

# Project 1

## Scientific Computing III

R.G.A. Deckers

---

### *Contents*

<b>1</b>	<b>Creating the matrices <math>K</math>, <math>M</math>, and <math>A</math></b>	<b>1</b>
<b>2</b>	<b>Power method</b>	<b>1</b>
a	Implementation . . . . .	1
b	Convergence . . . . .	1
c	Results . . . . .	2
c.1	Eigenvector . . . . .	2
c.2	Iterations vs. Matrix Size . . . . .	2
<b>3</b>	<b>Inverse iteration</b>	<b>2</b>
a	Convergence . . . . .	2
b	Implementation . . . . .	2
c	results . . . . .	2
c.1	iterations vs. matrix size . . . . .	2
c.2	error vs. matrix size . . . . .	2
<b>A</b>	<b>Source Code</b>	<b>3</b>

### *Creating the matrices $K$ , $M$ , and $A$*

First, before we can do anything else, we of course have to create our matrices. the function we use for doing so is given in listing 4.

### *Power method*

#### *Implementation*

The implementation of the power-method is given in listing 5.

#### *Convergence*

Let  $1 = |\lambda_1| \geq \dots \geq |\lambda_n|$  be the eigenvalues of a matrix  $A$  and  $\vec{e}_1, \dots, \vec{e}_n$  the corresponding eigenvectors. Then  $A^k \vec{x} = \sum_{i=1}^n a_i \lambda_i^k \vec{e}_i$ . where  $a_i$  are the eigenspace coefficients of  $\vec{x}$ . All eigenvectors with eigenvalues less than one converges to zero for  $k \rightarrow \infty$ , therefore if all  $\lambda_i$  for  $i \neq 1$  are smaller than  $\lambda_1$  the rate of convergence in the first order will be  $|\lambda_1/\lambda_2|$ . If the dominant eigenvalue is not unique (down to a sign) than the eigenvalue will still convert, at a rate defined by the next eigenvalue less than  $\lambda_1$  in the absolute, but the eigenvector will not.

## ***Results***

### ***Eigenvector***

### ***Iterations vs. Matrix Size***

### ***Inverse iteration***

### ***Convergence***

Note that the inverse iteration algorithm is equivalent to the power method over  $A - \mu I$  instead of  $A$ , where  $\mu$  is the eigenvalue of interest. It is therefore trivial to deduce that the convergence rate is given by  $|\bar{\lambda}_1/\bar{\lambda}_2|$  where  $\bar{\lambda}_i$  are the original eigenvalues shifted by  $\mu$ .

### ***Implementation***

The implementation of the the inverse iteration algorithm is given in listing 3.

### ***results***

### ***iterations vs. matrix size***

### ***error vs. matrix size***

## Source Code

Listing 1: Matlab script used to generate the data for task 1.

```
1 figure(1)
2 n= 16;
3 iteration = 1;
4 steps = [];
5 N = [];
6 while n<=512
7     A = create_matrices(n);
8     real_eigenvalue = max(eig(full(A)));
9     eigenvalues = eig_power(A);
10    steps(iteration) = size(eigenvalues,2)
11    N(iteration) = n
12    iteration = iteration + 1
13    errors = abs(eigenvalues-real_eigenvalue);
14    loglog(errors(1:end));
15    hold on;
16    n = n*2;
17 end
18 figure(2)
19 plot(N,steps)
```

Listing 2: Matlab script used to generate the data for task 2.

```
1 figure(1)
2 n= 16;
3 iteration = 1;
4 steps = [];
5 N = [];
6 while n<=512
7     A = create_matrices(n);
8     real_max_eigenvalue = max(eig(full(A)));
9     real_min_eigenvalue = min(eig(full(A)));
10    eigenvalues_max = eig_power(A);
11    B = A-eigenvalues_max(end)*eye(n);
12    eigenvalues_min = eig_power(B)+eigenvalues_max(end);
13    steps(iteration) = size(eigenvalues_max,2)
14    N(iteration) = n;
15    iteration = iteration + 1;
16    errors_max = abs(eigenvalues_max-real_max_eigenvalue);
17    errors_min = abs(eigenvalues_min-real_min_eigenvalue);
18    loglog(errors_max(1:end));
19    hold on;
20    loglog(errors_min(1:end));
21    n = n*2;
22 end
```

```

23 figure(2)
24 plot(N,steps)

```

Listing 3: Matlab function for computing the eigenvector corresponding to a specific eigenvalue using inverse iteration.

```

1 function [ eigenvalue ] = inverse_iteration( A, mu )
2 %INVERSE_ITERATION Summary of this function goes here
3 % Detailed explanation goes here
4     B = A-sparse(mu*diag(ones(size(A,1),1),0));
5     eigenvalue = power_iteration(B)+mu;
6 end

```

Listing 4: Matlab function for creating our problem matrices.

```

1 function [ A, K, M ] = create_matrices( N )
2 %CREATE_MATRICES Summary of this function goes here
3 % Detailed explanation goes here
4 h = 1.0/N;
5
6 K = sparse(-diag(ones(N-1,1),-1)-diag(ones(N-1,1),+1)+diag(2*ones(N,1),0));
7 K(N,N) = 1;
8 K = K*N;
9
10 M = sparse(diag(ones(N-1,1),-1)+diag(ones(N-1,1),+1)+diag(4*ones(N,1),0));
11 M(N,N) = 2;
12 M = h/6*M;
13
14 A = M\K;
15
16 end

```

Listing 5: Matlab function for computing the dominant eigenvalue of a matrix using the power method.

```

1 function [ eigenvalues, eigenvector ] = eig_power( matrix, tolerance )
2 %POWER_ITERATION Computed the largest eigenvalue by power iteration
3 %returns an array of the computed eigenvalues
4
5 %first some basic checks on the input
6 assert(ismatrix(matrix));
7 assert(size(matrix,1) == size(matrix,2));
8
9 %if the tolerance isn't specified, set it to 1e-4
10 if ~exist('tolerance','var')
11     tolerance = 1e-4;
12 end
13
14 %preset the delta to always get 1 loop.
15 eigen_delta = 2*tolerance;

```

```

16
17 %first guess
18 z = ones(size(matrix,1),1);
19 y = z/norm(z);
20 z = matrix*y;
21 eigenvalues(1) = dot(y, z);
22
23 i = 2;
24 while eigen_delta > tolerance
25     y = z/norm(z);
26     z = matrix*y;
27     eigenvalues(i) = dot(y, z);
28     eigen_delta = abs(eigenvalues(i-1)-eigenvalues(i));
29     i = i +1;
30 end
31
32 eigenvector = z/norm(z);
33
34 end

```