

homework5

April 6, 2025

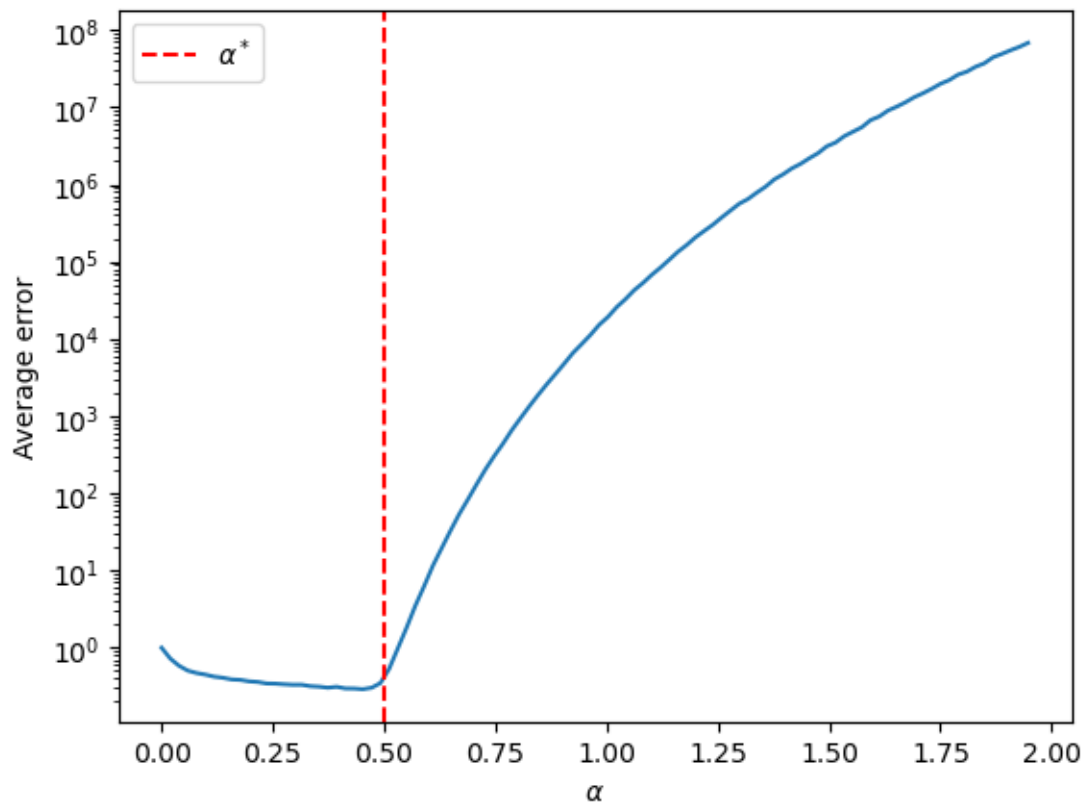
```
[33]: import numpy as np
import matplotlib.pyplot as plt
```

1 Problem 19

```
[34]: def make_tridiagonal(n, a, b, c):
    diag = np.diag(b * np.ones(n))
    sub_diag = np.diag(a * np.ones(n-1), -1)
    sup_diag = np.diag(c * np.ones(n-1), 1)
    return sub_diag + diag + sup_diag
```

```
[35]: def richardson_test(n, m, k, alpha):
    A = make_tridiagonal(n, -1, 2, -1)
    error = 0
    for _ in range(m):
        u_0 = np.random.rand(n)
        u_sol = np.random.rand(n)
        u = u_0.copy()
        b = A @ u_sol
        for _ in range(k):
            u = u + alpha * (b - A @ u)
        error += np.linalg.norm(u - u_sol) / np.linalg.norm(u_0 - u_sol)
    return error / m
```

```
[49]: n = 100
endpoint = 1 / np.cos(np.pi / n+1)
optimal_alpha = 1/2
alphas = np.linspace(0, endpoint, 100)
errors_richardson = [richardson_test(n, 100, 10, alpha) for alpha in alphas]
plt.semilogy(alphas, errors_richardson)
plt.axvline(optimal_alpha, color='red', linestyle='--', label=r'$\alpha^{*}$')
plt.xlabel(r'$\alpha$')
plt.ylabel('Average error')
plt.legend()
plt.show()
```



2 Problem 20

Algorithm which avoids multiplying with A twice

```
[37]: def make_spd_matrix(n, condition_number=2):
    Q, _ = np.linalg.qr(np.random.randn(n, n))

    eigvals = np.linspace(1, condition_number, n)
    D = np.diag(eigvals)

    A = Q @ D @ Q.T
    return A

def steepest_gradient_descent(A, b, x0, tol=1e-15, max_iter=10000,
    verbose=True):
    x = [x0]
    r = b - A @ x[-1]

    for iter_count in range(max_iter):
```

```

    Ar = A @ r
    alpha = (r.T @ r) / ( r.T @ Ar )
    x.append(x[-1] + alpha * r)
    if np.linalg.norm(r) < tol:
        if verbose:
            print(f"Converged in {iter_count} iterations.")
        return np.array(x)
    r = r - alpha * Ar
if verbose:
    print(f"Did not converge in {max_iter} iterations.")
return np.array(x)

```

```

[38]: n = 1000

A = make_spd_matrix(n, condition_number=3)

sol = np.random.rand(n)
b = A @ sol
x0 = np.random.rand(n)

x = steepest_gradient_descent(A, b, x0)
errors = x - sol

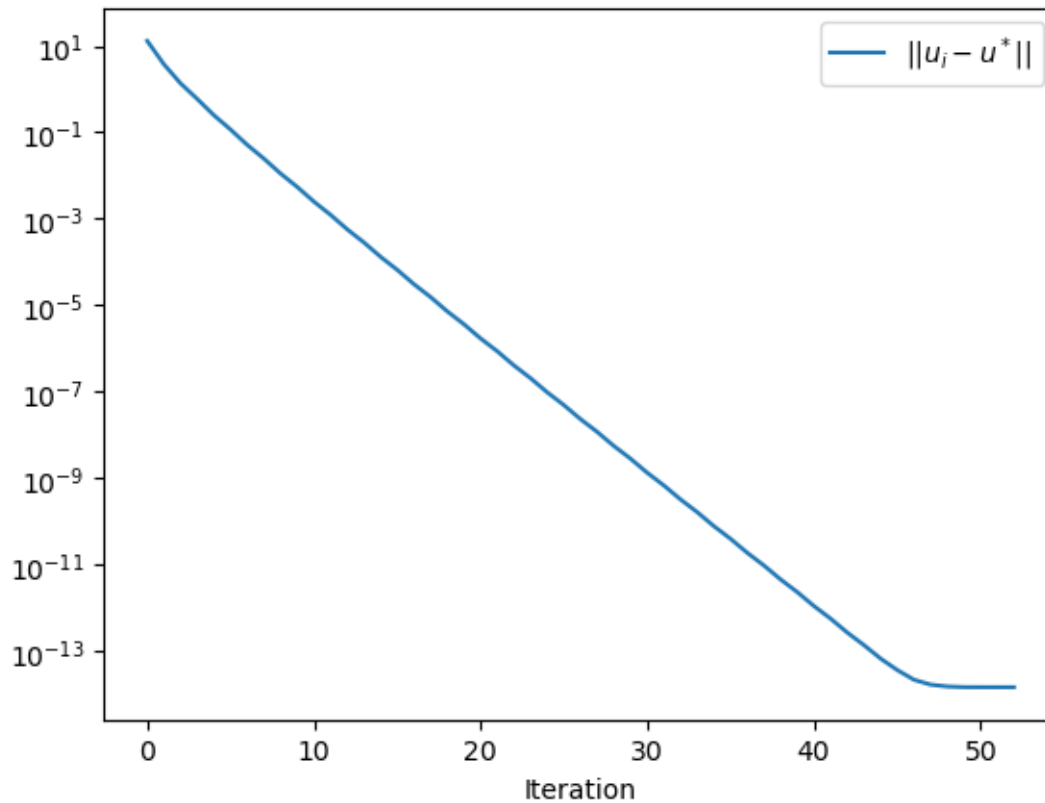
```

Converged in 51 iterations.

```

[39]: plt.semilogy(np.linalg.norm(errors, axis=1), label=r'$||u_i - u^*||$')
plt.xlabel('Iteration')
plt.legend()
plt.show()

```



2.1 Starred Exercise of Problem 20

```
[40]: n = 100
A = make_tridiagonal(n, -1, 2, -1)
m = 100
k = 10

error = 0

for _ in range(m):
    u_0 = np.random.rand(n)
    u_sol = np.random.rand(n)
    b = A @ u_sol

    u = steepest_gradient_descent(A, b, u_0, max_iter=k, verbose=False)
    e_k = np.linalg.norm(u[-1] - u_sol)
    e_0 = np.linalg.norm(u[0] - u_sol)

    error += e_k / e_0

error /= m
print(f"Average error: {error}")
```

```
print(f"Minimal average error with Richardson: {np.min(errors_richardson)}")
```

Average error: 0.3046466749355231

Minimal average error with Richardson: 0.29760259280854245

3 Problem 22

```
[41]: n = 1000

A = make_spd_matrix(n, condition_number=3)
L = np.linalg.cholesky(A)
sol = np.random.rand(n)
b = A @ sol
x0 = np.random.rand(n)

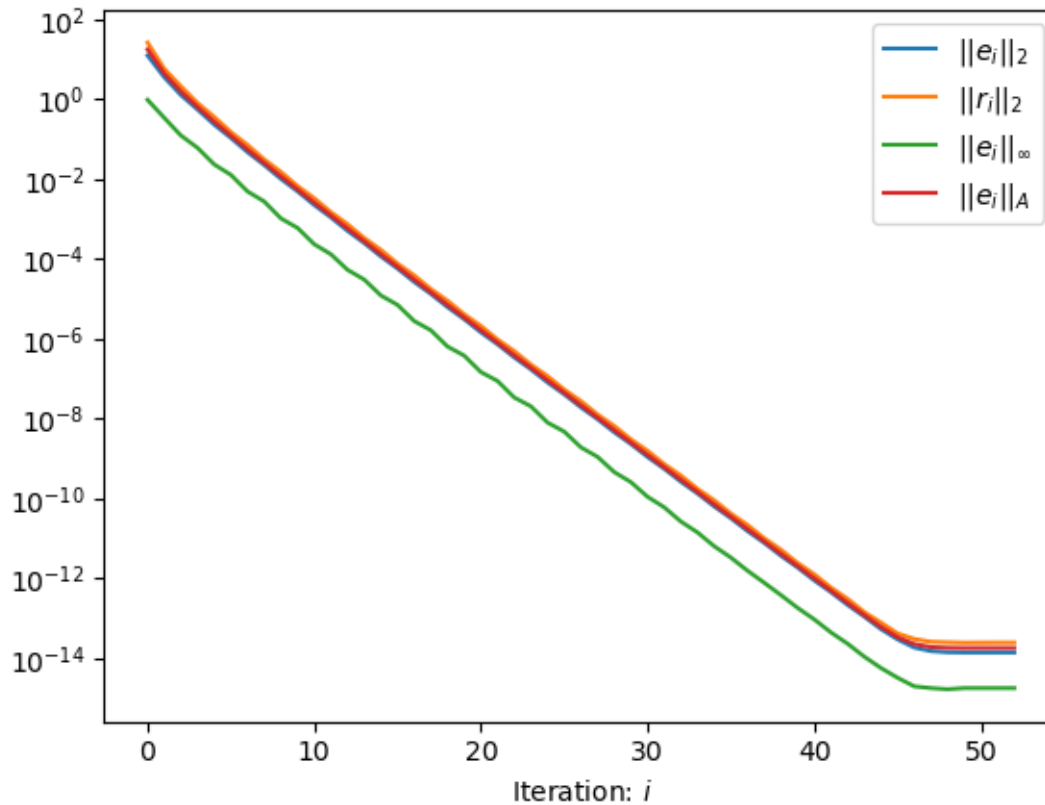
x = steepest_gradient_descent(A, b, x0)

errors = x - sol
residuals = b - [A @ x[i] for i in range(len(x))]

errors_l2 = np.linalg.norm(errors, axis=1, ord=2)
errors_inf = np.linalg.norm(errors, axis=1, ord=np.inf)
# Compute  $\sqrt{x^T A x}$  by computing 2-norm of  $L.T @ x$  with  $LL^T = A$ 
errors_A_norm = np.linalg.norm([L.T @ errors[i] for i in range(len(x))],
                                axis=1, ord=2)
residuals_l2 = np.linalg.norm(residuals, axis=1)
```

Converged in 51 iterations.

```
[42]: plt.semilogy(errors_l2, label=r'$||e_i||_2$')
plt.semilogy(residuals_l2, label=r'$||r_i||_2$')
plt.semilogy(errors_inf, label=r'$||e_i||_{\infty}$')
plt.semilogy(errors_A_norm, label=r'$||e_i||_A$')
plt.xlabel('Iteration: $i$')
plt.legend()
plt.show()
```



4 Problem 23

```
[43]: def make_preconditioner(n, k):
        Q = make_tridiagonal(n, 1, 0, 1)
        return np.sum([1/2**(i+1) * np.linalg.matrix_power(Q, i) for i in
        ↪ range(k+1)], axis=0)

[44]: n = 1000
A = make_tridiagonal(n, -1, 2, -1)
u0 = np.random.rand(n)
sol = np.random.rand(n)
b = A @ sol

A_tilde = lambda k: make_preconditioner(n, k) @ A
b_tilde = lambda k: make_preconditioner(n, k) @ b

[45]: ratios = []
for k in range(0, 11):
    x = steepest_gradient_descent(A_tilde(k), b_tilde(k), u0, max_iter=1,
    ↪ verbose=False)
```

```
e0 = np.linalg.norm(x[0] - sol)
e1 = np.linalg.norm(x[1] - sol)
ratio = e1 / e0
ratios.append(ratio)
```

```
[46]: plt.plot(range(0, 11), ratios)
plt.xlabel('k')
plt.ylabel(r'$\frac{||e_1||}{||e_0||}$')
```

```
[46]: Text(0, 0.5, '$\frac{||e_1||}{||e_0||}$')
```

