```
fprintf('Function (b) with limits of integration from 0 to 1')
```

Function (b) with limits of integration from 0 to 1

```
[R, n_romberg] = romberg(f_b, 0, 1, E, N_MAX);
```

(Romberg) n: 18

(Romberg) Approximation: 0.45339

(Romberg) Number of function evaluations: 524306

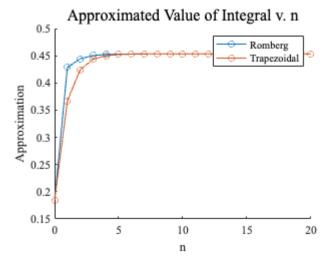
```
[T, n_trapezoidal] = trapezoidal(f_b, 0, 1, E, N_MAX);
```

(Trapezoidal) n: 20

(Trapezoidal) Approximation: 0.45339

(Trapezoidal) Number of function evaluations: 1048577

plot_romberg_trapezoidal(R, n_romberg, T, n_trapezoidal, 'lab5/b01')



Absolute Difference Betweeen Current and Previous Approximations v. n Romberg Trapezoidal 10-6 10-8 10-10 5 10 15 20

```
R_b_01 = R(n_romberg+1, n_romberg+1);
```

```
% Function (b) with limits of integration from 1 to 2 fprintf('----\n')
```

```
fprintf('Function (b) with limits of integration from 1 to 2')
```

Function (b) with limits of integration from 1 to 2

```
[R, n_romberg] = romberg(f_b, 1, 2, E, N_MAX);
```

(Romberg) n: 5

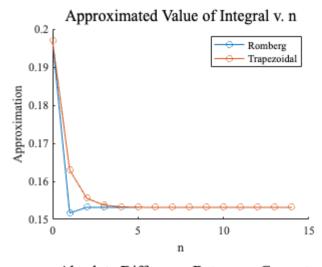
(Romberg) Approximation: 0.15316

(Romberg) Number of function evaluations: 69

[T, n_trapezoidal] = trapezoidal(f_b, 1, 2, E, N_MAX);

```
(Trapezoidal) n: 14
(Trapezoidal) Approximation: 0.15316
(Trapezoidal) Number of function evaluations: 16385
```

```
plot_romberg_trapezoidal(R, n_romberg, T, n_trapezoidal, 'lab5/b12')
```



n

```
R_b_12 = R(n_romberg+1, n_romberg+1);
```

2.

```
% Gauss points and weights from:
% https://pomax.github.io/bezierinfo/legendre-gauss.html

max_n = 11;
a_01 = zeros(max_n, 1);
a_12 = zeros(max_n, 1);
b_01 = zeros(max_n, 1);
b_12 = zeros(max_n, 1);
% n = 1
```

```
ai_xi1 = [2, 0];
% n = 2
ai_xi2 = [
   -0.5773502691896257;
    0.5773502691896257
];
% n = 3
ai_xi3 = [
   0.888888888888888888888888888888
                           0.00000000000000000;
   0.55555555555556,
                          -0.7745966692414834;
   0.555555555555556,
                           0.7745966692414834
];
% n = 4
ai_xi4 = [
   0.6521451548625461,
                           -0.3399810435848563;
                           0.3399810435848563;
   0.6521451548625461,
   0.3478548451374538,
                         -0.8611363115940526;
   0.3478548451374538,
                           0.8611363115940526
];
% n = 5
ai_xi5 = [
   0.5688888888888889, 0.0000000000000000;
   0.4786286704993665,
                          -0.5384693101056831;
    0.4786286704993665,
                           0.5384693101056831;
                         -0.9061798459386640;
    0.2369268850561891,
   0.2369268850561891,
                          0.9061798459386640
];
% n = 6
ai xi6 = [
   0.3607615730481386,
                           0.6612093864662645;
   0.3607615730481386,
                           -0.6612093864662645;
   0.4679139345726910,
                           -0.2386191860831969;
    0.4679139345726910,
                           0.2386191860831969;
   0.1713244923791704,
                          -0.9324695142031521;
    0.1713244923791704,
                           0.9324695142031521
];
% n = 7
ai_xi7 = [
   0.4179591836734694,
                           0.00000000000000000;
   0.3818300505051189,
                           0.4058451513773972;
    0.3818300505051189,
                           -0.4058451513773972;
   0.2797053914892766,
                           -0.7415311855993945;
    0.2797053914892766,
                           0.7415311855993945;
    0.1294849661688697,
                           -0.9491079123427585;
```

```
0.1294849661688697,
                            0.9491079123427585;
];
% n = 8
ai xi8 = [
    0.3626837833783620,
                            -0.1834346424956498;
                            0.1834346424956498;
    0.3626837833783620,
    0.3137066458778873,
                            -0.5255324099163290;
    0.3137066458778873,
                            0.5255324099163290;
    0.2223810344533745,
                            -0.7966664774136267;
    0.2223810344533745,
                            0.7966664774136267;
    0.1012285362903763,
                            -0.9602898564975363;
    0.1012285362903763,
                            0.9602898564975363
];
% n = 9
ai_xi9 = [
    0.3302393550012598,
                            0.00000000000000000;
    0.1806481606948574,
                            -0.8360311073266358;
    0.1806481606948574,
                            0.8360311073266358;
    0.0812743883615744,
                            -0.9681602395076261;
    0.0812743883615744,
                            0.9681602395076261;
    0.3123470770400029,
                            -0.3242534234038089;
    0.3123470770400029,
                            0.3242534234038089;
    0.2606106964029354,
                            -0.6133714327005904;
    0.2606106964029354,
                            0.6133714327005904
];
% n = 10
ai_xi10 = [
    0.2955242247147529,
                            -0.1488743389816312;
    0.2955242247147529,
                            0.1488743389816312;
    0.2692667193099963,
                            -0.4333953941292472;
    0.2692667193099963,
                            0.4333953941292472;
    0.2190863625159820,
                            -0.6794095682990244;
    0.2190863625159820,
                            0.6794095682990244;
    0.1494513491505806,
                            -0.8650633666889845;
    0.1494513491505806,
                            0.8650633666889845;
    0.0666713443086881,
                            -0.9739065285171717;
    0.0666713443086881,
                            0.9739065285171717
];
% n = 11
ai_xi11 = [
    0.2729250867779006,
                            0.00000000000000000;
    0.2628045445102467,
                            -0.2695431559523450;
    0.2628045445102467,
                            0.2695431559523450;
    0.2331937645919905,
                            -0.5190961292068118;
    0.2331937645919905,
                            0.5190961292068118;
                            -0.7301520055740494;
    0.1862902109277343,
```

```
0.1862902109277343.
                           0.7301520055740494;
    0.1255803694649046,
                           -0.8870625997680953;
    0.1255803694649046,
                           0.8870625997680953;
    0.0556685671161737,
                          -0.9782286581460570;
    0.0556685671161737,
                           0.9782286581460570;
];
ai_xi = [
    ai_xi1; ai_xi2; ai_xi3;
    ai_xi4; ai_xi5; ai_xi6;
    ai_xi7; ai_xi8; ai_xi9;
    ai xi10; ai xi11
];
```

```
row = 1
```

row = 1

```
for n = 1 : max_n
    fprintf('n = %d -
n)
    ai = ai_xi(row:row + n - 1, 1);
    xi = ai xi(row:row + n - 1, 2);
    % fprintf('Function (a) with limits of integration from 0 to 1')
    a 01(n) = gaussian(f a, 0, 1, n, ai, xi);
    % fprintf('Function (a) with limits of integration from 1 to 2')
    a_12(n) = gaussian(f_a, 1, 2, n, ai, xi);
    % fprintf('Function (b) with limits of integration from 0 to 1')
    b_01(n) = gaussian(f_b, 0, 1, n, ai, xi);
    % fprintf('Function (b) with limits of integration from 1 to 2')
    b_12(n) = gaussian(f_b, 1, 2, n, ai, xi);
    % fprintf('Number of function evaluations: %d\n', n);
    row = row + n;
end
```

Gaussian approximation: 0.19470
Evaluations: 1
Gaussian approximation: 0.23715
Evaluations: 1
Gaussian approximation: 0.55070
Evaluations: 1
Gaussian approximation: 0.12909
Evaluations: 1

n = 2 Gaussian approximation: 0.18832 Evaluations: 2 Gaussian approximation: 0.23439 Evaluations: 2 Gaussian approximation: 0.45820 Evaluations: 2 Gaussian approximation: 0.15414 Evaluations: 2 n = 3 Gaussian approximation: 0.18954 Evaluations: 3 Gaussian approximation: 0.23320 Evaluations: 3 Gaussian approximation: 0.45590 Evaluations: 3 Gaussian approximation: 0.45590 Evaluations: 3 Gaussian approximation: 0.15316 Evaluations: 3 Gaussian approximation: 0.18947 Evaluations: 4 Gaussian approximation: 0.23325 Evaluations: 4 Gaussian approximation: 0.45455 Evaluations: 4 Gaussian approximation: 0.15316 Evaluations: 4 n = 5 Gaussian approximation: 0.15316 Evaluations: 5 Gaussian approximation: 0.23325 Evaluations: 5 Gaussian approximation: 0.23325 Evaluations: 5 Gaussian approximation: 0.23325 Evaluations: 5 Gaussian approximation: 0.45402 Evaluations: 5 Gaussian approximation: 0.45402 Evaluations: 6 Gaussian approximation: 0.18947 Evaluations: 6 Gaussian approximation: 0.18947 Evaluations: 6 Gaussian approximation: 0.18947 Evaluations: 6 Gaussian approximation: 0.45377 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 7 Gaussian approximation: 0.18947 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 7 Gaussian approximation: 0.45364 Evaluations: 8 Gaussian approximation: 0.45365 Evaluations: 8 Gaussian approximation: 0.45366 Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8 Gaussian approximation: 0.45366 Evaluations: 8 Gaussian approximation: 0.45366 Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8		
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Evaluations: 3 n = 4		0. 15316
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Gaussian approximation: 0.23325 Evaluations: 6 Gaussian approximation: 0.45377 Evaluations: 6 Gaussian approximation: 0.15316 Evaluations: 6 n = 7		0.18947
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Gaussian approximation: 0.45377 Evaluations: 6 Gaussian approximation: 0.15316 Evaluations: 6 n = 7	_	0.23325
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<pre>n = 7</pre>	Evaluations: 6	
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Evaluations: 7 Gaussian approximation: 0.15316 Evaluations: 7 n = 8		
Gaussian approximation: 0.15316 Evaluations: 7 n = 8		0.45364
Evaluations: 7 n = 8		0.15316
n = 8		0.12310
Gaussian approximation: 0.18947 Evaluations: 8 Gaussian approximation: 0.23325 Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8 Gaussian approximation: 0.15316 Evaluations: 8	n - 8	
Evaluations: 8 Gaussian approximation: 0.23325 Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8 Gaussian approximation: 0.15316 Evaluations: 8		
Gaussian approximation: 0.23325 Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8 Gaussian approximation: 0.15316 Evaluations: 8		
Evaluations: 8 Gaussian approximation: 0.45356 Evaluations: 8 Gaussian approximation: 0.15316 Evaluations: 8		0.23325
Evaluations: 8 Gaussian approximation: 0.15316 Evaluations: 8		
Gaussian approximation: 0.15316 Evaluations: 8	Gaussian approximation:	0.45356
Evaluations: 8		
		0.15316
n = 9		

Gaussian approximation: 0.18947

Evaluations: 9

Gaussian approximation: 0.23325

Evaluations: 9

Gaussian approximation: 0.45351

Evaluations: 9

Gaussian approximation: 0.15316

Evaluations: 9

n = 10 -----

Gaussian approximation: 0.18947

Evaluations: 10

Gaussian approximation: 0.23325

Evaluations: 10

Gaussian approximation: 0.45348

Evaluations: 10

Gaussian approximation: 0.15316

Evaluations: 10

n = 11 -----

Gaussian approximation: 0.18947

Evaluations: 11

Gaussian approximation: 0.23325

Evaluations: 11

Gaussian approximation: 0.45346

Evaluations: 11

Gaussian approximation: 0.15316

Evaluations: 11

t = table((1 : max_n).', a_01, a_12, b_01, b_12)

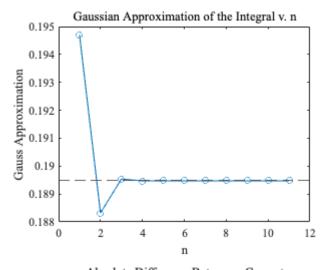
$t = 11 \times 5$ table

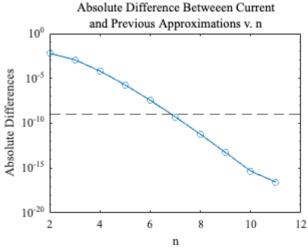
	Var1	a_01	a_12	b_01	b_12
1	1	0.1947	0.2371	0.5507	0.1291
2	2	0.1883	0.2344	0.4582	0.1541
3	3	0.1895	0.2332	0.4559	0.1532
4	4	0.1895	0.2333	0.4545	0.1532
5	5	0.1895	0.2333	0.4540	0.1532
6	6	0.1895	0.2333	0.4538	0.1532
7	7	0.1895	0.2333	0.4536	0.1532
8	8	0.1895	0.2333	0.4536	0.1532
9	9	0.1895	0.2333	0.4535	0.1532
10	10	0.1895	0.2333	0.4535	0.1532
11	11	0.1895	0.2333	0.4535	0.1532

fprintf('Function (a) with limits of integration from 0 to 1');

Function (a) with limits of integration from 0 to 1

plot_gaussian_results(a_01, max_n, R_a_01, 'lab5/a01')

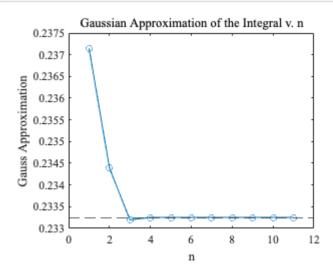


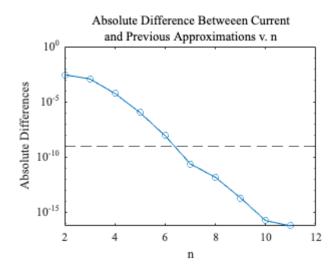


fprintf('Function (a) with limits of integration from 1 to 2')

Function (a) with limits of integration from 1 to 2

plot_gaussian_results(a_12, max_n, R_a_12, 'lab5/a12')

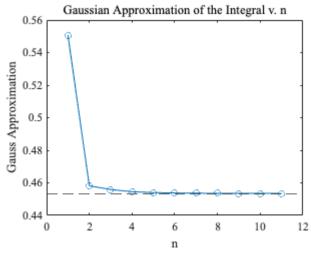


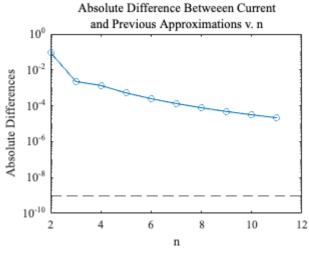


fprintf('Function (b) with limits of integration from 0 to 1')

Function (b) with limits of integration from 0 to 1

plot_gaussian_results(b_01, max_n, R_b_01, 'lab5/b01')

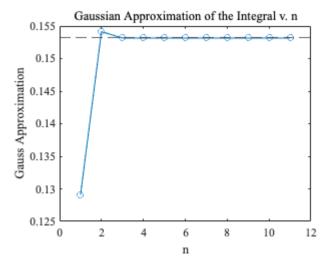


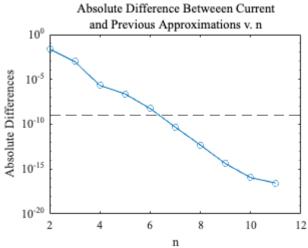


```
fprintf('Function (b) with limits of integration from 1 to 2')
```

Function (b) with limits of integration from 1 to 2

```
plot_gaussian_results(b_12, max_n, R_b_12, 'lab5/b12')
```





3.

```
% Interval a <= t <= b
a = 0;
b = 3;

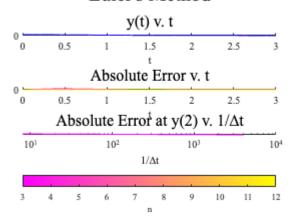
% Initial value problem
dy_dt = @(y, t) -2 * y * t / (1 + t^2);
y0 = 1;

% Analytical solution
y = @(t) 1 ./ (1 + t.^2);

n_min = 3;
n_max = 12;</pre>
```

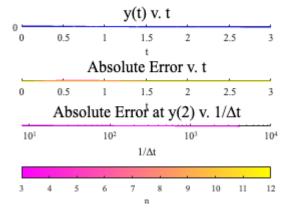
```
% Euler's method
y2_error_euler = approximate(dy_dt, y0, y, a, b, n_min, n_max, @euler,
'Euler''s Method');
saveas(gcf, 'lab5/euler.png');
```

Euler's Method



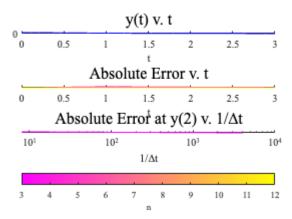
```
% Midpoint rule
y2_error_midpoint = approximate(dy_dt, y0, y, a, b, n_min, n_max,
@midpoint_rule, 'Midpoint Rule');
saveas(gcf, 'lab5/midpoint.png')
```

Midpoint Rule



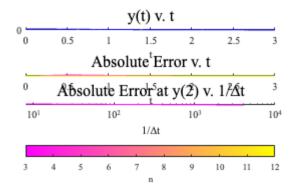
```
% Modified Euler's method
y2_error_modified_euler = approximate(dy_dt, y0, y, a, b, n_min, n_max,
@modified_euler, 'Modified Euler''s Method');
saveas(gcf, 'lab5/modified_euler.png')
```

Modified Euler's Method



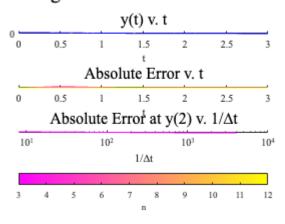
```
% 2-step Adams-Bashforth/Adams-Moulton predictor corrector
y2_error_adams = approximate(dy_dt, y0, y, a, b, n_min, n_max, @adams,
{'Adams-Bashforth & Adams-Moulton', 'Predictor Corrector'});
saveas(gcf, 'lab5/adams.png')
```

Adams-Bashforth & Adams-Moulton Predictor Corrector



```
% Runge-Kutta 4th order method
y2_error_runge_kutta = approximate(dy_dt, y0, y, a, b, n_min, n_max,
@runge_kutta_4, 'Runge-Kutta 4th Order Method');
saveas(gcf, 'lab5/runge_kutta.png')
```

Runge-Kutta 4th Order Method



```
% All y(2) v. 1/Δt
inverse_delta_t = 2 .^ (n_min : n_max);
figure
hold on
set(groot, 'DefaultAxesFontName', 'Times New Roman', 'DefaultAxesFontSize',
12)
set(groot, 'DefaultTextFontName', 'Times New Roman', 'DefaultTextFontSize',
14)
plot(inverse_delta_t, y2_error_euler, '-o');
plot(inverse_delta_t, y2_error_midpoint, '-o');
plot(inverse_delta_t, y2_error_modified_euler, '-o')
plot(inverse_delta_t, y2_error_adams, '-o');
plot(inverse_delta_t, y2_error_runge_kutta, '-o');
hold off
set(gca, 'YScale', 'log', 'XScale', 'log')
xlabel('1/Δt')
ylabel('Absolute Error')
title('Absolute Error at y(2) v. 1/Δt', 'FontSize', 20, 'FontWeight',
'normal');
legend({'Euler', 'Midpoint', 'Modified Euler', 'Adams', 'Runge-Kutta'},
'Location', 'bestoutside');
saveas(gcf, 'lab5/y2_errors.png')
```


Functions

```
function [R, n, romberg_evaluations] = romberg(f, a, b, e, n_max)
   %{
        Approximates the integral from a to b of f(x) with Romberg's method.
        Input:
            f: function
            a: lower limit
            b: upper limit
            e: error
        Returns:
            R: Romberg approximation
            n: number of times subintervals were halved
            romberg evaluations: number of function evaluations
    %}
    R = zeros(n_max + 1, n_max + 1);
    romberg_evaluations = 0;
    % Calculate the Romberg table row by row
    for n = 0: n_max
        for j = 0 : n
            % Initialize R with a composite trapezoidal approximation
            if j == 0
                [R(n+1, j+1), trapezoidal_evaluations] =
composite_trapezoidal(f, a, b, 2^n);
                romberg evaluations = romberg evaluations +
trapezoidal_evaluations;
            else
```

```
R(n+1, j+1) = (4^j * R(n+1, j) - R(n, j)) / (4^j - 1);
            end
        end
        % Check if R(n, n) is close enough to R(n-1, n-1)
        if n > 0 \&\& abs(R(n+1, n+1) - R(n, n)) < e
            break:
        end
    end
    fprintf('(Romberg) n: %d\n', n);
    fprintf('(Romberg) Approximation: %0.5f\n', R(n+1, n+1));
    fprintf('(Romberg) Number of function evaluations: %d',
romberg evaluations);
end
function [I, n, trapezoidal_evaluations] = trapezoidal(f, a, b, e, n_max)
    %{
        Approximates the integral from a to b of f(x) with the Composite
        Trapezoidal rule.
        Input:
            f: function
            a: lower limit
            b: upper limit
            e: error
        Returns:
            I: Composite Trapezoidal approximation
            n: number of times subintervals were halved
            trapezoidal evaluations: number of function evaluations
    %}
    I = zeros(n_max + 1, 1);
    trapezoidal evaluations = 0;
    for n = 0: n_max
        [I(n+1), trapezoidal_evaluations] = composite_trapezoidal(f, a, b,
2^n);
        % Check if I(n) is close enough to I(n-1)
        if n > 0 \&\& abs(I(n+1) - I(n)) < e
            break:
        end
    end
    fprintf('(Trapezoidal) n: %d\n', n);
    fprintf('(Trapezoidal) Approximation: %0.5f\n', I(n+1));
```

```
fprintf('(Trapezoidal) Number of function evaluations: %d',
trapezoidal_evaluations);
end
function [I, evaluations] = composite_trapezoidal(f, a, b, n)
    %{
        Approximates the integral from a to b of f(x) with n subintervals.
        Input:
            f: function
            a: lower limit
            b: upper limit
            n: number of subintervals
        Returns:
            I: approximated value of integral
            evaluations: number of function evaluations
    %}
    h = (b - a)/n;
    I = 0;
    evaluations = 0;
    for i = 0 : n
        x = a + i * h;
        if i == 0 || i == n
            I = I + f(x);
        else
            I = I + 2 * f(x);
        end
        evaluations = evaluations + 1;
    end
    I = I * h/2;
end
function plot_romberg_trapezoidal(R, n_romberg, T, n_trapezoidal, file)
    %{
        Plots approximated value of integral v. n.
        Plots absolute differences v. n.
        Input:
            R: Romberg table
            n_romberg: number of steps (2^n)
            T: Trapezoidal approximations
            n_trapezoidal: number of steps (2^n)
            file: output file prefix
    %}
```

```
% Get Romberg approximations along the diagonal
   romberg approximations = zeros(n romberg + 1, 1);
   for i = 1 : n_romberg + 1
        romberg approximations(i) = R(i, i);
   end
   trapezoidal_approximations = T(1:n_trapezoidal + 1);
   title_font = {'FontSize', 16, 'FontWeight', 'normal'};
   % Plot approximations v. n
   figure
   set(groot, 'DefaultAxesFontName', 'Times New Roman',
'DefaultAxesFontSize', 12)
   set(groot, 'DefaultTextFontName', 'Times New Roman',
'DefaultTextFontSize', 14)
   hold on
   plot(0:n romberg, romberg approximations, '-o')
   plot(0:n_trapezoidal, trapezoidal_approximations, '-o')
   hold off
   legend('Romberg', 'Trapezoidal', 'Location', 'northeast')
   xlabel('n')
   ylabel('Approximation')
   title('Approximated Value of Integral v. n', title_font{:})
   saveas(gcf, sprintf('%s-rt-approx.png', file))
   % Compute errors
   romberg_errors = zeros(n_romberg, 1);
   for i = 1 : n_romberg
       romberg errors(i) = abs(R(i+1, i+1) - R(i, i));
   end
   trapezoidal_errors = zeros(n_trapezoidal, 1);
   for i = 1 : n trapezoidal
       trapezoidal_errors(i) = abs(T(i+1) - T(i));
   end
   % Plot errors v. n
   figure
   set(groot, 'DefaultAxesFontName', 'Times New Roman',
'DefaultAxesFontSize', 12)
   set(groot, 'DefaultTextFontName', 'Times New Roman',
'DefaultTextFontSize', 14)
```

```
hold on
    plot(1:n romberg, romberg errors, '-o')
    plot(1:n_trapezoidal, trapezoidal_errors, '-o')
    vline(1e-9, '--')
    hold off
    set(gca, 'YScale', 'log')
    legend('Romberg', 'Trapezoidal', 'Location', 'northeast')
    xlabel('n')
    ylabel('Absolute Difference')
    title({'Absolute Difference Betweeen Current', 'and Previous
Approximations v. n'}, title_font{:})
    saveas(gcf, sprintf('%s-rt-diff.png', file))
end
function [G, evaluations] = gaussian(f, a, b, n, ai, xi)
        Approximates the integral from a to b of f(x) with Gaussian
quadrature.
        Input:
            f: function
            a: lower limit
            b: upper limit
            n: number of points
            ai: coefficients
            xi: Gauss points
        Returns:
            G: approximated value of integral
            evaluations: number of function evaluations
    %}
        % Map Gauss points from (-1, 1) to (a, b)
        yi = (a+b)/2 + (b-a)/2 * xi;
    G = 0;
    evaluations = 0;
    for i = 1 : n
        G = G + ai(i) * f(yi(i));
        evaluations = evaluations + 1;
    end
    G = (b-a)/2 * G;
    fprintf('Gaussian approximation: %0.5f\n', G);
```

```
fprintf('Evaluations: %d\n\n', evaluations);
end
function plot_gaussian_results(G, n, romberg_approximation, file)
   %{
        Plots results of the Gaussian quadrature.
        Input:
            G: Gaussian approximations
            n: maximum number of points (1 to n)
            romberg_approximation: Romberg value of the integral
            file: output file prefix
    %}
   title_font = {'FontSize', 16, 'FontWeight', 'normal'};
   % Plot approximation v. n, with a yline for Romberg approximation
    figure
    set(groot, 'DefaultAxesFontName', 'Times New Roman',
'DefaultAxesFontSize', 12)
    set(groot, 'DefaultTextFontName', 'Times New Roman',
'DefaultTextFontSize', 14)
    plot(1:n, G, '-o')
    yline(romberg_approximation, '--')
    xlabel('n')
    ylabel('Gauss Approximation')
    title('Gaussian Approximation of the Integral v. n', title_font{:})
    saveas(gcf, sprintf('%s-q-approx.png', file))
    % Plot differences v. n
    gaussian_differences = zeros(n-1);
    for i = 1:n-1
        gaussian differences(i) = abs(G(i+1) - G(i));
    end
    figure
    set(groot, 'DefaultAxesFontName', 'Times New Roman',
'DefaultAxesFontSize', 12)
    set(groot, 'DefaultTextFontName', 'Times New Roman',
'DefaultTextFontSize', 14)
    plot(2:n, gaussian_differences, '-o')
    yline(1e-9, '--')
```

```
set(gca, 'YScale', 'log')
    xlabel('n')
    vlabel('Absolute Differences')
    title({'Absolute Difference Betweeen Current', 'and Previous
Approximations v. n'}, title_font{:})
    saveas(gcf, sprintf('%s-g-diff.png', file))
end
function w = euler(dy_dt, a, b, h, y0)
    %{
        Performs Euler's method to approximate y(t) given dy/dt and h.
        Input:
            dy_dt: derivative f(y, t)
            a: lower t limit
            b: upper t limit
            h: step size
            y0: initial condition
        Returns:
            w: approximated y values
    %}
    % Number of subintervals
    n = (b - a) / h;
   w = zeros(n + 1, 1);
   w(1) = y0;
    for i = 1 : n
        t = a + (i-1) * h;
        w(i + 1) = w(i) + h * dy_dt(w(i), t);
    end
end
function w = midpoint_rule(dy_dt, a, b, h, y0)
     %{
        Uses the midpoint rule (RK-2) to approximate y(t) given dy/dt and h.
        Input:
            dy_dt: derivative f(y, t)
            a: lower t limit
            b: upper t limit
            h: step size
            y0: initial condition
        Returns:
            w: approximated y values
```

```
%}
     % Number of subintervals
     n = (b - a) / h;
     w = zeros(n + 1, 1);
     w(1) = y0;
     for i = 1 : n
         t = a + (i-1) * h;
         w_{mid} = w(i) + h/2 * dy_{dt}(w(i), t);
         t_mid = t + h/2;
         w(i + 1) = w(i) + h * dy_dt(w_mid, t_mid);
     end
end
function w = modified_euler(dy_dt, a, b, h, y0)
   %{
        Approximates y(t) given dy/dt and h with Modified Euler's method.
        Input:
            dy_dt: derivative f(y, t)
            a: lower t limit
            b: upper t limit
            h: step size
            y0: initial condition
        Returns:
            w: approximated y values
    %}
    % Number of subintervals
    n = (b - a) / h;
   % Initial condition
    w = zeros(n + 1, 1);
   w(1) = y0;
    for i = 1 : n
        t = a + (i-1) * h;
        % Predict w(i+1) with Euler's method
        w(i + 1) = w(i) + h * dy_dt(w(i), t);
        % Correct w(i+1) with the trapezoidal rule
        w(i + 1) = w(i) + h/2 * (dy_dt(w(i), t) + dy_dt(w(i + 1), t + h));
    end
end
```

```
function w = adams(dy_dt, a, b, h, y0)
   %{
        A 2-step Adams-Bashforth/Adams-Moulton predictor corrector scheme
        with a single correction.
        Input:
            dy_dt: derivative f(y, t)
            a: lower t limit
            b: upper t limit
            h: step size
            y0: initial condition
        Returns:
            w: approximated y values
   %}
    % Number of subintervals
    n = (b - a) / h;
   % Initial conditions
   w = zeros(n + 1, 1);
    f = zeros(n + 1, 1);
   w(1) = y0;
    f(1) = dy_dt(y0, a);
   % Compute the first step using RK-4
    f1 = dy_dt(y0, a);
    f2 = dy_dt(y0 + h/2 * f1, a + h/2);
    f3 = dy_dt(y0 + h/2 * f2, a + h/2);
    f4 = dy_dt(y0 + h * f3, a + h);
   w(2) = y0 + h/6 * (f1 + 2 * (f2 + f3) + f4);
   % y = @(t) 1 ./ (1 + t.^2);
    % w(2) = y(a + h);
    for i = 2 : n
        t = a + (i-1) * h;
        f(i) = dy_dt(w(i), t);
        % Predict w(i+1) with the Adams-Bashforth 2-step predictor
        w(i+1) = w(i) + h/2 * (3 * f(i) - f(i-1));
        f(i+1) = dy_dt(w(i+1), t + h);
        % Correct w(i+1) with the Adams-Moulton 2-step corrector
        w(i+1) = w(i) + h/12 * (5 * f(i+1) + 8 * f(i) - f(i-1));
    end
end
```

```
function w = runge_kutta_4(dy_dt, a, b, h, y0)
        Approximates y(t) given dy/dt and h with the Runge-Kutta 4th order
method.
        Input:
            dy_dt: derivative f(y, t)
            a: lower t limit
            b: upper t limit
            h: step size
            y0: initial condition
        Returns:
            w: approximated y values
    %}
    % Number of subintervals
    n = (b - a) / h;
   w = zeros(n + 1, 1);
   w(1) = y0;
    for i = 1 : n
        t = a + (i-1) * h;
        f1 = dy_dt(w(i), t);
        f2 = dy_dt(w(i) + h/2 * f1, t + h/2);
        f3 = dy_dt(w(i) + h/2 * f2, t + h/2);
        f4 = dy_dt(w(i) + h * f3, t + h);
        w(i + 1) = w(i) + h/6 * (f1 + 2 * (f2 + f3) + f4);
    end
end
function y2_error = approximate(dy_dt, y0, y, a, b, n_min, n_max, method,
method_name)
   %{
        Approximates y(t) given dy/dt and h.
        Input:
            method: function handle, method(dy_dt, a, b, h, y0)
            dy_dt: derivative f(y, t)
            y0: initial condition
            v: exact solution
            a: lower t limit
            b: upper t limit
            n_min: largest step size, 2^(-n_min)
            n_max: smallest step size, 2^(-n_max)
    %}
```

```
tiledlayout('vertical');
   % Font
   set(groot, 'DefaultAxesFontName', 'Times New Roman',
'DefaultAxesFontSize', 12)
   set(groot, 'DefaultTextFontName', 'Times New Roman',
'DefaultTextFontSize', 12)
   title_font = {'FontSize', 20, 'FontWeight', 'normal'};
   subtitle_font = {'FontSize', 16, 'FontWeight', 'normal'};
   % Colors
   colormap spring
   c = spring(n_max - n_min + 1);
   colororder(c);
   ax w = nexttile;
   ax_e = nextfile;
   ax_y2 = nextfile;
   title(ax_w, method_name, title_font{:})
   subtitle(ax_w, 'y(t) v. t', subtitle_font{:})
   ylabel(ax_w, 'y');
xlabel(ax_w, 't');
   hold(ax_w, 'on');
   title(ax_e, 'Absolute Error v. t', subtitle_font{:})
   ylabel(ax_e, 'Absolute Error');
   xlabel(ax_e, 't');
   hold(ax_e, 'on');
   title(ax_y2, 'Absolute Error at y(2) v. 1/Δt', subtitle_font{:});
   ylabel(ax_y2, 'Absolute Error');
   xlabel(ax_y2, \frac{1}{\Delta t});
   hold(ax y2, 'on');
   % Absolute value of the error at y(2) versus 1/Δt on a log-log scale
   y2\_error = zeros(n\_max - n\_min + 1, 1);
   for n = n_min : n_max
       % Step size
       h = 2^{(-n)};
       % Evaluated t
       t = a:h:b;
       % Use the method to approximate y
       w = method(dy_dt, a, b, h, y0);
        plot(ax_w, t, w);
```

```
% Plot errors
        e = abs(y(t).' - w);
        plot(ax_e, t, e);
        % Save calculated w at t = 2
        y2_error(n - n_min + 1) = abs(w(2^n + 1) + 1) - y(2));
    end
   % Plot the analytical solution in y(t) v. t
   t = a:1e-3:b;
    plot(ax_w, t, y(t), 'b')
   % Plot absolute error at y(2) v. 1/\Delta t
   inverse_delta_t = 2 .^ (n_min : n_max);
    plot(ax_y2, inverse_delta_t, y2_error);
   set(ax_y2, 'YScale', 'log', 'XScale', 'log');
   % Colorbar
    clim([n_min, n_max]);
    cb = colorbar();
    cb.Layout.Tile = 'south';
    cb.Label.String = 'n';
    cb.Ticks = n_min:1:n_max;
end
```