數值 HW7

第一題:

•[3]: import numpy as np

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A = np.array([[ 4, -1, 0, -1, 0, 0],
                    [-1, 4, -1, 0, -1, 0],
                    [ 0, -1, 4, 0, 1, -1],
                    [-1, 0, 0, 4, -1, -1],
                     [0, -1, 0, -1, 4, -1],
                     [ 0, 0, -1, 0, -1, 4]], dtype=float)
       b = np.array([0, -1, 9, 4, 8, 6], dtype=float)
       def jacobi(A, b, x0, tol=1e-10, max_iter=1000):
          D = np.diag(np.diag(A))
          R = A - D
          x = x0.copy()
          for i in range(max_iter):
              x_new = np.linalg.inv(D) @ (b - R @ x)
              if np.linalg.norm(x_new - x, np.inf) < tol:</pre>
                  return x new, i + 1
              x = x new
           return x, max_iter
       # 初始猜測
       x0 = np.zeros_like(b)
       # 執行
       x_jacobi, it_jacobi = jacobi(A, b, x0)
       # 顯示結果
       print(f"Jacobi result in {it_jacobi} iterations:\n{x_jacobi}")
       Jacobi result in 47 iterations:
       [1.17478856 1.64317358 2.44824809 3.05598067 3.94965767 3.09947644]
第二題:
[5]: def gauss_seidel(A, b, x0, tol=1e-10, max_iter=1000):
          x = x0.copy()
          n = len(b)
          for it in range(max_iter):
              x_new = x.copy()
              for i in range(n):
                  s1 = np.dot(A[i, :i], x_new[:i])
                  s2 = np.dot(A[i, i+1:], x[i+1:])
                  x_{new[i]} = (b[i] - s1 - s2) / A[i, i]
              if np.linalg.norm(x_new - x, np.inf) < tol:</pre>
                  return x_new, it + 1
              x = x_new
          return x, max_iter
      # 初始猜測
      x0 = np.zeros_like(b)
      x_gs, it_gs = gauss_seidel(A, b, x0)
      print(f"Gauss-Seidel result in {it_gs} iterations:\n{x_gs}")
      Gauss-Seidel result in 19 iterations:
      [1.17478856 1.64317358 2.44824809 3.05598067 3.94965767 3.09947644]
```

第三題:

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•[7]: import numpy as np
      A = np.array([[ 4, -1, 0, -1, 0, 0],
                    [-1, 4, -1, 0, -1, 0],
                    [ 0, -1, 4, 0, 1, -1],
                    [-1, 0, 0, 4, -1, -1],
                    [0, -1, 0, -1, 4, -1],
                    [ 0, 0, -1, 0, -1, 4]], dtype=float)
      b = np.array([0, -1, 9, 4, 8, 6], dtype=float)
      def sor(A, b, x0, w=1.2, tol=1e-10, max iter=1000):
          x = x0.copy()
          n = len(b)
          for it in range(max_iter):
              x_new = x.copy()
              for i in range(n):
                  s1 = np.dot(A[i, :i], x_new[:i])
                  s2 = np.dot(A[i, i+1:], x[i+1:])
                  x_{new[i]} = (1 - w) * x[i] + (w / A[i, i]) * (b[i] - s1 - s2)
              if np.linalg.norm(x_new - x, np.inf) < tol:</pre>
                  return x_new, it + 1
              x = x new
          return x, max_iter
      # 初始猜測
      x0 = np.zeros_like(b)
      x_sor, it_sor = sor(A, b, x0, w=1.25)
      print(f"SOR result (\omega=1.25) in {it_sor} iterations:\n{x_sor}")
      SOR result (\omega=1.25) in 25 iterations:
      [1.17478856 1.64317358 2.44824809 3.05598067 3.94965767 3.09947644]
```

第四題:

```
•[9]: import numpy as np
      A = np.array([[ 4, -1, 0, -1, 0, 0],
                    [-1, 4, -1, 0, -1, 0],
                    [ 0, -1, 4, 0, 1, -1],
                    [-1, 0, 0, 4, -1, -1],
                    [ 0, -1, 0, -1, 4, -1],
                    [ 0, 0, -1, 0, -1, 4]], dtype=float)
      b = np.array([0, -1, 9, 4, 8, 6], dtype=float)
      def conjugate_gradient(A, b, x0, tol=1e-10, max_iter=1000):
          x = x0.copy()
          r = b - A @ x
          p = r.copy()
          rs_old = np.dot(r, r)
          for i in range(max_iter):
              Ap = A @ p
              alpha = rs_old / np.dot(p, Ap)
              x += alpha * p
              r -= alpha * Ap
              rs_new = np.dot(r, r)
              if np.sqrt(rs_new) < tol:</pre>
                  return x, i + 1
              p = r + (rs_new / rs_old) * p
              rs_old = rs_new
          return x, max_iter
      # 初始猜測
      x0 = np.zeros_like(b)
      x_cg, it_cg = conjugate_gradient(A, b, x0)
      print(f"Conjugate Gradient result in {it_cg} iterations:\n{x_cg}")
      Conjugate Gradient result in 1000 iterations:
      [1.17656665 1.64269366 2.44433267 3.06002082 3.95260785 3.09922059]
```