

Ma423: Matrix Computation Lab 8

```
format long e;
```

Question 1

```
% Question 1
fprintf(' ---- Question 1 ---- \n');
```

```
---- Question 1 ----
```

```
A = [2 -12; 1 -5];
x = [1; 1];
k = 8;
[iter, lambda] = PowerMethod(A, x, k);
disp('Iterates matrix:');
```

Iterates matrix:

```
disp(iter);
```

1.000000000000000e+00	1.000000000000000e+00	1.000000000000000e+00	1.000000000000000e+00	1.000000000000000e+00
4.000000000000000e-01	3.571428571428570e-01	3.437499999999999e-01	3.382352941176471e-01	3.357142857142857e-01

```
disp('Dominant Eigenvalue:');
```

Dominant Eigenvalue:

```
disp(lambda);
```

-2.006993006993008e+00

Observation: It can be seen that the iterates seem to converge to the dominant eigenvector $[3 \ 1]'$ and λ is close to the dominant eigenvalue -2.

Question 2

```
fprintf(' ---- Question 2 ---- ');
```

```
---- Question 2 ----
```

```
A1 = [1 1 1; -1 9 2; 0 -1 2];
A2 = [1 1 1; -1 9 2; -4 -1 2];
A3 = [1 1 1; -1 3 2; -4 -1 2];
x = [1 1 1]';
k = 20;

matrices = {A1, A2, A3};
matrix_names = {'A1', 'A2', 'A3'};

for idx = 1:length(matrices)
    A = matrices{idx};
```


Observations:

- **Part (a):** All 3 eigen values are different, thus the ratio of norms are also very close to $|\lambda_2| / |\lambda_1|$ which means the iterations converge linearly.
- **Part (b):** Here $|\lambda_2| = |\lambda_3|$, hence there is no definite convergence to the ratios, and the iterations do not converge linearly.
- **Part (c):** Here $|\lambda_1| = |\lambda_2|$, hence there is no definite convergence to the ratios, and the iterations do not converge linearly. → check $|\lambda_2| \approx |\lambda_3|$

Q37

Question 4

```
fprintf(' ---- Question 4 ---- ');
```

```
---- Question 4 ----
```

```
A = [4, 1, 0; 1, 3, 1; 0, 1, 2];  
x = rand(3, 1);  
s = 2.5;  
k = 10;  
[iter, lambda] = Shiftinv(A, x, s, k);  
fprintf('Approximate eigenvalue closest to s: %.6f\n', lambda);
```

```
Approximate eigenvalue closest to s: 3.000000
```

```
disp('Approximate eigenvector:');
```

```
Approximate eigenvector:
```

```
disp(iter(:, end));
```

```
-9.993845912336913e-01  
9.991653939304213e-01  
1.000000000000000e+00
```

Question 5

```
fprintf(' ---- Question 5 ---- ');
```

```
---- Question 5 ----
```

```
A = [4, -1, 0, 0;  
     -2, 4, -1, 0;  
     0, -2, 4, -1;  
     0, 0, -2, 4];  
  
[L, U, p] = gepphess(A);  
  
disp('Permutation vector p:');
```

```
Permutation vector p:
```

```
disp(p);
```

```
1 2 3 4
```

```
disp('Lower triangular matrix L:');
```

Lower triangular matrix L:

```
disp(L);
```

```
1.000000000000000e+00 0 0 0
-5.000000000000000e-01 1.000000000000000e+00 0 0
0 -5.714285714285714e-01 1.000000000000000e+00 0
0 0 -5.833333333333333e-01 1.000000000000000e+00
```

```
disp('Upper triangular matrix U:');
```

Upper triangular matrix U:

```
disp(U);
```

```
4.000000000000000e+00 -1.000000000000000e+00 0 0
0 3.500000000000000e+00 -1.000000000000000e+00 0
0 0 3.428571428571429e+00 -1.000000000000000e+00
0 0 0 3.416666666666667e+00
```

```
A_permuted = A(p, :);
```

```
A_reconstructed = L * U;
```

```
disp('A(p,:) - L * U (should be zero matrix):');
```

A(p,:) - L * U (should be zero matrix):

```
disp(A_permuted - A_reconstructed);
```

```
0 0 0 0
0 0 0 0
0 0 0 0
0 0 0 0
```

Question 7

```
fprintf(' ---- Question 7 ---- ');
```

---- Question 7 ----

```
A1 = [1 1 1; -1 9 2; 0 -1 2];
```

```
A2 = [1 1 1; -1 9 2; -4 -1 2];
```

```
A3 = [1 1 1; -1 3 2; -4 -1 2];
```

```
x = [1; 1; 1];
```

```
k = 20;
```

```
matrices = {A1, A2, A3};
```

```
matrix_names = {'A1', 'A2', 'A3'};
```

```

for m = 1:length(matrices)
    A = matrices{m};
    fprintf('--- Matrix %s ---\n', matrix_names{m});

    [v, lam] = eig_simplified(A);
    lam1 = lam(1);
    lam2 = lam(2);

    rate_power = abs(lam2 / lam1);

    % Power Method
    [iter_power, lambda_power] = PowerMethod(A, x, k);
    errors_power = zeros(k,1);
    for i = 1:k
        errors_power(i) = norm(iter_power(:,i) - v1);
    end

    % Shift and Invert Method
    s = lam1 - 0.1;
    [iter_shiftinv, lambda_shiftinv] = Shiftinv(A, x, s, k);
    errors_shiftinv = zeros(k,1);
    for i = 1:k
        errors_shiftinv(i) = norm(iter_shiftinv(:,i) - v1);
    end

    % Rayleigh Quotient Iteration
    [iter_rayleigh, lambda_rayleigh] = Rayleigh(A, x, k);
    errors_rayleigh = zeros(k,1);
    for i = 1:k
        errors_rayleigh(i) = norm(iter_rayleigh(:,i) - v1);
    end

    % Plotting the Convergence
    figure;
    semilogy(1:k, errors_power, 'b-', 'LineWidth', 2);
    hold on;
    semilogy(1:k, errors_shiftinv, 'r--', 'LineWidth', 2);
    semilogy(1:k, errors_rayleigh, 'g-.', 'LineWidth', 2);
    xlabel('Iteration');
    ylabel('Error (log scale)');
    title(['Convergence Comparison for ', matrix_names{m}]);
    legend('Power Method', 'Shift and Invert', 'Rayleigh Quotient');
    grid on;

    fprintf('Power Method Eigenvalue: %.6f\n', lambda_power);
    fprintf('Shift and Invert Eigenvalue: %.6f\n', lambda_shiftinv);
    fprintf('Rayleigh Quotient Eigenvalue: %.6f\n', lambda_rayleigh);
    fprintf('True Dominant Eigenvalue: %.6f\n', lam1);

```

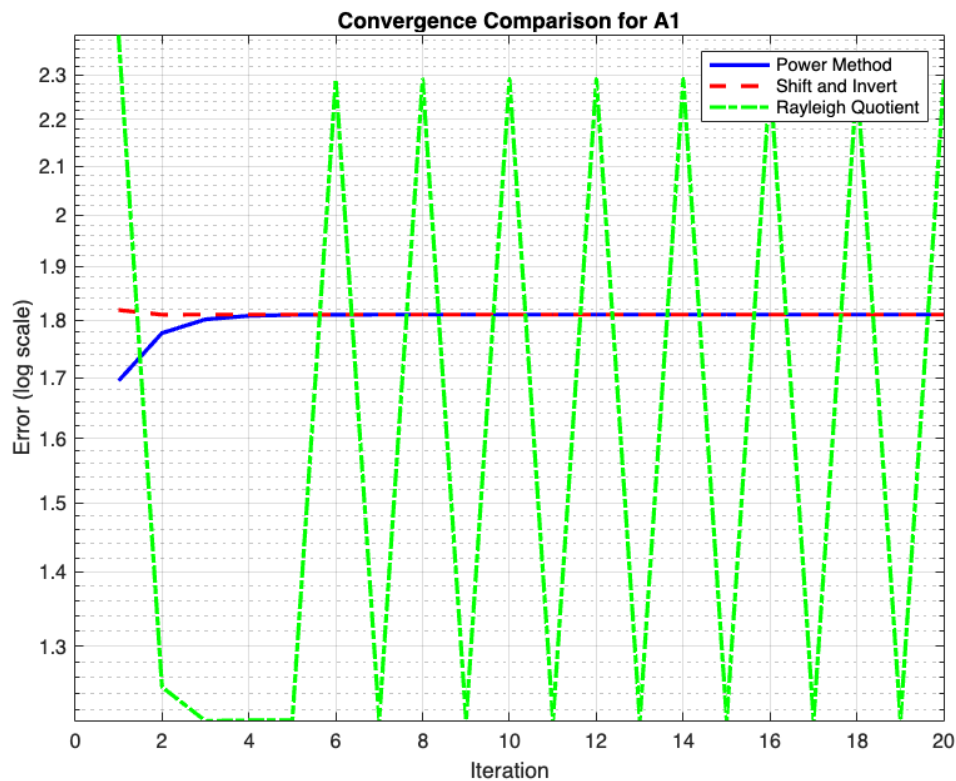
where do you define v_1 ?

```
fprintf('Power Method Convergence Rate: %.6f\n', rate_power);
fprintf('\n');
```

end

--- Matrix A1 ---

Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 1.395170e-17.
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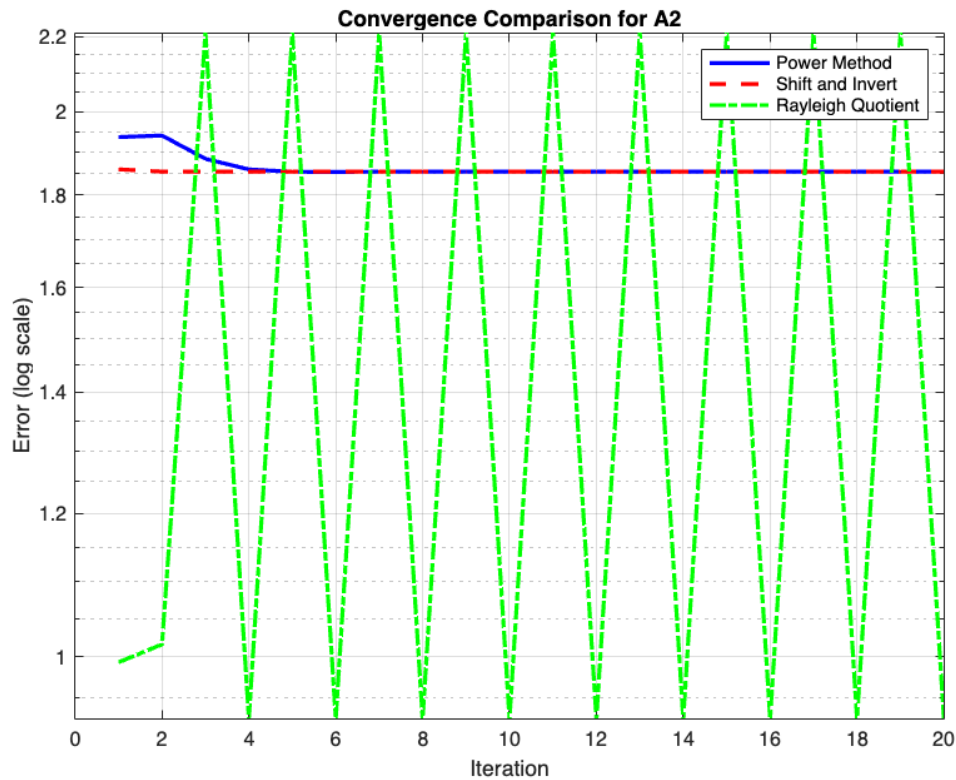


Power Method Eigenvalue: 8.584428
Shift and Invert Eigenvalue: 8.584428
Rayleigh Quotient Eigenvalue: 2.194882
True Dominant Eigenvalue: 8.584428
Power Method Convergence Rate: 0.255682

--- Matrix A2 ---

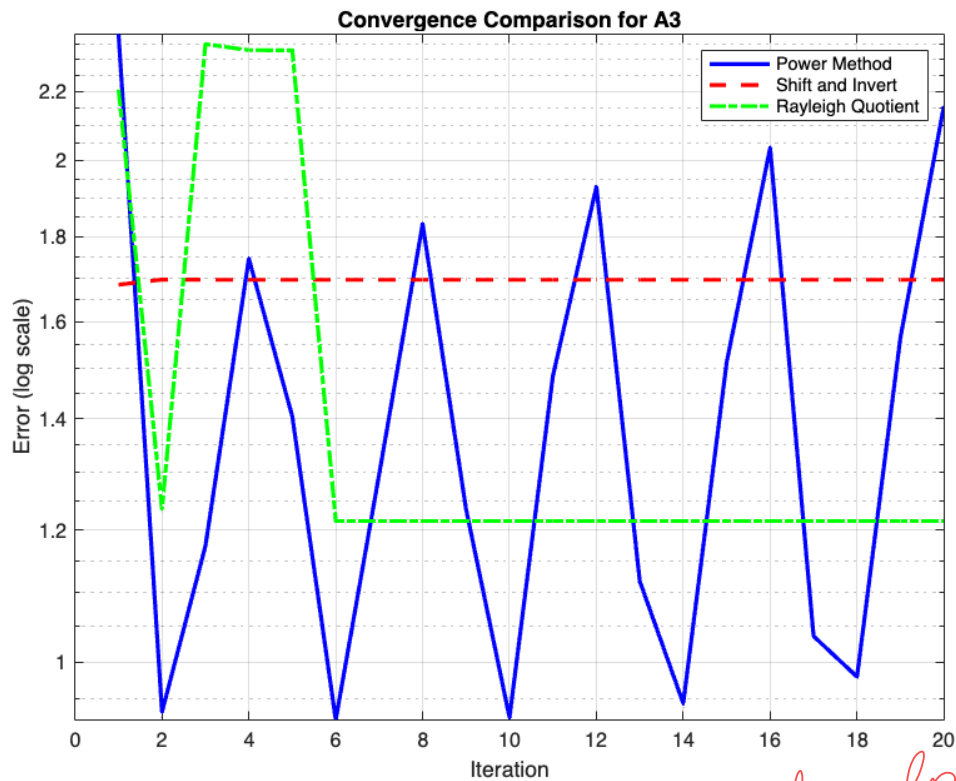
Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 1.247086e-16.
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Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 1.247086e-16.



Power Method Eigenvalue: 8.455873
Shift and Invert Eigenvalue: 8.455873
Rayleigh Quotient Eigenvalue: 8.455873
True Dominant Eigenvalue: 8.455873
Power Method Convergence Rate: 0.290434
--- Matrix A3 ---

Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 2.692686e-17.
Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 2.692686e-17.
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Warning: Matrix is close to singular or badly scaled. Results may be inaccurate. RCOND = 2.692686e-17.



Power Method Eigenvalue: 3.593669
 Shift and Invert Eigenvalue: 2.380066
 Rayleigh Quotient Eigenvalue: 1.239868
 True Dominant Eigenvalue: 2.380066
 Power Method Convergence Rate: 1.000000

what do you observe?

Functions used:

Question 1

```
function [iter, lambda] = PowerMethod(A, x, k)
    % Initialize variables
    n = length(x);
    iter = zeros(n, k);
    % Power iteration loop
    for i = 1:k
        x = A * x;
        [lambda, x] = normalize(x); % Matrix-vector multiplication
        % Normalize x and extract scaling factor as
        iter(:, i) = x; % Store iteration result
    end
end

function [lambda, x] = normalize(x)
    % Normalize x and return the scaling factor (lambda)
    [~, idx] = max(abs(x)); % Find index with max magnitude
```

need to scale x before computing Ax.


```

lambda = x(idx);           % Scaling factor as eigenvalue approximation
x = x / lambda;           % Normalize vector x
end

```

Question 2

```

function [v, lam] = eig_simplified(A)
    % Compute eigenvalues and eigenvectors
    [V, eigenvalues] = eig(A, 'vector');

    % Sort eigenvalues by absolute value in descending order
    [~, idx] = sort(abs(eigenvalues), 'descend');
    lam = eigenvalues(idx); % Sorted eigenvalues
    v = V(:, idx(1));       % Dominant eigenvector

    % Normalize the dominant eigenvector
    v = v / max(abs(v));
end

```

the scaling factor is wrong.

Question 3

```

function [v1, v2, v3, lam1, lam2, lam3] = eig_simplified_full(A)
    % Compute eigenvalues and eigenvectors
    [V, eigenvalues] = eig(A, 'vector');

    % Sort eigenvalues by absolute magnitude in descending order
    [~, idx] = sort(abs(eigenvalues), 'descend');

    % Extract top three eigenvalues and corresponding eigenvectors
    lam1 = eigenvalues(idx(1));
    lam2 = eigenvalues(idx(2));
    lam3 = eigenvalues(idx(3));
    v1 = normalize_vector(V(:, idx(1)));
    v2 = normalize_vector(V(:, idx(2)));
    v3 = normalize_vector(V(:, idx(3)));
end

function v = normalize_vector(v)
    % Normalize vector by its maximum absolute component
    v = v / max(abs(v));
end

```

Question 4

```

function [iter, lambda] = Shiftinv(A, x, s, k)
    n = length(x);
    iter = zeros(n, k);

```

```
I = eye(n);
[L, U] = lu(A - s * I);
```

$$[L, U, P] = \text{lu}(A - sI)$$

```
% Normalize x with a different technique
x = x / norm(x, Inf);
```

```
for j = 1:k
    % Solve Ly = x, then Uz = y in one step
    y = L \ x;
    x = U \ y;
```

```
% Renormalize to prevent overflow or underflow
x = x / norm(x, Inf);
```

```
iter(:, j) = x;
```

```
end
```

```
% Compute the Rayleigh quotient for lambda
lambda = (x' * A * x) / (x' * x);
```

```
end
```

The scaling factor is wrong

Question 5

```
function [L, U, p] = gepphess(A)
    n = size(A, 1);
    p = 1:n;

    for i = 1:n-1
        [~, max_idx] = max(abs(A(i:i+1, i)));
        if max_idx == 2
            p([i, i + 1]) = p([i + 1, i]);
            A([i, i + 1], :) = A([i + 1, i], :);
        end

        if A(i, i) == 0
            continue
        end

        A(i + 1, i) = A(i + 1, i) / A(i, i);
        A(i + 1, i + 1 : end) = A(i + 1, i + 1 : end) - A(i + 1, i) .* A(i, i + 1 : end);
    end;
    L = eye(n) + tril(A, -1);
    U = triu(A);
end
```

Question 6

```
function [iter, lambda] = Rayleigh(A, x, k)
```

```

n = length(x);
iter = zeros(n, k);

% Check if A is nearly upper Hessenberg; if not, convert
if ~isequal(A, triu(A, -1))
    H = hess(A);
else
    H = A;
end

% Initial normalization
x = x / norm(x, Inf);

for j = 1:k
    % Rayleigh quotient to approximate the eigenvalue
    lambda = (x' * H * x) / (x' * x);
    ShiftedMatrix = H - lambda * eye(n);

    % Factorize shifted matrix with partial pivoting
    [L, U, p] = gepphess(ShiftedMatrix); % Updated to match the function name
used previously
    x_perm = x(p);

    % Solve Ly = x_perm, Uz = y
    y = L \ x_perm;
    x = U \ y;

    % Renormalize x to maintain numerical stability
    x = x / norm(x, Inf);

    % Store the current iteration vector
    iter(:, j) = x;
end
end

```