

1 Question 1

* All codes are writtern and compiled by Matlab R2017b.

1.1

Function for solving an $n \times n$ tridiagonal matrix.

```
1  %Solving linear system Ax=(abc)x=d;
2  function x = tridisolve(a,b,c,d)
3  x = d;
4  n = length(x);
5
6  for j = 1:n-1
7  mu = a(j)/b(j);
8  b(j+1) = b(j+1) - mu*c(j);
9  x(j+1) = x(j+1) - mu*x(j);
10 end
11
12 x(n) = x(n)/b(n);
13 for j = n-1:-1:1
14 x(j) = (x(j)-c(j)*x(j+1))/b(j);
15 end
16
17 end
```

Where

$$A = \begin{pmatrix} b_1 & c_1 & & & \\ a_1 & b_2 & c_2 & & \\ & a_2 & b_3 & c_3 & \\ & & \ddots & \ddots & \ddots \\ & & & a_{n-2} & b_{n-1} & c_{n-1} \\ & & & & a_{n-1} & b_n \end{pmatrix} \quad (1)$$

and $a = (a_1, \dots, a_{n-1})$, $b = (b_1, \dots, b_n)$, $c = (c_1, \dots, c_{n-1})$

1.2

Function for Computing the coefficients $\{c_i\}_{i=-1}^{n+1}$ of the natural C^2 -Cubic B-spline $q_3^n(x)$ that interpolates f

```
1  %Function f
2  function F=func(x)
3
4  F=exp(-x).*cos(6*pi*x);
5
6  end
```

where F would be a vector such that $F(x) = (F(x_1), \dots, F(x_d)), x_i \in R$.

```

1      %Computing coefficients
2      function C= Bspline(f)
3      n=length(f);
4      trid=@tridisolve;
5      a=zeros(1,n-3);
6      b=zeros(1,n-2);
7      c=zeros(1,n-3);
8      d=zeros(1,n-2);
9      a(1:n-3)=1;
10     b(1:n-2)=4;
11     c(1:n-3)=1;
12
13     d(1)=f(2)-f(1)/6;%f1-f0/6
14     d(2:n-3)=f(3:n-2);
15     d(n-2)=f(n-1)-f(n)/6;
16
17     C=zeros(1,n+2);
18     C(3:n)=trid(a,b,c,d);
19     C(2)=f(1)/6; %C0
20     C(1)=2*C(2)-C(3); %first-derivative conditions ,C-1
21     C(n+1)=f(n)/6;
22     C(n+2)=2*C(n+1)-C(n);
23     end

```

1.3

Function for computing values of the spline $q_3^n(x)$

```

1      function [xhat,q]= evaluate(Coef,xnode,a,b)
2      B=@splinesfunc;
3      n=length(xnode);%index of xi is from 0 to n-1
4      h=xnode(2)-xnode(1);
5      distance=(b-a)/(20*n-20);%20*(n-1)+1 point with same distances
6
7      xhat=zeros(1,20*n-19);
8      q=zeros(1,20*n-19);
9      xhat(1)=xnode(1);
10     xhat(end)=xnode(end);
11     q(1)=Coef(1)+4*Coef(2)+Coef(3); %condition for q(x) in node points
12     q(end)=Coef(end-2)+4*Coef(end-1)+Coef(end);
13
14     for i=2:20*n-20
15         xhat(i)=xhat(i-1)+distance;
16         k=floor((xhat(i)-xnode(1))/h)+3; %location of xhat(i)=k-2
17         if (k<4)
18             %B-1(x)=B((x-x0+h)/h)
19             q(i)=Coef(1)*B((xhat(i)-xnode(1)+h)/h)+Coef(2)*B((xhat(i)-xnode(1))/h)
20             +Coef(3)*B((xhat(i)-xnode(2))/h)+Coef(4)*B((xhat(i)-xnode(3))/h);
21         elseif (k>n)
22             %B(n+1)(x)=B((x-xn-h)/h)
23             q(i)=Coef(k-2)*B((xhat(i)-xnode(k-3))/h)+Coef(k-1)*B((xhat(i)-xnode(k-2))/h)
24             +Coef(k)*B((xhat(i)-xnode(k-1))/h)+Coef(k+1)*B((xhat(i)-xnode(k-1)-h)/h);
25         else
26             q(i)=Coef(k-2)*B((xhat(i)-xnode(k-3))/h)+Coef(k-1)*B((xhat(i)-xnode(k-2))/h)
27             +Coef(k)*B((xhat(i)-xnode(k-1))/h)+Coef(k+1)*B((xhat(i)-xnode(k))/h);
28         end
29     end

```

where B is one of the spline function

$$B(x) = \begin{cases} 0, & x \leq -2 \\ (x+2)^3, & -2 \leq x \leq -1 \\ 1+3(x+1)+3(x+1)^2-3(x+1)^3, & -1 \leq x \leq 0 \\ 1+3(1-x)+3(1-x)^2-3(1-x)^3, & 0 \leq x \leq 1 \\ (2-x)^3, & 1 \leq x \leq 2 \\ 0, & 2 \leq x \end{cases} \quad (2)$$

```

1 %Spline function B
2 function B=splinefunc(x)
3 if (x<=-2)
4 B=0;
5 elseif (-2<=x && x<=-1)
6 B=(x+2)^3;
7 elseif (-1<=x && x<=0)
8 B=1+3*(x+1)+3*(x+1)^2-3*(x+1)^3;
9 elseif (0<=x && x<=1)
10 B=1+3*(1-x)+3*(1-x)^2-3*(1-x)^3;
11 elseif (1<=x && x<=2)
12 B=(2-x)^3;
13 else
14 B=0;
15 end
16 end

```

1.4

Table 1 illustrates $\|f - q_3^n\|_{\infty, [-1,1]}$ with different n and Figure 1 shows $\log_{10}(\|f - q_3^n\|_{\infty, [-1,1]})$ against $\log_{10}(n)$.

	n=16	n=32	n=64	n=128
$\ f - q_3^n\ _{\infty, [-1,1]}$	1.087447370563231	0.216700752320041	0.048126171304783	0.011648941831482

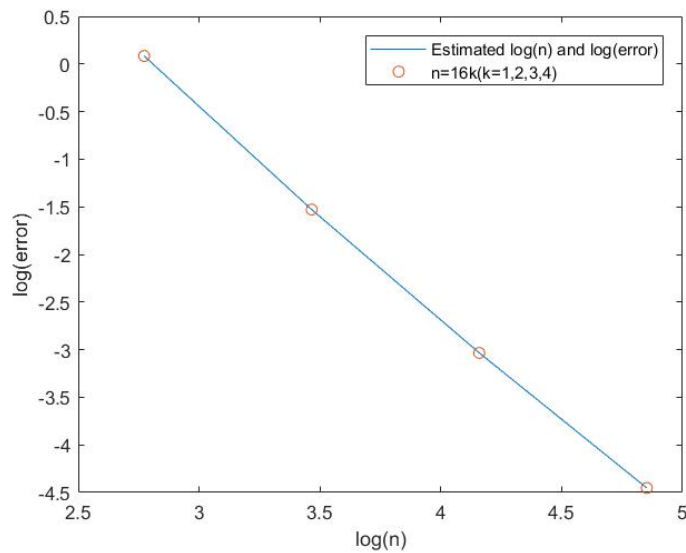


Figure 1: $\log(n)$ and $\log(\text{error})$

As Figure 1 shows, the relationship between $\log_{10}(\|f - q_3^n\|_{\infty, [-1,1]})$ and $\log_{10}(n)$ is almost linear. Moreover, we will get a more accurate solution when test multiple n by using OLS regression. By testing $n = 16, 18, 20, \dots, 128$, we get the relationship between actual rate of convergence($\log(\text{error})$) and $\log(n)$.

$$\log_{10}(\|f - q_3^n\|_{\infty, [-1,1]}) = -2.158 \log_{10}(n) + 5.972 \quad (3)$$

Theoretically, $\log_{10}(\|f - q_3^n\|_{\infty, [-1,1]})$ satisfies

$$\log_{10}(\|f - q\|_{\infty}) \leq \log_{10}\left(\frac{5}{384}h^4\|f^{(4)}\|_{\infty}\right) = \log_{10}\left(\frac{5}{384}\frac{(b-a)^4}{n^4}\|f^{(4)}\|_{\infty}\right) = -4\log_{10}(n) + \log C \quad (4)$$

where C is a constant number.

Figure 2,3,4 and 5 show the trajectories of $q_3^n(x)$ with $n=16, 32, 64, 128$ respectively.

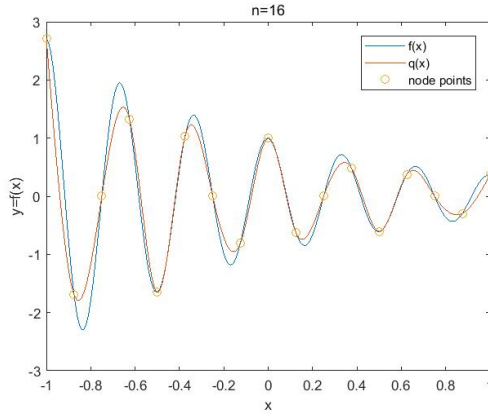


Figure 2: $n=16$

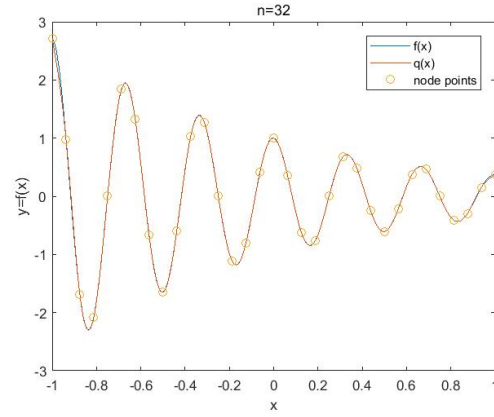


Figure 3: $n=32$

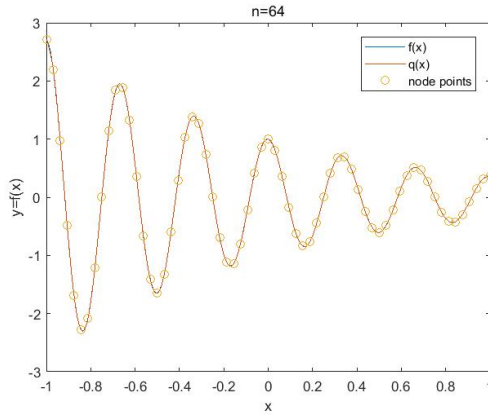


Figure 4: $n=64$

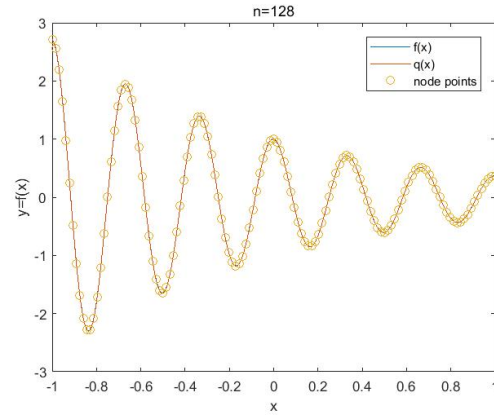


Figure 5: $n=128$

2 Question 2

2.1

Function for integrating a function using a Gauss–Legendre quadrature with $n = 10$

```
1  %function to integrate a function using a -GaussLegendre quadrature
2  function G=Gauss(f,n) %f is a vector valued function with dimension n
3
4  G=zeros(n,1);
5  %first column are weights and second one are quadrature points
6  Gaussnode=[0.2955242247147529, -0.1488743389816312;
7  0.2955242247147529, 0.1488743389816312;
8  0.2692667193099963, -0.4333953941292472;
9  0.2692667193099963, 0.4333953941292472;
10 0.2190863625159820, -0.6794095682990244;
11 0.2190863625159820, 0.6794095682990244;
12 0.1494513491505806, -0.8650633666889845;
13 0.1494513491505806, 0.8650633666889845;
14 0.0666713443086881, -0.9739065285171717;
15 0.0666713443086881, 0.9739065285171717];
16
17 G=f(Gaussnode(:,2)',n)*Gaussnode(:,1);
18 end
```

We take testgaussfunc as example(note that $n=1$)

```
1  %Vector-valued function to do Gauss integrate
2  function F=testguassfunc(x,n)
3  dimension=length(x); %Dimension of vector x
4
5  F=zeros(n,dimension);
6  if n==1
7  F(1,:)=exp(-x.^2);
8  end
```

Then we get the approximation

$$\int_{-1}^1 e^{-x^2} dx \approx 1.493648265624351 \quad (5)$$

And the error is

$$err = |\sqrt{\pi} \operatorname{erf}(1) - 1.493648265624351| = 5.033751193650460e - 13 \quad (6)$$

2.2

Function that will compute the LSQ coefficients for a given f and n

```
1  %Compute the LSQ coefficients
2  function C=Lsqcoef(Func,n)
3  G=@Gauss;
4  C=G(Func,n); %Compute inner product(f(x),Pi(x))
5  for i=1:n+1
6  C(i)=(2*i-1)*C(i)/2; %inner product(Pk(x),Pk(x)=2/(2k+1)
7  end
```

Where Func is Vector-valued function such that $Func(x) = f(x)(P_1(x), \dots, P_n(x))^T$ where $P(x)$ is Legendre polynomials and inner product is computing by Gauss-integrate which is introduced in 2.1.

```

1  %F is vector-valued function that F(i)=f(x)*Pi(x) where Pi(x) is Legendre
2  function F=Func(x,n)
3  P=@Legendre;
4  f=@func;
5  dimension=length(x);%length of vector x
6  F=zeros(n+1,dimension);
7  z=P(x,n);
8  for i=1:n+1
9  F(i,:)=f(x).*z(i,:);%F(i,:)=(Fi(x0),Fi(x1),...,Fi(xd))
10 end
11 end

```

And Legendre polynomial

```

1  %Vector-valued function that P(i)=Pi(x) where Pi(x) is Legendre
2  function P=Legendre(x,n)
3  dimension=length(x);
4  P=zeros(n+1,dimension);
5  P(1,:)=1;
6  P(2,:)=x;
7  for i=3:n+1
8  %P(i,:)=(Pi(x0),Pi(x1),...,Pi(xd))
9  P(i,:)=(2*i-3)*x.*P(i-1,:)/(i-1)-(i-2)*P(i-2,:)/(i-1);
10 end
11 end

```

Actually, we shall get a matrix with given vector x in both codes above, which means

$$P(x) = \begin{pmatrix} P_0(x_0) & P_0(x_1) & \dots & P_0(x_d) \\ P_1(x_0) & P_1(x_1) & \dots & P_1(x_d) \\ \dots & \dots & \dots & \dots \\ P_n(x_0) & P_n(x_1) & \dots & P_n(x_d) \end{pmatrix} \quad (7)$$

and

$$F(x) = \begin{pmatrix} f(x_0)P_0(x_0) & f(x_1)P_0(x_1) & \dots & f(x_d)P_0(x_d) \\ f(x_0)P_1(x_0) & f(x_1)P_1(x_1) & \dots & f(x_d)P_1(x_d) \\ \dots & \dots & \dots & \dots \\ f(x_0)P_n(x_0) & f(x_1)P_n(x_1) & \dots & f(x_d)P_n(x_d) \end{pmatrix} \quad (8)$$

Then we can easily get LSQ approximation with given vector a

$$LSQ(a) = C * P(a) = (C_0, C_1, \dots, C_n) * P(a) \quad (9)$$

2.3

Test the code for $f(x) = e^{-x} \cos(\pi x)$

```

1  %Function to test the LSQ
2  function f=func(x)
3  f=exp(-x).*cos(pi*x);
4  end

```

Figure 6,7 and 8 shows the results with different $n=3,5,7$ respectively.

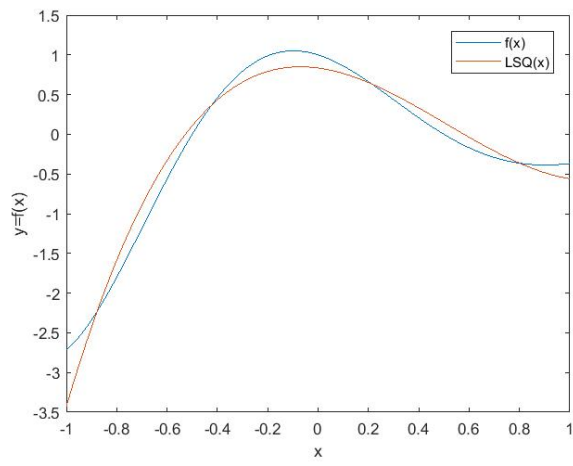


Figure 6: $n=3$

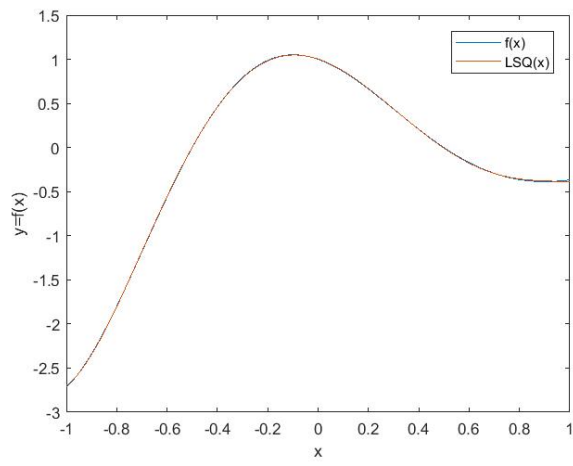


Figure 7: $n=5$

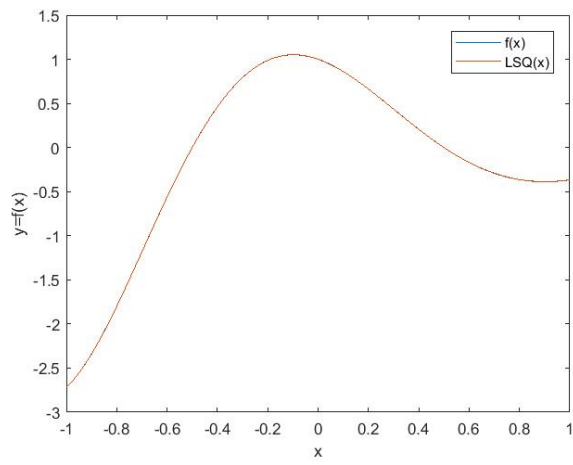


Figure 8: $n=7$

3 Question 3

3.1

Function to compute the Jacobian of a smooth vector-valued function F.

```
1 function J=Jaccobi(Func,x,epsilon)
2 n=length(x);
3 J=zeros(n,n);
4 deltax=zeros(1,n);
5 for i=1:n
6 deltax(i)=epsilon; %deltax=x+epsilon*ej
7 J(:,i)=((Func(x+deltax)-Func(x))/epsilon)';
8 deltax(i)=0;
9 end
10 end
```

3.2

Function to solve the k-dimension nonlinear system of equations using Newton's method to get $(n-1)^{th}$ iteration.

```
1 function X=Newton(n,Func,x0,epsilon)
2 J=@Jaccobi;
3 k=length(x0);
4 Xs=zeros(n,k);
5 Xs(1,:)=x0;
6 for i=2:n
7 Xs(i,:)=Xs(i-1,:)+(J(Func,Xs(i-1,:),epsilon)\(-Func(Xs(i-1,:))))';
8 end
9 X=Xs(end,:); %(n-1)th iteration
10 end
```

Test the code with a 2-dimension function.

```
1 function F=Func(x) %x is a two-dimension vector
2 F=zeros(2,1);
3 F(1)=x(1)^2-x(2);
4 F(2)=x(1)^2+x(2)^2-2;
5 end
```

Table 2 shows the fifth iteration $(x_1^{(5)}, x_2^{(5)})$ with different \mathcal{E} .

	$\mathcal{E}=0.005$	$\mathcal{E}=0.05$	$\mathcal{E}=0.5$
$x_1^{(5)}$	1.00000	1.000001554936712	1.001443216203944
$x_2^{(5)}$	1.00000	1.000000088939143	1.000164972604241