

# Home assignment 2

Numerical Optimization and its Applications - Spring 2019

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## 1 The efficiency of different iterative methods for solving a linear system

(a) Following are the implementation for the four methods:

**Jacobi:**

```
from numpy import diag, matmul, array
from numpy.linalg import inv, norm

def weighted_jacobi(A, b, x_0, maxIter, epsilon, w):
    D = diag(diag(A))
    x = x_0
    res = [norm(matmul(A, x) - b)]
    for i in range(maxIter):
        x = x + w * matmul(inv(D), b - matmul(A, x))
        res.append(norm(matmul(A, x) - b))
        if res[-1] / norm(b) < epsilon:
            break
    return x, array(res)
```

**Gauss-Seidel:**

```

from numpy import matmul, tril, array
from numpy.linalg import inv, norm

def weighted_gauss_seidel(A, b, x_0, maxIter, epsilon, w):
    L_D = tril(A, k=0)
    x = x_0
    res = [norm(matmul(A, x) - b)]
    for i in range(maxIter):
        x = x + w * matmul(inv(L_D), b - matmul(A, x))
        res.append(norm(matmul(A, x) - b))
        if res[-1] / norm(b) < epsilon:
            break
    return x, array(res)

```

- (b) Following are the system and parameters definition, methods calls, residual vector norm and convergence factor plotting:

```

import numpy as np
from scipy.sparse import spdiags
import matplotlib.pyplot as plt
from py_files.part_1_gauss_seidel import weighted_gauss_seidel
from py_files.part_1_jacobi import weighted_jacobi

n = 100
# TODO: not sure if .toarray() is the right approach
A = spdiags(np.array([-np.ones(n), 2.1 * np.ones(n), -np.ones(n)]),
            np.array([-1, 0, 1]), n, n).toarray()
x_0 = np.zeros(n)
b = np.random.rand(n)
maxIter = 100
epsilon = 1e-6

res = dict()
x, res['weighted_jacobi_1'] = weighted_jacobi(A, b, x_0, maxIter, epsilon, 1)
#print('weighted_jacobi_1 result:', x)
x, res['weighted_jacobi_0_75'] = weighted_jacobi(A, b, x_0, maxIter, epsilon, 0.75)
#print('weighted_jacobi_0_75 result:', x)
x, res['weighted_gauss_seidel_1'] = weighted_gauss_seidel(A, b, x_0, maxIter, epsilon, 1)
#print('weighted_gauss_seidel_1 result:', x)

convergence_factor = dict()
for alg_res in res:

```

```

convergence_factor[alg_res] = res[alg_res][1:] / res[alg_res][::-1]

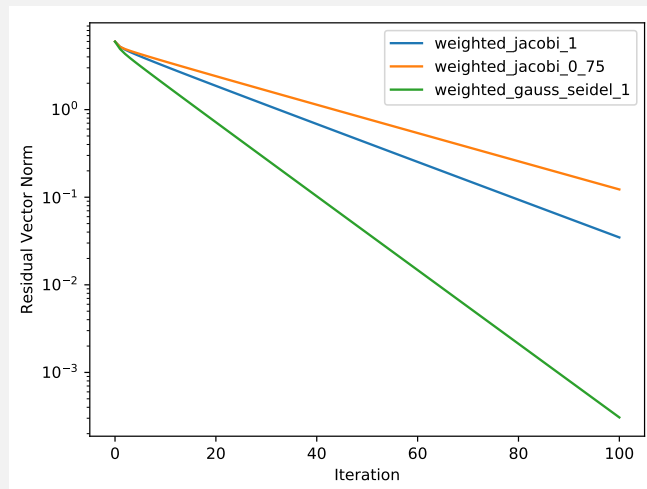
plt.figure()
for alg_res in res:
    plt.semilogy(res[alg_res], label=alg_res)
plt.legend()
plt.xlabel("Iteration")
plt.ylabel("Residual Vector Norm")

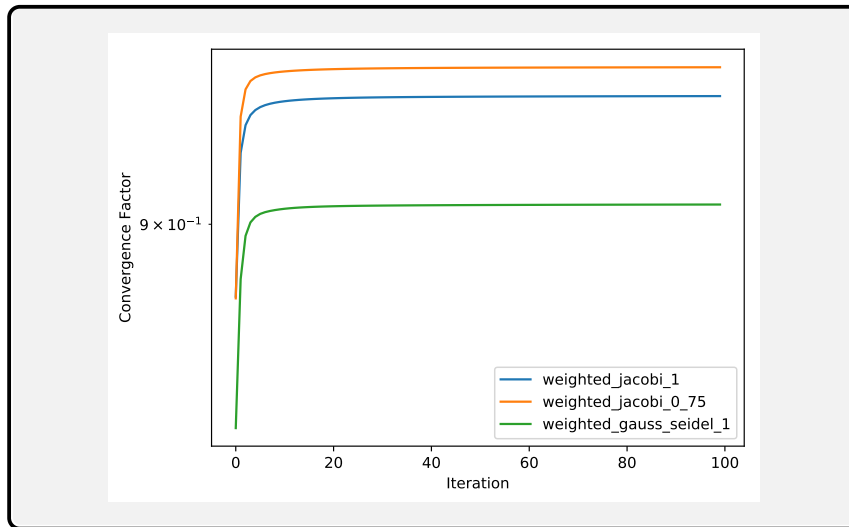
# Save the plot as .pdf and include it in the .tex document
plt.savefig("myplot1.pdf", bbox_inches="tight")
print(r"\saveandshowplot{myplot1.pdf}")

plt.figure()
for alg_con in convergence_factor:
    plt.semilogy(convergence_factor[alg_con], label=alg_con)
plt.legend()
plt.xlabel("Iteration")
plt.ylabel("Convergence Factor")

# Save the plot as .pdf and include it in the .tex document
plt.savefig("myplot2.pdf", bbox_inches="tight")
print(r"\saveandshowplot{myplot2.pdf}")

```





## 2 Convergence properties

- (a) t
- (b) t
- (c) t

## 3 GMRES(1) method

- (a) t
- (b) t
- (c) t
- (d) t
- (e) t

## 4 Convexity

- (a) t
- (b) t
- (c) t

## 5 Non Linear Optimization

(a) t

(b) t

(c) t