

1.

(a)

---

**Algorithm 1:** Solving  $\mathbf{Ax} = \mathbf{b}$  by using the Cholesky Decomposition Algorithm

---

**Require:**  $\mathbf{A}$  is symmetric positive definite

**Input:** Matrix  $\mathbf{A}$   $[a_{ij}]$  of  $n \times n$  and matrix  $\mathbf{b}$   $[b_{ij}]$  of  $n \times m$

**Output:** Solution of equations  $\mathbf{Ax} = \mathbf{b}$ , matrix  $\mathbf{x}$   $[x_{ij}]$  of  $n \times m$

1 **Function** Solve( $\mathbf{A}, \mathbf{b}$ ):

2     /\*Applying Cholesky Decomposition Algorithm to calculate matrix  $\mathbf{C}$  first\*/

3     **for**  $i \leftarrow 1$  **to**  $n$  **do**

4         **for**  $j \leftarrow i+1$  **to**  $n$  **do**

5              $c[ij] \leftarrow 0$ ;

6         **end**

7     **end**

8     **for**  $j \leftarrow 1$  **to**  $n$  **do**

9          $c_{jj} \leftarrow \sqrt{a_{jj}}$ ;

10        **for**  $i \leftarrow j+1$  **to**  $n$  **do**

11             $c_{ij} \leftarrow a_{ij}/c_{jj}$ ;

12            **for**  $k \leftarrow j$  **to**  $i$  **do**

13                 $a_{ik} \leftarrow a_{ik} - c_{ij} \cdot c_{kj}$

14            **end**

15        **end**

16     **end**

17     /\*Solving equation  $\mathbf{Cy} = \mathbf{b}$ \*/

18     **for**  $i \leftarrow 1$  **to**  $n$  **do**

19         **for**  $j \leftarrow 1$  **to**  $i - 1$  **do**

20              $\mathbf{b}_i \leftarrow \mathbf{b}_i - c_{ij} \cdot \mathbf{y}_j$ ;

/\* $\mathbf{y}_i, \mathbf{b}_i$  denote the  $i^{th}$  row of  $\mathbf{y}, \mathbf{b}$ \*/

21         **end**

22          $\mathbf{y}_i \leftarrow \mathbf{b}_i/c_{ii}$ ;

23     **end**

24     /\*Solving equation  $\mathbf{C}^T \mathbf{x} = \mathbf{y}$ \*/

25     **for**  $i \leftarrow n$  **to**  $1$  **step**  $-1$  **do**

26         **for**  $j \leftarrow i + 1$  **to**  $n$  **do**

27              $\mathbf{y}_i \leftarrow \mathbf{y}_i - c_{ji} \cdot \mathbf{x}_j$ ;

/\* $\mathbf{x}_i$  denote the  $i^{th}$  row of  $\mathbf{x}$ \*/

28         **end**

29          $\mathbf{x}_i \leftarrow \mathbf{y}_i/c_{ii}$ ;

30     **end**

31     **return**  $\mathbf{x}$ ;

32 **end**

---

(b)

```

1 function [x] = Gauss_Solver(A, b)
2     [n, m] = size(b);
3     e(1) = 0; f(1) = A(1, 1); g(1) = A(1, 2);
4     e(n) = A(n, n-1); f(n) = A(n, n); g(n) = 0;
5     for i = 2 : n - 1
6         e(i) = A(i, i-1); f(i) = A(i, i); g(i) = A(i, i+1);
7     end
8     for i = 2 : n
9         q = e(i)/f(i-1);
10        f(i) = f(i) - q * g(i-1);
11        for k = 1 : m
12            b(i, k) = b(i, k) - q * b(i - 1, k);
13        end
14    end
15    for k = 1 : m
16        x(n, k) = b(n, k)/f(n);
17        for i = n - 1 : -1 : 1
18            x(i, k) = (b(i, k) - g(i) * x(i+1, k))/f(i);
19        end
20    end
21 end

```

```

1 function [x] = Choleksy(a, b)
2     [n, m] = size(b); c(n, n) = 0;
3     for j = 1 : n
4         c(j, j) = sqrt(a(j, j));
5         for i = j + 1 : n
6             c(i, j) = a(i, j)/c(j, j);
7             for k = j : i
8                 a(i, k) = a(i, k) - c(i, j) * c(k, j);
9             end
10        end
11    end
12    x(n, m) = 0; y(n, m) = 0;
13    for i = 1 : n
14        for j = 1 : i - 1
15            for k = 1 : m
16                b(i, k) = b(i, k) - c(i, j) * y(j, k);
17            end
18        end
19        for k = 1 : m
20            y(i, k) = b(i, k)/c(i, i);
21        end
22    end
23    for i = n : -1 : 1
24        for j = i + 1 : n
25            for k = 1 : m
26                y(i, k) = y(i, k) - c(j, i) * x(j, k);
27            end
28        end
29        for k = 1 : m
30            x(i, k) = y(i, k)/c(i, i);
31        end
32    end
33 end

```

```

1 - for i = 1 : 3
2 -     N = 10 * 10^i;
3 -     z = [2 1 zeros(1, N - 2)] ;
4 -     A = toeplitz(z, z);
5 -     b = ones(N, 1);
6 -     disp("For N = " + N + ", Tridiagonal method: ");
7 -     tic; Gauss_Solver(A, b); toc;
8 -     disp("For N = " + N + ", Choleksy method: ")
9 -     tic; Choleksy(A, b); toc;
10 -    disp("For N = " + N + ", Backslash method: ")
11 -    tic; A\b; toc;
12 - end
13

```

For N = 100, Tridiagonal method:  
时间已过 0.001796 秒。

For N 100, Choleksy method:  
时间已过 0.005401 秒。

For N 100, Backslash method:  
时间已过 0.002063 秒。

For N = 1000, Tridiagonal method:  
时间已过 0.001050 秒。

For N 1000, Choleksy method:  
时间已过 0.622900 秒。

For N 1000, Backslash method:  
时间已过 0.013007 秒。

For N = 10000, Tridiagonal method:  
时间已过 0.003580 秒。

For N 10000, Choleksy method:  
时间已过 1357.763879 秒。

For N 10000, Backslash method:  
时间已过 2.747848 秒。

	n	100	1000	10000
Elapsed	Tridiagonal	0.001796	0.005401	0.002063
time (in	Choleksy	0.001050	0.622900	0.013007
seconds)	Backslash	0.003580	1357.763879	2.747848

## 2.

---

**Algorithm 2:** Finding the eignvalue of a given matrix  $A$  of  $n \times n$

---

**Require:**  $A$  is symmetric positive definite

**Input:** Matrix  $A$  of  $n \times n$

**Output:** eignvalues of a given matrix  $A$

```

1 Function eignvalue( $A$ ):
2   while true do
3      $Q, R \leftarrow \text{HHQR}(A)$ ;
4      $A1 \leftarrow RQ$ ;
5      $\text{sum} \leftarrow 0$ ;
6     for  $i \leftarrow 2$  to  $n$  do
7       for  $j \leftarrow 1$  to  $i - 1$  do
8          $\text{sum} \leftarrow \text{sum} + |A1(i, j)|$ ; /*sum of elements in lower triangle part of
           $A_1^*$ 
9       end
10    end
11    if  $\text{sum} < 1 \times 10^{-5}$  then
12      break;
13    end
14     $A \leftarrow A1$ ;
15  end
16  return  $\text{diag}(A)$ ;
17 end
18 Function HHQR( $A_{n \times n}$ ):
19    $Q \leftarrow I_{n \times n}$ ;
20    $R \leftarrow A$ ;
21   for  $j \leftarrow 1$  to  $n - 1$  do
22      $a_1 \leftarrow R[1 : (j - 1), j]$ ;
23      $a_2 \leftarrow R[j : n, j]$ ;
24      $c \leftarrow \text{sign}(R(j, j)) \cdot \|a_2\|$ ;
25      $v[1 : (j - 1)] \leftarrow 0_{(j-1) \times 1}$ ;
26      $v[j : n] \leftarrow a_2$ ;
27      $v \leftarrow v - ce_j$ ;
28      $H \leftarrow I_{n \times n} - 2vv^T / (v^T v)$ ;
29      $R \leftarrow HR$ ;
30      $Q \leftarrow QH$ ;
31   end
32   return ( $Q, R$ );
33 end

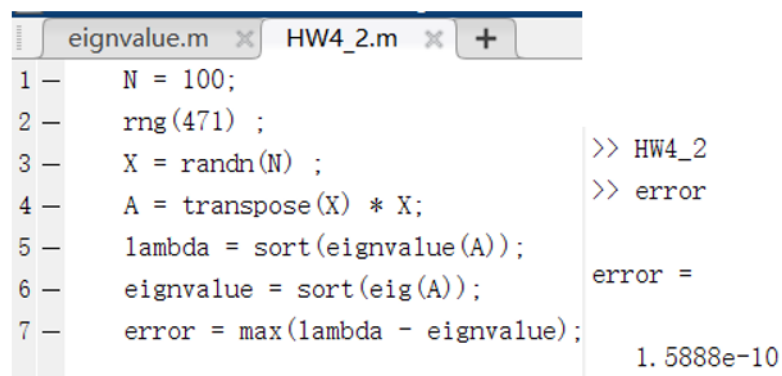
```

---

```

1 function [lambda] = eignvalue(A)
2     [n, ~] = size(A);
3     while 1
4         [Q, R] = HHQR(A);
5         A1 = R * Q;
6         sum = 0;
7         for i = 2 : n
8             for j = 1 : i - 1
9                 sum = sum + abs(A1(i, j));
10            end
11        end
12        if sum < 1e-5
13            break;
14        end
15        A = A1;
16    end
17    lambda = diag(A);
18 end
19
20 function [Q, R] = HHQR(A)
21     n = size(A, 1);
22     I = eye(n);
23     Q = I;
24     R = A;
25     for j = 1 : n - 1
26         a2 = R(j : n, j);
27         c = sign(R(j, j)) * norm(a2);
28         v(1 : (j - 1), 1) = zeros(j - 1, 1);
29         v(j : n, 1) = a2;
30         v = v - c * I(:, j);
31         H = I - 2 * (v * v') / (v' * v);
32         R = H * R;
33         Q = Q * H;
34     end
35 end

```



The screenshot shows a MATLAB command window with two tabs: 'eignvalue.m' and 'HW4\_2.m'. The command window displays the following commands and their outputs:

```

1 - N = 100;
2 - rng(471) ;
3 - X = randn(N) ;
4 - A = transpose(X) * X;
5 - lambda = sort(eignvalue(A));
6 - eignvalue = sort(eig(A));
7 - error = max(lambda - eignvalue);

```

On the right side of the command window, the following prompts and results are shown:

```

>> HW4_2
>> error
error =
1.5888e-10

```

---

**Algorithm 3:** Computing eigenvector of matrix  $\mathbf{A}$

---

**Require:**  $\mathbf{A}$  has  $n$  different eigenvalues

**Input:** Matrix  $\mathbf{A}$  of  $n \times n$ , and its eigenvalue  $\lambda$

**Output:** eigenvector of  $\mathbf{A}$  corresponding to each eigenvalue

```

1 Function eigenvector ( $\mathbf{A}, \lambda$ ):
2   for  $m \leftarrow 1$  to  $n$  do
3      $\mathbf{a} \leftarrow \mathbf{A} - \lambda_m \cdot \mathbf{I}_{n \times n}$ ;
4     for  $k \leftarrow 1$  to  $n - 1$  do
5       /*Try to find a row in which  $a_{lk} \neq 0$ , where  $a_{lk}$  denotes the element in
6         matrix  $\mathbf{a}^*$ */
7        $l \leftarrow k$ ;
8       while  $l \leq n$  do
9         if  $a_{lk} = 0$  then
10          |  $l \leftarrow l + 1$ ;
11        else
12          | break;
13        end
14      end
15      if  $l > n$  then
16        | continue; /*If fail, we have done Gauss elimination for  $k^{th}$  column*/
17      else
18        | swap( $a_l, a_k$ ) /*Exchange row  $k$  with row  $l$  to make sure  $a_{kk} \neq 0$ */
19      end
20      for  $i \leftarrow k + 1$  to  $n - 1$  do
21        |  $m \leftarrow a_{ik}/a_{kk}$ ;
22        |  $a_{ik} \leftarrow 0$ ;
23        | for  $j \leftarrow k + 1$  to  $n$  do
24          |  $a_{ij} \leftarrow a_{ij} - m \cdot a_{kj}$ ;
25        | end
26      end
27       $\mathbf{v}_m \leftarrow 0$ ; /* $\mathbf{v}_m$  denotes the  $m^{th}$  column of matrix  $\mathbf{V}^*$ */
28       $v(n, m) \leftarrow 1$ ; /*Only the last row of  $\mathbf{A}$  would contain only zeros*/
29      for  $i \leftarrow n - 1$  to  $1$  step -1 do
30        |  $y_i \leftarrow 0$ ;
31        | for  $j \leftarrow i + 1$  to  $n$  do
32          |  $y_i \leftarrow y_i - a_{ji} \cdot v_{jm}$ ; /* $\mathbf{x}_i$  denote the  $i^{th}$  row of  $\mathbf{x}^*$ */
33        | end
34        |  $v_{im} \leftarrow y_i/a_{ii}$ ;
35      end
36    end
37    return  $\mathbf{V}$ ; /*each column of matrix  $\mathbf{V}$  is an eigenvector of  $\mathbf{A}^*$ */
38 end

```

---

### 3.

#### (a)

For any  $(t, y_1), (t, y_2) \in \mathcal{A}$ ,  $y_1 < y_2$ , since  $\mathcal{A}$  is convex,  $\forall y \in (y_1, y_2)$ ,  $(t, y) \in \mathcal{A}$ . Since  $\Phi(t, y)$  is differentiable on  $y\mathcal{A}$ , according to mean value theorem,  $\forall (t, y_1), (t, y_2) \in \mathcal{A}$ ,  $y_1 < y_2$ ,  $\exists \xi \in (y_1, y_2)$ , such that

$$\left. \frac{\partial \Phi(t, y)}{\partial y} \right|_{y=\xi} = \frac{\Phi(t, y_1) - \Phi(t, y_2)}{y_1 - y_2}$$

since  $(t, \xi) \in \mathcal{A}$ ,

$$\left| \left. \frac{\partial \Phi(t, y)}{\partial y} \right|_{y=\xi} \right| \leq c$$

and therefore,  $\forall (t, y_1), (t, y_2) \in \mathcal{A}$ ,

$$\left| \frac{\Phi(t, y_1) - \Phi(t, y_2)}{y_1 - y_2} \right| \leq c$$

so  $\Phi$  satisfies a Lipschitz condition in  $y$  on  $\mathcal{A}$  with Lipschitz constant  $c$ .

#### (b)

For all  $(t_1, y_1), (t_2, y_2) \in \mathcal{D}$ ,  $(\lambda t_1 + (1 - \lambda)t_2, \lambda y_1 + (1 - \lambda)y_2)$ ,  $\lambda \in [0, 1]$  denotes a point on line segment joining these two points. Since  $y_1, y_2 \in \mathbb{R}$ ,  $\lambda y_1 + (1 - \lambda)y_2 \in \mathbb{R}$ . Since  $t_1, t_2 \in [t_0, T]$ ,

$$\lambda t_1 + (1 - \lambda)t_2 \in [\min\{t_1, t_2\}, \max\{t_1, t_2\}] \subset [t_0, T]$$

so  $(\lambda t_1 + (1 - \lambda)t_2, \lambda y_1 + (1 - \lambda)y_2) \in \mathcal{D}$  holds for any  $\lambda \in [0, 1]$ . And therefore,  $\mathcal{D}$  is convex.

#### (c)

Define  $\mathcal{B} = \{(t, y) | 0 \leq t \leq 1, y \in \mathbb{R}\}$ , it is convex. Define  $\Phi(t, y) = \frac{4t^3 y}{1 + t^4}$ , for all  $(t, y) \in \mathcal{B}$ ,

$$\left| \frac{\partial \Phi(t, y)}{\partial y} \right| = \frac{4t^3}{1 + t^4} \leq 4$$

and therefore  $\Phi$  satisfies a Lipschitz condition in  $y$  on  $\mathcal{B}$  with Lipschitz constant  $c = 4$ . So  $\dot{y} = \Phi(t, y)$  has a unique solution.

#### (d)

We can solve the differential equation analytically

$$\begin{aligned} \dot{y} = 1 + y^2, \quad y(0) = 0 &\Rightarrow \int_0^y \frac{1}{1 + u^2} du = \int_0^t 1 dx \\ &\Rightarrow \arctan(y) = t \end{aligned}$$

so we can see that for  $t \geq \frac{\pi}{2}$ , the ODE has no solution. So we cannot find the solution for  $0 \leq t \leq 3$  numerically.

If we use Euler's forward method,

$$\hat{y}(t_{i+1}) = \hat{y}(t_i) + h \cdot (1 + \hat{y}(t_i)^2)$$

it will not return our such kind of information. It seems to work, while actually, it has failed.

## 4.

### (a)

Define  $\mathcal{A} = \{(t, y) | t_0 \leq t \leq T, y \in \mathbb{R}\}$ , it is convex. Define  $\Phi(t, y) = \arctan(y)$ , for all  $(t, y) \in \mathcal{A}$ ,

$$\left| \frac{\partial \Phi(t, y)}{\partial y} \right| = \frac{1}{1 + y^2} \leq 1$$

and therefore  $\Phi$  satisfies a Lipschitz condition in  $y$  on  $\mathcal{A}$  with Lipschitz constant  $c = 1$ , i.e. a Lipschitz constant for  $\arctan(y)$  is  $c = 1$ .

### (b)

Take derivative on the ODE,

$$|\dot{y}| = \left| \frac{d \arctan(y)}{dx} \right| = \frac{1}{1 + y^2} \cdot |\dot{y}| = \frac{|\arctan(y)|}{1 + y^2} \leq \frac{\pi}{2}$$

So an upper bound on  $|\ddot{y}|$  is  $\frac{\pi}{2}$ .

### (c)

For the local discretisation error,

$$\begin{aligned} |\tau_k| &= |y(t_k) - (y(t_{k-1}) + h \cdot \arctan(y(t_{k-1})))| \\ &\leq |y(t_k) - y(t_{k-1})| + h \cdot |\arctan(y(t_{k-1}))| \\ &\leq 1 \cdot |t_k - t_{k-1}| + h \cdot \frac{\pi}{2} \\ &= (1 + \frac{\pi}{2})h \end{aligned}$$

then

$$|e_k| \leq \frac{|\tau^*|}{h \cdot 1} (e^{1 \cdot (t_k - t_0)} - 1) \leq (1 + \frac{\pi}{2})(e^{kh} - 1)$$

where  $h$  is the step size and  $k$  is the step number.



## 5.

### (a)

```

1 function [t, y] = Euler_f(Phi, t0, T, y0, n)
2     f = inline(Phi, 't', 'y');
3     h = (T - t0)/n;
4     t(1) = t0;
5     y(1) = y0;
6     for i = 2 : n + 1
7         t(i) = t(i - 1) + h;
8         y(i) = y(i - 1) + h * f(t(i - 1), y(i - 1));
9     end
10 end

```

Here is the running result of Euler's method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}(t)$	4.00E+00	1.20E+01	3.61E+01	1.10E+02	3.39E+02	1.07E+03	3.48E+03	1.17E+04
$t$	8	9	10	11	12	13	14	15
$\hat{y}(t)$	4.09E+04	1.49E+05	5.67E+05	2.27E+06	9.54E+06	4.24E+07	1.99E+08	9.86E+08

Table 1: Running result of Euler's method

### (b)

```

1 function [t, y] = Euler_b(Phi, t0, T, y0, n)
2     f = inline(Phi, 't', 'y');
3     h = (T - t0)/n;
4     t(1) = t0;
5     y(1) = y0;
6     for i = 2 : n + 1
7         t(i) = t(i - 1) + h;
8         y(i) = y(i - 1)/(1 - f(t(i), h));
9     end
10 end

```

Here is the running result of backward Euler's method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}(t)$	4.00E+00	-3.96E+00	3.81E+00	-3.49E+00	3.01E+00	-2.41E+00	1.77E+00	-1.19E+00
$t$	8	9	10	11	12	13	14	15
$\hat{y}(t)$	7.25E-01	-4.01E-01	2.00E-01	-9.06E-02	3.71E-02	-1.38E-02	4.66E-03	-1.44E-03

Table 2: Running result of backward Euler's method

### (c)

```

1 function [t, y] = Taylor_sec(phi, phi_t, phi_y, t0, T, y0, n)
2     Phi = inline(phi, 't', 'y');
3     Phi_t = inline(phi_t, 't', 'y');

```

```

4   Phi_y = inline(phi_y, 't', 'y');
5   t(1) = t0;
6   y(1) = y0;
7   h = (T - t0)/n;
8   for i = 2 : n + 1
9       t(i) = t(i - 1) + h;
10      p = Phi(t(i - 1), y(i - 1));
11      pt = Phi_t(t(i - 1), y(i - 1));
12      py = Phi_y(t(i - 1), y(i - 1));
13      y(i) = y(i - 1) + h * p + 1/2 * h * h * (pt + py * p);
14  end
15 end

```

Here is the running result of second-order Taylor's method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}(t)$	4.00E+00	2.00E+01	1.01E+02	5.18E+02	2.75E+03	1.52E+04	8.87E+04	5.50E+05
$t$	8	9	10	11	12	13	14	15
$\hat{y}(t)$	3.66E+06	2.64E+07	2.07E+08	1.78E+09	1.69E+10	1.77E+11	2.06E+12	2.66E+13

Table 3: Running result of second-order Taylor's method

(d)

```

1 function [t, y] = Henu(Phi, t0, T, y0, n)
2     h = (T - t0)/n;
3     t(1) = t0;
4     y(1) = y0;
5     f = inline(Phi, 't', 'y');
6     for i = 2 : n + 1
7         t(i) = t(i - 1) + h;
8         k1 = f(t(i - 1), y(i - 1));
9         ystar = y(i - 1) + h * k1;
10        k2 = f(t(i), ystar);
11        y(i) = y(i - 1) + 1/2 * h * (k1 + k2);
12    end
13 end

```

Here is the running result of Henu's method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}(t)$	4.00E+00	2.01E+01	1.02E+02	5.29E+02	2.85E+03	1.60E+04	9.56E+04	6.09E+05
$t$	8	9	10	11	12	13	14	15
$\hat{y}(t)$	4.17E+06	3.10E+07	2.52E+08	2.25E+09	2.21E+10	2.41E+11	2.93E+12	3.96E+13

Table 4: Running result of Henu's method

(e)

```

1 function [t, y] = Adams(Phi, t0, T, y0, n)
2     h = (T - t0)/n;
3     t(1) = t0;

```

```

4  y(1) = y0;
5  f = inline(Phi, 't', 'y');
6  t(2) = t0 + h;
7  k1 = f(t(1), y(1));
8  ystar = y(1) + h * k1;
9  k2 = f(t(2), ystar);
10 y(2) = y(1) + 1/2 * h * (k1 + k2);
11 for i = 3 : n + 1
12     t(i) = t(i - 1) + h;
13     y(i) = y(i - 1) + 1/2 * h * (3 * f(t(i - 1), y(i - 1)) - f(t(i - 2), y(i - 2)));
14 end
15 end

```

Here is the running result of two-step Adams-Bashforth method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}(t)$	4.00E+00	2.01E+01	7.65E+01	2.91E+02	1.12E+03	4.46E+03	1.83E+04	7.81E+04
$t$	8	9	10	11	12	13	14	15
$\hat{y}(t)$	3.48E+05	1.63E+06	8.04E+06	4.19E+07	2.32E+08	1.36E+09	8.49E+09	5.64E+10

Table 5: Running result of two-step Adams-Bashforth method

(f)

---

**Algorithm 4:** Second-order Runge-Kutta method using the mid-point rule

---

**Require:**  $\mathbf{A}$  has  $n$  different eignvalues

**Input:** function  $\Phi$ , endpoints  $t_0, T$ , initial condition  $y_0$ , number of steps  $n$

**Output:**  $t$  contains  $n + 1$  equally spaced mesh points, starting from  $t_0$  to  $T$   $y$  contains  $y_0$  and approximations to the solution at the corresponding  $t$

1 **Function** Runge\_Kutta\_mid ( $\Phi, t_0, T, y_0, n$ ):

2      $h \leftarrow (T - t_0)/n$ ;

3      $t[0] \leftarrow t_0$ ;

4      $y[0] \leftarrow y_0$ ;

5     **for**  $i \leftarrow 1$  **to**  $n$  **do**

6          $t[i] \leftarrow t[i - 1] + h$ ;

7          $k_1 \leftarrow \Phi(t[i - 1], y[i - 1])$ ;

8          $ystar \leftarrow y[i - 1] + h \cdot k_1$ ;

9          $ymid \leftarrow 1/2 \cdot (y[i - 1] + ystar)$ ;

10         $tmid \leftarrow t[i - 1] + h/2$ ;

11         $y[i] \leftarrow y[i - 1] + h \cdot \Phi(tmid, ymid)$ ;

12     **end**

13     **return** ( $t, y$ );

14 **end**

---

```

1 function [t, y] = Runge_Kutta_mid(Phi, t0, T, y0, n)
2     h = (T - t0)/n;
3     t(1) = t0;
4     y(1) = y0;
5     f = inline(Phi, 't', 'y');
6     for i = 2 : n + 1

```

```

7      t(i) = t(i - 1) + h;
8      k1 = f(t(i - 1), y(i - 1));
9      ystar = y(i - 1) + h * k1;
10     ymid = 1/2 * (y(i - 1) + ystar);
11     tmid = t(i - 1) + h/2;
12     y(i) = y(i - 1) + h * f(tmid, ymid);
13 end
14 end

```

Here is the running result of Runge-Kutta method using the mid-point rule, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}$	4.00E+00	2.00E+01	1.01E+02	5.23E+02	2.79E+03	1.56E+04	9.18E+04	5.77E+05
$t$	8	9	10	11	12	13	14	15
$\hat{y}$	3.90E+06	2.85E+07	2.27E+08	1.99E+09	1.92E+10	2.06E+11	2.44E+12	3.23E+13

Table 6: Running result of Runge-Kutta method using the mid-point rule

(g)

```

1 function [t, y] = Runge_Kutta_quad(Phi, t0, T, y0, n)
2     h = (T - t0)/n;
3     t(1) = t0;
4     y(1) = y0;
5     f = inline(Phi, 't', 'y');
6     for i = 2 : n + 1
7         t(i) = t(i - 1) + h;
8         k1 = f(t(i - 1), y(i - 1));
9         k2 = f(t(i - 1) + h/2, y(i - 1) + h/2 * k1);
10        k3 = f(t(i - 1) + h/2, y(i - 1) + h/2 * k2);
11        k4 = f(t(i - 1) + h, y(i - 1) + h * k3);
12        y(i) = y(i - 1) + h/6 * (k1 + k2 + k3 + k4);
13    end
14 end

```

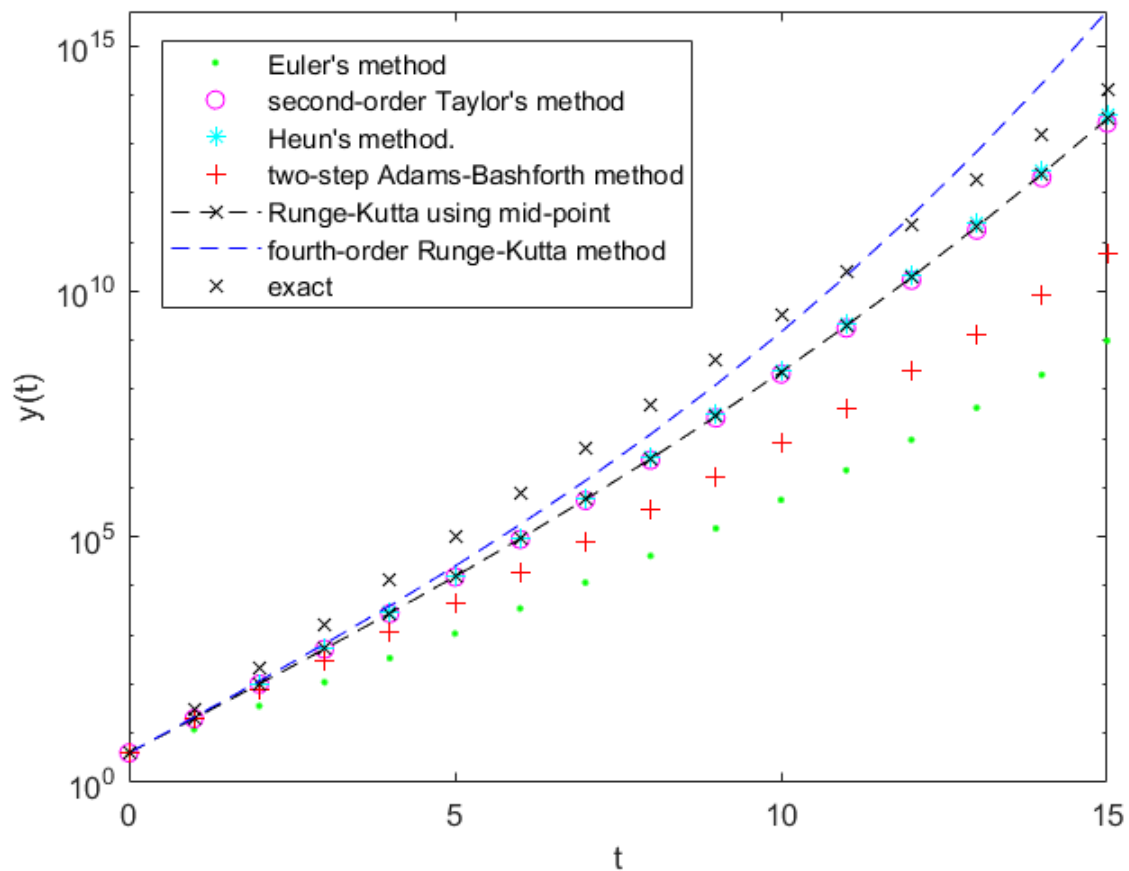
Here is the running result of fourth-order Runge-Kutta method, the data has been rounded to 2 digital numbers in scientific notation

$t$	0	1	2	3	4	5	6	7
$\hat{y}$	4.00E+00	2.14E+01	1.17E+02	6.61E+02	3.95E+03	2.54E+04	1.78E+05	1.39E+06
$t$	8	9	10	11	12	13	14	15
$\hat{y}$	1.23E+07	1.25E+08	1.48E+09	2.06E+10	3.43E+11	6.89E+12	1.68E+14	5.04E+15

Table 7: Running result of fourth-order Runge-Kutta method

(h)

All of the approximations and the exact solution is plotted on the figure below. The axis of  $y$  is in logarithm to show all lines clearly. And the approximation done by backward Euler's method is not shown on this figure, since there is obvious error.



(i)

```

1 function [P] = Newton_interpolation(t, y, n, x)
2     P = y(1);
3     for i = 2 : n
4         prod = divided_difference(1, i, y, t);
5         for k = 1 : i - 1
6             prod = prod * (x - t(k));
7         end
8         P = P + prod;
9     end
10 end
11
12 function [x] = divided_difference(i, k, y, t)
13     if k - i == 1
14         x = (y(k) - y(i)) / (t(k) - t(i));
15     else
16         x = (divided_difference(i + 1, k, y, t) - divided_difference(i, k - 1, y, t)) / (t(k) - t(i));
17     end
18 end

```

```
>> [t,y1]=Euler_f(' (2+0.01*t^2)*y', 0, 15, 4, 15);
>> Newton_interpolation(t, y1, 16, 9.625)

ans =

3.4085e+05
```

So the value of  $y$  at  $t = 9.625$  by using the approximation from Euler's method and interpolation in Newton's form is  $\hat{y}(9.625) = 3.4085 \times 10^5$ .

## 6.

Set  $y_1 = y, y_2 = \dot{y}, y_3 = \ddot{y}$ , then

$$\begin{pmatrix} \dot{y}_1 \\ \dot{y}_2 \\ \dot{y}_3 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -\frac{2}{t^3} & \frac{2}{t^2} & -\frac{1}{t} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ 8 - \frac{2}{t^3} \end{pmatrix}$$

with the initial condition

$$\begin{pmatrix} y_1(1) \\ y_2(1) \\ y_3(1) \end{pmatrix} = \begin{pmatrix} 2 \\ 8 \\ 6 \end{pmatrix}$$

Then we use Matlab to apply fourth-order Runge-Kutta method

```
1 clear all clc;
2 h = 0.1;
3 t(1) = 1;
4 n = (2 - 1)/h;
5 y1(1) = 2;
6 y2(1) = 8;
7 y3(1) = 6;
8 f = @(t,y1,y2,y3) [[0, 1, 0];[0, 0, 1];[-2/t^3, 2/t^2, -1/t]] * [y1; y2; y3]
   + [0; 0; 8 - 2/t^3];
9 for i = 2 : n + 1
10     t(i) = t(i - 1) + h;
11     k1 = f(t(i - 1), y1(i - 1), y2(i - 1), y3(i - 1));
12     k2 = f(t(i - 1) + h/2, y1(i - 1) + h/2 * k1(1), y2(i - 1) + h/2 * k1(2),
   y3(i - 1) + h/2 * k1(3));
13     k3 = f(t(i - 1) + h/2, y1(i - 1) + h/2 * k2(1), y2(i - 1) + h/2 * k2(2),
   y3(i - 1) + h/2 * k2(3));
14     k4 = f(t(i - 1) + h, y1(i - 1) + h * k3(1), y2(i - 1) + h * k3(2), y3(i
   - 1) + h * k3(3));
15     y1(i) = y1(i - 1) + h/6 * (k1(1) + k2(1) + k3(1) + k4(1));
16     y2(i) = y2(i - 1) + h/6 * (k1(2) + k2(2) + k3(2) + k4(2));
17     y3(i) = y3(i - 1) + h/6 * (k1(3) + k2(3) + k3(3) + k4(3));
18 end
```

and we find the approximation to solution is (results have been rounded to 2 digital numbers)

$t$	1	1.1	1.2	1.3	1.4	1.5	1.6	1.7	1.8	1.9	2
$\hat{y}(t)$	2.00	2.55	3.14	3.76	4.42	5.11	5.85	6.63	7.46	8.34	9.26

Table 8: Approximation to solution by using fourth-order Runge-Kutta method

compared with the exact solution, we can see the error is listed in below

$t$	1	1.1	1.2	1.3	1.4	1.5	1.6	1.7	1.8	1.9	2
$\hat{y}(t)$	2.00	2.55	3.14	3.76	4.42	5.11	5.85	6.63	7.46	8.34	9.26
$y(t)$	2.00	2.83	3.73	4.72	5.79	6.96	8.23	9.61	11.12	12.74	14.50
relative error(%)	0.00	-9.79	-15.90	-20.27	-23.68	-26.50	-28.90	-31.01	-32.89	-34.59	-36.14

## 7.

### (a)

Use test function  $u(x)$  which satisfies that  $u(1) = u(3) = u'(1) = u'(3) = 0$ ,

$$\begin{aligned}
 & \int_1^3 u(x) \cdot x^3 y^{(4)} dx + \int_1^3 u(x) \cdot 6x^2 y^{(3)} dx + \int_1^3 u(x) \cdot 6xy'' dx - \int_1^3 u(x) \cdot 10x dx = 0 \\
 \Rightarrow & u(x) \cdot (x^3 y'')'|_1^3 - \int_1^3 u'(x) \cdot (x^3 y'')' dx = \int_1^3 u(x) \cdot 10x dx \\
 \Rightarrow & -u'(x) \cdot (x^3 y'')'|_1^3 + \int_1^3 u''(x) \cdot (x^3 y'') dx = \int_1^3 u(x) \cdot 10x dx \\
 \Rightarrow & \int_1^3 x^3 y''(x) u''(x) dx = \int_1^3 u(x) \cdot 10x dx
 \end{aligned}$$

### (b)

Assume  $u(x) = y(x)$ , and set the energy function

$$I[u] = \int_1^3 x^3 (u''(x))^2 dx - 2 \int_1^3 u(x) \cdot 10x dx$$

Let  $\hat{y}(x) = \sum_{j=1}^n c_j \phi_j(x)$ , substituting  $\hat{y}(x)$  in to  $I[u]$ , we have

$$I = \int_1^3 x^3 \left( \sum_{j=1}^n c_j \phi_j''(x) \right)^2 dx - 2 \int_1^3 \sum_{j=1}^n c_j \phi_j(x) \cdot 10x dx$$

then

$$\frac{\partial I}{\partial c_i} = \int_1^3 x^3 (2\phi_i''(x) \sum_{j=1}^n c_j \phi_j''(x)) dx - 2 \int_1^3 \phi_i(x) \cdot 10x dx$$

set it to zero and we obtain a set of equations  $\mathbf{Ax} = \mathbf{b}$ , where

$$a_{ij} = \int_1^3 x^3 \phi_i''(x) \phi_j''(x) dx, \quad b_i = \int_1^3 10x \phi_i(x) dx$$

Choose  $\phi(x)$  as

$$\phi_i(x) = \begin{cases} 0 & , x \in [1, x_{i-1}] \\ \frac{(x - x_{i-1})(x_{i+1} - x)}{(x_i - x_{i-1})(x_{i+1} - x_i)} & , x \in (x_{i-1}, x_{i+1}] \\ 0 & , x \in (x_{i+1}, 3] \end{cases} \Rightarrow \phi_i''(x) = \begin{cases} 0 & , x \in [1, x_{i-1}] \\ -\frac{2}{(x_i - x_{i-1})(x_{i+1} - x_i)} & , x \in (x_{i-1}, x_{i+1}] \\ 0 & , x \in (x_{i+1}, 3] \end{cases}$$

then

$$\begin{aligned} a_{ii} &= \int_1^3 x^3 \left( \frac{2}{(x_i - x_{i-1})(x_{i+1} - x_i)} \right)^2 dx \\ &= \int_{x_{i-1}}^{x_{i+1}} x^3 \left( \frac{2}{(x_i - x_{i-1})(x_{i+1} - x_i)} \right)^2 dx \\ &= \frac{x_{i+1}^4 - x_{i-1}^4}{(x_i - x_{i-1})^2 (x_{i+1} - x_i)^2} \\ a_{i,i+1} &= \int_1^3 x^3 \frac{2}{(x_i - x_{i-1})(x_{i+1} - x_i)} \frac{2}{(x_{i+1} - x_i)(x_{i+2} - x_{i+1})} dx \\ &= \int_{x_i}^{x_{i+1}} x^3 \frac{4}{(x_i - x_{i-1})(x_{i+1} - x_i)(x_{i+1} - x_i)(x_{i+2} - x_{i+1})} dx \\ &= \frac{x_{i+1}^4 - x_i^4}{(x_i - x_{i-1})(x_{i+1} - x_i)(x_{i+1} - x_i)(x_{i+2} - x_{i+1})} \end{aligned}$$

else  $a_{ij} = 0$ . And

$$\begin{aligned} b_i &= \int_1^3 10x\phi_i(x)dx = \int_{x_{i-1}}^{x_{i+1}} 10x \frac{(x - x_{i-1})(x_{i+1} - x)}{(x_i - x_{i-1})(x_{i+1} - x_i)} dx \\ &= \frac{\frac{5}{6}(x_{i+1}^4 - x_{i-1}^4) - \frac{5}{3}x_{i-1}x_{i+1}(x_{i+1}^2 - x_{i-1}^2)}{(x_i - x_{i-1})(x_{i+1} - x_i)} \end{aligned}$$

(c)

To find the exact solution,

$$\begin{aligned} x^3 y^{(4)} + 6x^2 y^{(3)} + 6xy'' - 10x &= 0 \\ \Leftrightarrow (x^3 y'')'' &= 10x \\ \Rightarrow x^3 y'' &= \frac{5}{3}x^3 + c_1 x + c_2 \\ \Rightarrow y'' &= \frac{5}{3} + \frac{c_1}{x^2} + \frac{c_2}{x^3} \\ \Rightarrow y' &= \frac{5}{3}x - \frac{c_1}{x} - \frac{c_2}{2x^2} + c_3 \\ \Rightarrow y &= \frac{5}{6}x^2 - c_1 \ln x + \frac{c_2}{2x} + c_3 x + c_4 \end{aligned}$$

Since  $y(1) = y(3) = y'(1) = y'(3) = 0$ ,

$$c_1 = -16.9012, \quad c_2 = 17.8518, \quad c_3 = -9.6420, \quad c_4 = -0.1173$$



