

Final Project Report

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1

当 $\alpha = 0$ 时, 记

$$\min_{\{u_1, u_2, \dots, u_{n-1}\}} \sum_{i=1}^n \left(\frac{1}{2} \left(\frac{u_i - u_{i-1}}{h} \right)^2 h - f_i u_i h \right) = \min_{\{u_1, u_2, \dots, u_{n-1}\}} G_n(u_1, u_2, \dots, u_{n-1})$$

求 G 对每个变量的偏微分, 并令它们为0, 得到

$$\begin{aligned} \frac{\partial G}{\partial u_1} &= u_1 - u_0 - f_1 h^2 + u_1 - u_2 = 0 \\ \frac{\partial G}{\partial u_2} &= u_2 - u_1 - f_2 h^2 + u_2 - u_3 = 0 \\ &\vdots \\ \frac{\partial G}{\partial u_{n-1}} &= u_{n-1} - u_{n-2} - f_{n-1} h^2 + u_{n-1} - u_n = 0 \end{aligned}$$

其中 $u_0 = u_n = 0$, 因此化简得

$$A_h u_h = f_h$$
$$\begin{pmatrix} \frac{2}{h^2} & -\frac{1}{h^2} & 0 & 0 & \cdots & 0 \\ -\frac{1}{h^2} & \frac{2}{h^2} & -\frac{1}{h^2} & 0 & \cdots & 0 \\ 0 & -\frac{1}{h^2} & \frac{2}{h^2} & -\frac{1}{h^2} & \cdots & 0 \\ 0 & 0 & -\frac{1}{h^2} & \frac{2}{h^2} & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & -\frac{1}{h^2} \\ 0 & 0 & 0 & 0 & -\frac{1}{h^2} & \frac{2}{h^2} \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ \vdots \\ \vdots \\ u_{n-1} \end{pmatrix} = \begin{pmatrix} f_1 \\ f_2 \\ \vdots \\ \vdots \\ \vdots \\ f_{n-1} \end{pmatrix}$$

2

为节省空间, 求解结果放在附录中

取 $e_h = ||u_h - u_e||_2$, $\epsilon = 10^{-10}$, $n = \{10, 20, 40, 80, 160\}$, Jacobi和Gauss-Seidel迭代法求解上述线性方程组误差列表为

	Jacobi iteration error	Gauss-Seidel iteration error
n=10	0.018482034928874364	0.018482032197992628
n=20	0.006510206868033681	0.006510189798674751
n=40	0.002299599922536855	0.002299502237862865
n=80	0.000812978994829882	0.000812425160801029
n=160	0.000288183348886839	0.000285044509734070

3

作log变换，则有 $\log(e_h) = \beta \log(h) + b$ 。

同时，2中相同n下Jacobi和Gauss-Seidel产生的不同 e_h 值取平均值，产生 n 和 $(\overline{e_h})$ 的一一对应关系

h	$\log(h)$	e_h	$\log(e_h)$
0.1	-2.30258509299404	0.018482034928874364	-3.99095610363200
0.1	-2.30258509299404	0.018482032197992628	-3.99095625139072
0.05	-2.99573227355399	0.006510206868033681	-5.03438404630283
0.05	-2.99573227355399	0.006510189798674751	-5.03438666824429
0.025	-3.68887945411393	0.002299599922536855	-6.07502011790062
0.025	-3.68887945411393	0.002299502237862865	-6.07506259778935
0.0125	-4.38202663467388	0.000812978994829882	-7.11480528536784
0.0125	-4.38202663467388	0.000812425160801029	-7.11548675778609
0.00625	-5.07517381523382	0.000288183348886839	-8.15191365231071
0.00625	-5.07517381523382	0.000285044509734070	-8.16286521538408

记 $x = \log(h), y = \log(e_h)$ ，作最小二乘法：

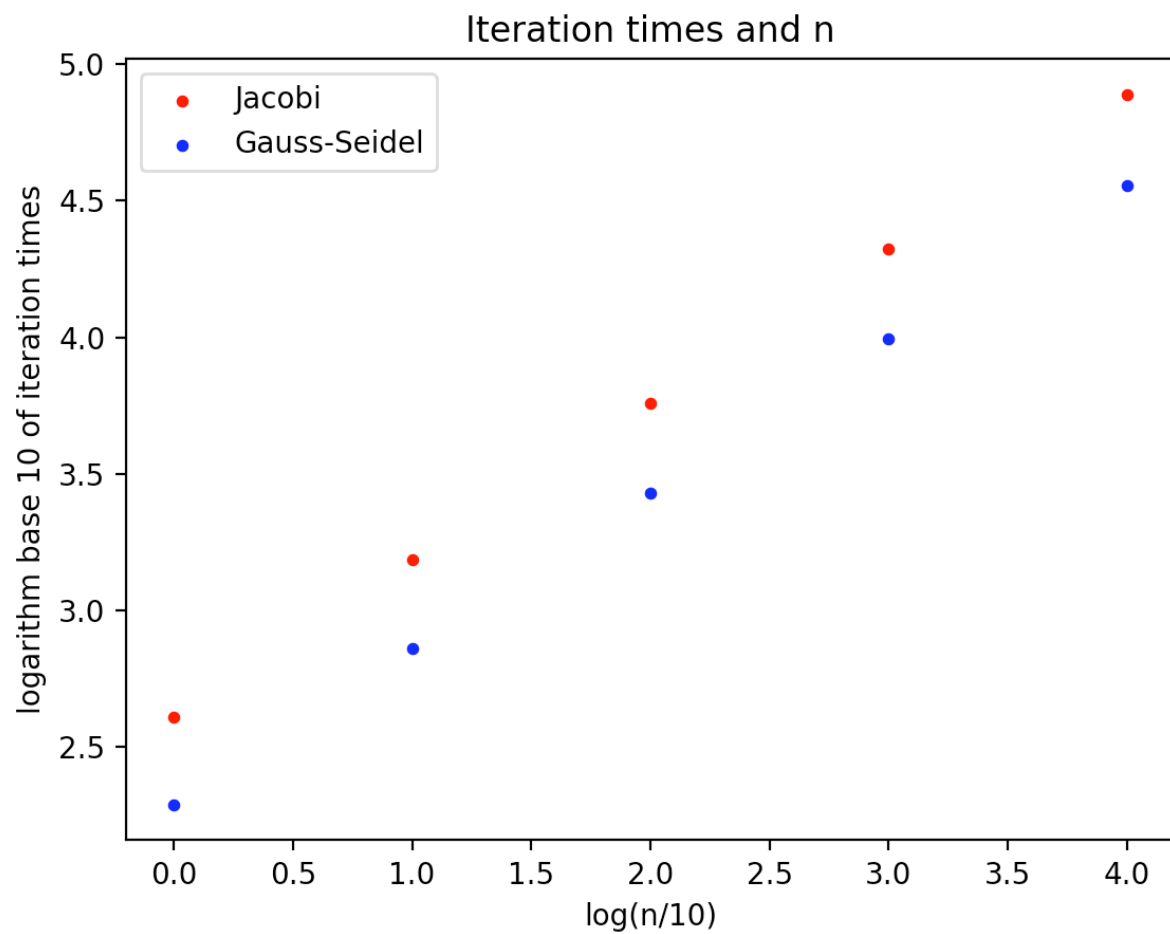
$$\begin{aligned} \sum xy &= 238.52042154548081 \\ \overline{x} &= -3.688879454113932 \\ \overline{y} &= -6.0745836696108535 \\ \sum x_i^2 &= 145.68737654820305 \\ \hat{\beta} &= \frac{\sum xy - 10\overline{x}\overline{y}}{\sum x_i^2 - 10\overline{x}^2} = 1.50237 \end{aligned}$$

2中Jacobi和Gauss-Seidel迭代法求解上述线性方程组到收敛的迭代次数分别为

	Jacobi iteration times	Gauss-Seidel iteration times
n=10	408	195
n=20	1539	729
n=40	5727	2697
n=80	21127	9891
n=160	77331	35970

观察结论：

- 相同 n 下，Jacobi迭代次数高于Gauss-Seidel
- 迭代次数随 n 增大而增大，且和 n 呈超线性关系：



为节省空间，求解结果放在附录中

当 $\alpha = 1$ 时，记

$$\min_{\{u_1, u_2, \dots, u_{n-1}\}} \sum_{i=1}^n \left(\frac{1}{2} \left(\frac{u_i - u_{i-1}}{h} \right)^2 h + \frac{\alpha}{4} u_i^4 h - f_i u_i h \right) = \min_{\{u_1, u_2, \dots, u_{n-1}\}} G_n(u_1, u_2, \dots, u_{n-1})$$

求 G 对每个变量的偏微分，并令它们为0，得到

$$\begin{aligned} \frac{\partial G}{\partial u_1} &= u_1 - u_0 + u_1^3 h^2 - f_1 h^2 + u_1 - u_2 = 0 \\ \frac{\partial G}{\partial u_2} &= u_2 - u_1 + u_2^3 h^2 - f_2 h^2 + u_2 - u_3 = 0 \\ &\vdots \\ \frac{\partial G}{\partial u_{n-1}} &= u_{n-1} - u_{n-2} + u_{n-1}^3 h^2 - f_{n-1} h^2 + u_{n-1} - u_n = 0 \end{aligned}$$

使用牛顿迭代法求解结果与精确解 $u_e(x) = \sin(\pi x)$ 的误差和 n 的关系见下表：

n	e_h
10	0.01501803282490173
20	0.00530088945911111
40	0.00187336860374472
80	0.00066226693305820
160	0.00023414062963731

同理3进行最小二乘法拟合：

h	$\log(h)$	e_h	$\log(e_h)$
0.1	-2.30258509299404	0.01501803282490173	-4.198503611602896
0.05	-2.99573227355399	0.00530088945911111	-5.239880650030883
0.025	-3.68887945411393	0.00187336860374472	-6.280017076300147
0.0125	-4.38202663467388	0.00066226693305820	-7.319841861157172
0.00625	-5.07517381523382	0.00023414062963731	-8.359588641767521

记 $x = \log(h)$, $y = \log(e_h)$ ，作最小二乘法：

$$\begin{aligned} \sum xy &= 123.03302474409142 \\ \bar{x} &= -3.6888794541139327 \\ \bar{y} &= -6.279566368171724 \\ \sum x_i^2 &= 72.84368827410151 \end{aligned}$$

$$\hat{\beta} = \frac{\sum xy - 5\bar{x}\bar{y}}{\sum x_i^2 - 5\bar{x}^2} = 1.50237$$

因此收敛阶仍约为1.5

附录

1. Jacobi和Gauss-Seidel迭代Python代码

```

1  from math import pi, sin, sqrt
2  import numpy as np
3
4  def gauss_seidel_iteration(A: np.array, b: np.array, epsilon: float):
5      iter_times = 0
6      n = len(A)
7      x_1 = np.zeros(n)
8      x_2 = np.ones(n)
9      while True:
10         iter_times += 1
11         x_1 = x_2.__copy__()
12         for i in range(n):
13             s = 0
14             for j in range(n):
15                 s += A[i][j] * x_2[j]
16             x_2[i] = (b[i] - s + A[i][i] * x_2[i]) / A[i][i]
17             if np.max(np.abs(x_2 - x_1)) < epsilon:
18                 break
19         return x_2, iter_times
20
21
22 def Jacobi_iteration(A: np.array, b: np.array, epsilon: float):
23     iter_times = 0
24     x_1 = np.zeros(len(A))
25     x_2 = np.ones(len(A))
26     while True:
27         iter_times += 1
28         x_1 = x_2.__copy__()
29         for i in range(len(A)):
30             s = 0
31             for j in range(len(A)):
32                 s += A[i][j] * x_1[j]
33             x_2[i] = (b[i] - s + A[i][i] * x_1[i]) / A[i][i]
34             if np.max(np.abs(x_2 - x_1)) < epsilon:
35                 break
36         return x_2, iter_times
37
38 def run_question_2(n: int, print_result: bool = False):

```

```

39     def f(x):
40         return pi * pi * sin(pi*x)
41     def u_e(x):
42         return sin(pi*x)
43     A = 2 * np.eye(n-1)
44     for i in range(n-2):
45         A[i][i+1] = -1
46         A[i+1][i] = -1
47     b = [f(i/n)/n/n for i in range(1, n)]
48     u_h_Jacobi, iter_times_Jacobi = Jacobi_iteration(A, b, 1e-10)
49     e_h_Jacobi = sqrt(sum([(u_h_Jacobi[i] - u_e((i+1)/n))**2 for i in range(n-
50 1)]))
51     u_h_Gauss, iter_times_Gauss = gause_seidel_iteration(A, b, 1e-10)
52     e_h_Gauss = sqrt(sum([(u_h_Gauss[i] - u_e((i+1)/n))**2 for i in range(n-1)]))
53     if print_result:
54         print(f"n={n}, Jacobi: {u_h_Jacobi}")
55         print(f"n={n}, Gauss Seidel: {u_h_Gauss}, Error: {e_h_Gauss}")
56         print(f"n={n}, Jacobi Error: {e_h_Jacobi}")
57         print(f"n={n}, Gauss Seidel Error: {e_h_Gauss}")
58         return (e_h_Jacobi, e_h_Gauss), (iter_times_Jacobi, iter_times_Gauss)
59
60 question_2_n = [10, 20, 40, 80, 160]
61 question_2_errors = []
62 question_2_iters = []
63 for n in question_2_n:
64     error, iters = run_question_2(n, True)
65     question_2_errors.append(error)
66     question_2_iters.append(iters)
67 print("Question 2 error table:", question_2_errors)
68 print("Question 2 iter times table:", question_2_iters)
69

```

2. Jacobi和Gauss- Seidel迭代输出解

```

1  n=10, Jacobi: [0.31157115 0.59264354 0.81570386 0.9589174 1.00826542 0.9589174
2  0.81570386 0.59264354 0.31157115]
3  n=10, Gauss Seidel: [0.31157115 0.59264354 0.81570386 0.9589174 1.00826542
4  0.9589174
5  0.81570386 0.59264354 0.31157115], Error: 0.018482036208807052
6  n=20, Jacobi: [0.15675652 0.30965317 0.45492513 0.58899533 0.70856251 0.81068252
7  0.89284085 0.95301446 0.9897217 1.00205871 0.9897217 0.95301446
8  0.89284085 0.81068252 0.70856251 0.58899533 0.45492513 0.30965317
9  0.15675652]
10 n=20, Gauss Seidel: [0.15675652 0.30965317 0.45492514 0.58899533 0.70856251
11 0.81068253
12 0.89284085 0.95301447 0.98972171 1.00205871 0.98972171 0.95301447

```

```

12 0.89284085 0.81068253 0.70856251 0.58899533 0.45492513 0.30965317
13 0.15675652], Error: 0.006510214971563716
14
15 n=40, Jacobi: [0.07849944 0.1565149 0.2335654 0.30917589 0.38288021 0.45422394
16 0.52276724 0.5880875 0.649782 0.70747038 0.76079697 0.809433
17 0.8530786 0.89146469 0.9243546 0.95154556 0.97286992 0.98819622
18 0.99742995 1.00051421 0.99742995 0.98819622 0.97286992 0.95154556
19 0.9243546 0.89146469 0.8530786 0.809433 0.76079697 0.70747038
20 0.649782 0.5880875 0.52276724 0.45422394 0.38288021 0.30917589
21 0.2335654 0.1565149 0.07849944]
22 n=40, Gauss Seidel: [0.07849944 0.15651491 0.23356541 0.3091759 0.38288021
0.45422395
23 0.52276724 0.5880875 0.64978201 0.70747039 0.76079698 0.809433
24 0.85307861 0.89146469 0.92435461 0.95154557 0.97286993 0.98819623
25 0.99742997 1.00051422 0.99742997 0.98819623 0.97286993 0.95154557
26 0.92435461 0.89146469 0.85307861 0.809433 0.76079698 0.70747039
27 0.649782 0.5880875 0.52276724 0.45422395 0.38288021 0.3091759
28 0.23356541 0.15651491 0.07849944], Error: 0.0022996463089598887
29
30 n=80, Jacobi: [0.03926486 0.07846918 0.11755251 0.15645457 0.1951154 0.23347537
31 0.27147534 0.30905672 0.34616155 0.38273262 0.41871355 0.45404886
32 0.48868405 0.52256573 0.55564165 0.58786081 0.61917353 0.64953153
33 0.678888 0.70719768 0.7344169 0.76050371 0.78541788 0.80912099
34 0.83157649 0.85274977 0.87260816 0.89112106 0.90825991 0.92399829
35 0.93831193 0.95117877 0.96257895 0.97249491 0.98091135 0.9878153
36 0.99319611 0.99704548 0.99935748 1.00012854 0.99935748 0.99704548
37 0.99319611 0.9878153 0.98091135 0.97249491 0.96257895 0.95117877
38 0.93831193 0.92399829 0.90825991 0.89112106 0.87260816 0.85274977
39 0.83157649 0.80912099 0.78541788 0.76050371 0.7344169 0.70719768
40 0.678888 0.64953153 0.61917353 0.58786081 0.55564165 0.52256573
41 0.48868405 0.45404886 0.41871355 0.38273262 0.34616155 0.30905672
42 0.27147534 0.23347537 0.1951154 0.15645457 0.11755251 0.07846918
43 0.03926486]
44 n=80, Gauss Seidel: [0.03926486 0.07846918 0.11755251 0.15645458 0.19511541
0.23347538
45 0.27147535 0.30905673 0.34616156 0.38273264 0.41871357 0.45404888
46 0.48868407 0.52256575 0.55564167 0.58786083 0.61917356 0.64953156
47 0.67888803 0.70719771 0.73441693 0.76050374 0.78541791 0.80912102
48 0.83157653 0.8527498 0.8726082 0.89112109 0.90825995 0.92399833
49 0.93831197 0.95117881 0.96257899 0.97249495 0.9809114 0.98781534
50 0.99319615 0.99704552 0.99935752 1.00012859 0.99935752 0.99704552
51 0.99319615 0.98781534 0.98091139 0.97249495 0.96257899 0.95117881
52 0.93831197 0.92399833 0.90825995 0.89112109 0.8726082 0.8527498
53 0.83157653 0.80912102 0.78541791 0.76050374 0.73441693 0.7071977
54 0.67888803 0.64953156 0.61917355 0.58786083 0.55564167 0.52256575
55 0.48868407 0.45404888 0.41871357 0.38273264 0.34616156 0.30905673
56 0.27147535 0.23347538 0.19511541 0.15645458 0.11755251 0.07846918
57 0.03926486], Error: 0.0008132435845628745
58
59 n=160, Jacobi: [0.01963433 0.03926108 0.0588727 0.07846162 0.0980203 0.11754118
60 0.13701676 0.15643951 0.17580194 0.19509661 0.21431606 0.23345289
61 0.25249971 0.2714492 0.29029403 0.30902695 0.32764074 0.34612821

```

62	0.36448224	0.38269576	0.40076175	0.41867323	0.4364233	0.45400513
63	0.47141193	0.48863698	0.50567367	0.5225154	0.53915569	0.55558813
64	0.57180638	0.58780419	0.60357539	0.6191139	0.63441372	0.64946897
65	0.66427384	0.67882262	0.69310969	0.70712956	0.72087682	0.73434617
66	0.74753241	0.76043047	0.77303536	0.78534223	0.79734634	0.80904306
67	0.82042788	0.8314964	0.84224437	0.85266764	0.86276218	0.87252412
68	0.88194968	0.89103523	0.89977727	0.90817243	0.91621748	0.9239093
69	0.93124494	0.93822156	0.94483649	0.95108716	0.95697117	0.96248625
70	0.96763027	0.97240125	0.97679735	0.98081688	0.98445829	0.98772016
71	0.99060126	0.99310045	0.99521679	0.99694945	0.99829777	0.99926123
72	0.99983945	1.00003222	0.99983945	0.99926123	0.99829777	0.99694945
73	0.99521679	0.99310045	0.99060126	0.98772016	0.98445829	0.98081688
74	0.97679735	0.97240125	0.96763027	0.96248625	0.95697117	0.95108716
75	0.94483649	0.93822156	0.93124494	0.9239093	0.91621748	0.90817243
76	0.89977727	0.89103523	0.88194968	0.87252412	0.86276218	0.85266764
77	0.84224437	0.8314964	0.82042788	0.80904306	0.79734634	0.78534223
78	0.77303536	0.76043047	0.74753241	0.73434617	0.72087682	0.70712956
79	0.69310969	0.67882262	0.66427384	0.64946897	0.63441372	0.6191139
80	0.60357539	0.58780419	0.57180638	0.55558813	0.53915569	0.5225154
81	0.50567367	0.48863698	0.47141193	0.45400513	0.4364233	0.41867323
82	0.40076175	0.38269576	0.36448224	0.34612821	0.32764074	0.30902695
83	0.29029403	0.2714492	0.25249971	0.23345289	0.21431606	0.19509661
84	0.17580194	0.15643951	0.13701676	0.11754118	0.0980203	0.07846162
85	0.0588727	0.03926108	0.01963433]			
86	n=160, Gauss Seidel: [0.01963433 0.03926109 0.05887271 0.07846164 0.09802032 0.1175412					
87	0.13701678	0.15643953	0.17580197	0.19509664	0.21431609	0.23345293
88	0.25249976	0.27144924	0.29029408	0.309027	0.32764079	0.34612827
89	0.36448231	0.38269583	0.40076181	0.4186733	0.43642338	0.4540052
90	0.47141201	0.48863707	0.50567375	0.52251549	0.53915579	0.55558823
91	0.57180648	0.58780429	0.60357549	0.619114	0.63441383	0.64946908
92	0.66427395	0.67882273	0.69310981	0.70712968	0.72087694	0.73434629
93	0.74753254	0.76043059	0.77303549	0.78534237	0.79734648	0.8090432
94	0.82042802	0.83149654	0.84224451	0.85266778	0.86276233	0.87252427
95	0.88194983	0.89103538	0.89977743	0.90817259	0.91621763	0.92390946
96	0.9312451	0.93822172	0.94483665	0.95108732	0.95697133	0.96248641
97	0.96763043	0.97240141	0.97679752	0.98081705	0.98445845	0.98772033
98	0.99060142	0.99310062	0.99521696	0.99694962	0.99829794	0.9992614
99	0.99983962	1.00003239	0.99983962	0.9992614	0.99829794	0.99694962
100	0.99521696	0.99310062	0.99060142	0.98772033	0.98445845	0.98081705
101	0.97679752	0.97240141	0.96763043	0.96248641	0.95697133	0.95108732
102	0.94483665	0.93822172	0.93124509	0.92390945	0.91621763	0.90817259
103	0.89977742	0.89103538	0.88194983	0.87252426	0.86276233	0.85266778
104	0.84224451	0.83149654	0.82042801	0.8090432	0.79734648	0.78534236
105	0.77303549	0.76043059	0.74753254	0.73434629	0.72087694	0.70712968
106	0.69310981	0.67882273	0.66427395	0.64946908	0.63441383	0.619114
107	0.60357549	0.58780429	0.57180648	0.55558823	0.53915578	0.52251549
108	0.50567375	0.48863707	0.471412	0.4540052	0.43642337	0.4186733
109	0.40076181	0.38269583	0.3644823	0.34612827	0.32764079	0.309027
110	0.29029408	0.27144924	0.25249975	0.23345292	0.21431609	0.19509664
111	0.17580197	0.15643953	0.13701678	0.1175412	0.09802031	0.07846164
112	0.05887271	0.03926109	0.01963433], Error: 0.00028968235456973407			


```

113
114 Question 2 error table: [(0.018482034928874364, 0.018482036208807052),
(0.006510206868033681, 0.006510214971563716), (0.0022995999225368553,
0.0022996463089598887), (0.0008129789948298825, 0.0008132435845628745),
(0.0002881833488868389, 0.00028968235456973407)]
115 Question 2 iter times table: [(408, 195), (1539, 729), (5727, 2697), (21127,
9891), (77331, 35970)]

```

3. Newton迭代Python代码

```

1  from math import pi, sin, sqrt, nan
2  import numpy as np
3
4  def newton_iteration(n:int , epsilon: float, Jacobi_function, F_function):
5      X = np.zeros(n-1)
6      iter_times = 0
7      while True:
8          iter_times += 1
9          Jacobi_matrix = Jacobi_function(X)
10         delta = np.linalg.inv(Jacobi_matrix).dot(F_function(X).reshape(n-1, 1))
11         X = X - delta.reshape(n-1)
12         if np.max(np.abs(delta)) < epsilon:
13             break
14     return X, iter_times
15
16 def run_question_6(n: int, print_result: bool = False):
17     h = 1/n
18     def f(x):
19         return pi * pi * sin(pi*x) + sin(pi*x) ** 3
20
21     def u_e(x):
22         return sin(pi*x)
23
24     def Jacobi(input:np.array):
25         res = np.zeros((n-1, n-1))
26         for i in range(n-1):
27             res[i][i] = 2 + 3 * input[i] * input[i] * h * h
28             if i > 0:
29                 res[i][i-1] = -1
30             if i < n-2:
31                 res[i][i+1] = -1
32         return res
33
34     def F(X:np.array):
35         res = np.zeros(n-1)
36         res[0] = 2 * X[0] - X[1] + X[0] ** 3 * h * h - f(h) * (h ** 2)
37         res[n-2] = 2 * X[n-2] - X[n-3] + X[n-2] ** 3 * h * h - f(1 - h) * (h ** 2)
38         for i in range(1, n-2):

```

```

39         res[i] = 2 * X[i] - X[i-1] - X[i+1] + X[i] ** 3 * h * h - f((i+1) * h)
40     * (h ** 2)
41     return res
42
43     X, iter_times = newton_iteration(n, 1e-8, Jacobi, F)
44     if print_result:
45         print(f"n={n}, Newton: {X}")
46         print(f"n={n}, Newton iter times: {iter_times}")
47         error = sqrt(sum([(X[i] - u_e((i+1)/n))**2 for i in range(n-1)]))
48         return error, iter_times
49
50 question_6_n = [10, 20, 40, 80, 160]
51 question_6_errors = []
52 question_6_iters = []
53 for n in question_6_n:
54     error, iters = run_question_6(n, True)
55     question_6_errors.append(error)
56     question_6_iters.append(iters)
57 print("Question 6 error table:", question_6_errors)
58 print("Question 6 iter times table:", question_6_iters)

```

4. Newton迭代输出解

```

1  n=10, Newton: [0.31114144 0.59179026 0.81446879 0.95740831 1.00665583 0.95740831
2    0.81446879 0.59179026 0.31114144]
3
4  n=20, Newton: [0.15670316 0.30954651 0.45476555 0.58878402 0.70830206 0.81037743
5    0.8924978 0.95264236 0.98933131 1.00166208 0.98933131 0.95264236
6    0.8924978 0.81037743 0.70830206 0.58878402 0.45476555 0.30954651
7    0.15670316]
8
9  n=40, Newton: [0.07849278 0.15650159 0.23354543 0.30914928 0.38284697 0.45418413
10   0.52272092 0.58803479 0.64972307 0.70740544 0.7607263 0.80935695
11   0.85299759 0.8913792 0.92426519 0.95145284 0.97277458 0.98809896
12   0.99733154 1.0004154 0.99733154 0.98809896 0.97277458 0.95145284
13   0.92426519 0.8913792 0.85299759 0.80935695 0.7607263 0.70740544
14   0.64972307 0.58803479 0.52272092 0.45418413 0.38284697 0.30914928
15   0.23354543 0.15650159 0.07849278]
16
17 n=80, Newton: [0.03926403 0.07846752 0.11755001 0.15645124 0.19511124 0.23347038
18   0.27146951 0.30905006 0.34615406 0.38272431 0.41870442 0.4540389
19   0.48867328 0.52255414 0.55562926 0.58784763 0.61915956 0.64951679
20   0.6788725 0.70718144 0.73439994 0.76048604 0.78539952 0.80910197
21   0.83155684 0.85272951 0.87258733 0.89109968 0.90823803 0.92397594
22   0.93828915 0.95115559 0.96255542 0.97247108 0.98088726 0.98779099
23   0.99317163 0.99702088 0.9993328 1.00010384 0.9993328 0.99702088
24   0.99317163 0.98779099 0.98088726 0.97247108 0.96255542 0.95115559
25   0.93828915 0.92397594 0.90823803 0.89109968 0.87258733 0.85272951]

```

```

26 0.83155684 0.80910197 0.78539952 0.76048604 0.73439994 0.70718144
27 0.6788725 0.64951679 0.61915956 0.58784763 0.55562926 0.52255414
28 0.48867328 0.4540389 0.41870442 0.38272431 0.34615406 0.30905006
29 0.27146951 0.23347038 0.19511124 0.15645124 0.11755001 0.07846752
30 0.03926403]
31
32 n=160, Newton: [0.01963422 0.03926087 0.05887238 0.0784612 0.09801977 0.11754055
33 0.13701602 0.15643866 0.17580099 0.19509555 0.21431489 0.23345162
34 0.25249834 0.27144772 0.29029244 0.30902526 0.32763894 0.34612631
35 0.36448024 0.38269365 0.40075953 0.41867091 0.43642088 0.4540026
36 0.47140929 0.48863425 0.50567083 0.52251246 0.53915265 0.55558499
37 0.57180314 0.58780085 0.60357194 0.61911035 0.63441008 0.64946523
38 0.66427 0.67881868 0.69310567 0.70712544 0.72087261 0.73434187
39 0.74752802 0.76042598 0.77303079 0.78533758 0.7973416 0.80903824
40 0.82042297 0.83149142 0.84223931 0.8526625 0.86275697 0.87251884
41 0.88194433 0.89102981 0.89977179 0.90816689 0.91621187 0.92390363
42 0.93123922 0.93821579 0.94483066 0.95108128 0.95696525 0.96248028
43 0.96762426 0.97239521 0.97679128 0.98081077 0.98445215 0.987714
44 0.99059507 0.99309425 0.99521057 0.99694322 0.99829153 0.99925498
45 0.9998332 1.00002596 0.9998332 0.99925498 0.99829153 0.99694322
46 0.99521057 0.99309425 0.99059507 0.987714 0.98445215 0.98081077
47 0.97679128 0.97239521 0.96762426 0.96248028 0.95696525 0.95108128
48 0.94483066 0.93821579 0.93123922 0.92390363 0.91621187 0.90816689
49 0.89977179 0.89102981 0.88194433 0.87251884 0.86275697 0.8526625
50 0.84223931 0.83149142 0.82042297 0.80903824 0.7973416 0.78533758
51 0.77303079 0.76042598 0.74752802 0.73434187 0.72087261 0.70712544
52 0.69310567 0.67881868 0.66427 0.64946523 0.63441008 0.61911035
53 0.60357194 0.58780085 0.57180314 0.55558499 0.53915265 0.52251246
54 0.50567083 0.48863425 0.47140929 0.4540026 0.43642088 0.41867091
55 0.40075953 0.38269365 0.36448024 0.34612631 0.32763894 0.30902526
56 0.29029244 0.27144772 0.25249834 0.23345162 0.21431489 0.19509555
57 0.17580099 0.15643866 0.13701602 0.11754055 0.09801977 0.0784612
58 0.05887238 0.03926087 0.01963422]
59
60 Question 6 error table: [0.015018032824901736, 0.005300889459111111,
0.0018733686037447255, 0.0006622669330582057, 0.00023414062963731416]
61 Question 6 iter times table: [5, 5, 5, 5, 5]

```