

pr2_math548_final_Lazizbek

May 2, 2024

1 Math 548, Final Exam. Problem 1. LS

(a)

For which values of b , the matrix A is positive definite? (answer is on paper)

$$\begin{bmatrix} 1 & -1/2 & b \end{bmatrix}$$

$$A = \begin{bmatrix} -1/2 & 2 & -1/2 \end{bmatrix}$$

$$\begin{bmatrix} b & -1/2 & 2 \end{bmatrix}$$

(b) Find the LU decomposition of this matrix when $b=0$, using the Gaussian elimination method. Write down the matrices L and U . Explain and justify every step in the process.

1.

$$\begin{bmatrix} 1 & -1/2 & b \end{bmatrix} \begin{bmatrix} x_1 \end{bmatrix} = \begin{bmatrix} 0 \end{bmatrix}$$

$$\begin{bmatrix} -1/2 & 2 & -1/2 \end{bmatrix} \begin{bmatrix} x_2 \end{bmatrix} = \begin{bmatrix} 0 \end{bmatrix}$$

$$\begin{bmatrix} b & -1/2 & 2 \end{bmatrix} \begin{bmatrix} x_3 \end{bmatrix} = \begin{bmatrix} 0 \end{bmatrix}$$

(c)

Now solve the system of equations given by

2.

$$\begin{bmatrix} 1 & -1/2 & b \end{bmatrix} \begin{bmatrix} x_1 \end{bmatrix} = \begin{bmatrix} 2 \end{bmatrix}$$

$$\begin{bmatrix} -1/2 & 2 & -1/2 \end{bmatrix} \begin{bmatrix} x_2 \end{bmatrix} = \begin{bmatrix} 3 \end{bmatrix}$$

$$\begin{bmatrix} b & -1/2 & 2 \end{bmatrix} \begin{bmatrix} x_3 \end{bmatrix} = \begin{bmatrix} 14 \end{bmatrix}$$

2 (b). I am introducing $b = [0, 0, 0]$ RHS vector to perform LU decomposition on the matrix A

```
[1]: import numpy as np

def lu_decomposition(A):
    # Function to perform LU decomposition of matrix A
    n = len(A)
    L = np.zeros((n, n)) # Initialize lower triangular matrix L
    U = np.zeros((n, n)) # Initialize upper triangular matrix U
    P = np.identity(n)   # Initialize permutation matrix P

    # Main loop for LU decomposition with partial pivoting
    for i in range(n):
        # Partial pivoting: find pivot row with maximum absolute value
        # pivot_row = max(range(i, n), key=lambda k: abs(A[k][i]))

        pivot_row_max_value = -1 # Initialize with a minimum value
        for k in range(i, n): # Iterate through rows from i to n-1
            absolute_value = abs(A[k][i]) # Compute the absolute value of
            ↪ element at column i and row k
            if absolute_value > pivot_row_max_value: # Check if the absolute
            ↪ value is greater than current maximum
                pivot_row_max_value = absolute_value # Update maximum absolute
            ↪ value
                pivot_row = k # Update the row index with the maximum absolute
            ↪ value

        # Swap current row with pivot row in A, P
        A[[i, pivot_row]] = A[[pivot_row, i]]
        P[[i, pivot_row]] = P[[pivot_row, i]]

        # Set diagonal of L to 1
        L[i][i] = 1
        # Calculate entries of U and L using formulae for LU decomposition
        for j in range(i, n):
            U[i][j] = A[i][j] - sum(L[i][k] * U[k][j] for k in range(i))
        for j in range(i + 1, n):
            L[j][i] = (A[j][i] - sum(L[j][k] * U[k][i] for k in range(i))) /
            ↪ U[i][i]

    return P, L, U

def forward_substitution(L, b):
    # Function to perform forward substitution to solve Ly = b
    n = len(b)
```

```

y = np.zeros(n) # Initialize solution vector y
for i in range(n):
    # Compute y[i] using forward substitution formula
    y[i] = b[i] - np.dot(L[i, :], y[:i])
return y

def back_substitution(U, y):
    # Function to perform back substitution to solve  $Ux = y$ 
    n = len(y)
    x = np.zeros(n) # Initialize solution vector x
    for i in range(n - 1, -1, -1):
        # Compute x[i] using back substitution formula
        x[i] = (y[i] - np.dot(U[i, i+1:], x[i+1:])) / U[i, i]
    return x

# Given matrix A and vector b
A = np.array([[ 2, -1/2,  0], [-1/2,  2, -1/2], [ 0, -1/2,  2]])
b = np.array([ 0,  0,  0])

# Keep the original A and b
original_A = A
original_b = b

# Perform LU decomposition of A
P, L, U = lu_decomposition(A)

# Permute b according to pivot matrix P
b_permuted = np.dot(P, b)

# Solve  $Ly = b_{\text{permuted}}$  using forward substitution
y = forward_substitution(L, b_permuted)

# Solve  $Ux = y$  using back substitution
x = back_substitution(U, y)

# Verify if  $A = LU$ 
is_A_LU = np.allclose(original_A, np.dot(L, U))

# Calculate the determinant of A
det_A = np.linalg.det(original_A)

# Construct the augmented matrix  $[A|b]$ 
augmented_matrix = np.hstack((original_A, np.expand_dims(original_b, axis=1)))

# Print all intermediate and final results
print("\nAugmented matrix  $[A|b]$ :")
print(augmented_matrix)

```

```

print("\nSolution vector x:")
print(x)
print("\nPermutation matrix P:")
print(P)
print("\nLower triangular matrix L:")
print(L)
print("\nUpper triangular matrix U:")
print(U)
print("\nIs A equal to LU:", is_A_LU)
print("\nDeterminant of A:", det_A)

```

Augmented matrix $[A|b]$:

```

[[ 2.  -0.5  0.   0. ]
 [-0.5  2.  -0.5  0. ]
 [ 0.  -0.5  2.   0. ]]

```

Solution vector x:

```

[0. 0. 0.]

```

Permutation matrix P:

```

[[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]]

```

Lower triangular matrix L:

```

[[ 1.         0.         0.         ]
 [-0.25       1.         0.         ]
 [ 0.        -0.26666667  1.         ]]

```

Upper triangular matrix U:

```

[[ 2.        -0.5         0.         ]
 [ 0.         1.875       -0.5         ]
 [ 0.         0.         1.86666667]]

```

Is A equal to LU: True

Determinant of A: 6.999999999999998

3 (c) Now solve the system of equations given by

```

[2]: import numpy as np

def lu_decomposition(A):
    # Function to perform LU decomposition of matrix A
    n = len(A)

```

```

L = np.zeros((n, n)) # Initialize lower triangular matrix L
U = np.zeros((n, n)) # Initialize upper triangular matrix U
P = np.identity(n)   # Initialize permutation matrix P

# Main loop for LU decomposition with partial pivoting
for i in range(n):
    # Partial pivoting: find pivot row with maximum absolute value
    # pivot_row = max(range(i, n), key=lambda k: abs(A[k][i]))

    pivot_row_max_value = -1 # Initialize with a minimum value
    for k in range(i, n): # Iterate through rows from i to n-1
        absolute_value = abs(A[k][i]) # Compute the absolute value of
        ↪ element at column i and row k
        if absolute_value > pivot_row_max_value: # Check if the absolute
        ↪ value is greater than current maximum
            pivot_row_max_value = absolute_value # Update maximum absolute
            ↪ value
            pivot_row = k # Update the row index with the maximum absolute
            ↪ value

    # Swap current row with pivot row in A, P
    A[[i, pivot_row]] = A[[pivot_row, i]]
    P[[i, pivot_row]] = P[[pivot_row, i]]

    # Set diagonal of L to 1
    L[i][i] = 1
    # Calculate entries of U and L using formulae for LU decomposition
    for j in range(i, n):
        U[i][j] = A[i][j] - sum(L[i][k] * U[k][j] for k in range(i))
    for j in range(i + 1, n):
        L[j][i] = (A[j][i] - sum(L[j][k] * U[k][i] for k in range(i))) /
        ↪ U[i][i]

    return P, L, U

def forward_substitution(L, b):
    # Function to perform forward substitution to solve Ly = b
    n = len(b)
    y = np.zeros(n) # Initialize solution vector y
    for i in range(n):
        # Compute y[i] using forward substitution formula
        y[i] = b[i] - np.dot(L[i, :i], y[:i])
    return y

def back_substitution(U, y):
    # Function to perform back substitution to solve Ux = y
    n = len(y)

```

```

x = np.zeros(n) # Initialize solution vector x
for i in range(n - 1, -1, -1):
    # Compute x[i] using back substitution formula
    x[i] = (y[i] - np.dot(U[i, i+1:], x[i+1:])) / U[i, i]
return x

# Given matrix A and vector b
A = np.array([[ 2, -1/2,  0], [-1/2,  2, -1/2], [ 0, -1/2,  2]])
b = np.array([ 2,  3, 14])

# Keep the original A and b
original_A = A
original_b = b

# Perform LU decomposition of A
P, L, U = lu_decomposition(A)

# Permute b according to pivot matrix P
b_permuted = np.dot(P, b)

# Solve Ly = b_permuted using forward substitution
y = forward_substitution(L, b_permuted)

# Solve Ux = y using back substitution
x = back_substitution(U, y)

# Verify if A = LU
is_A_LU = np.allclose(original_A, np.dot(L, U))

# Calculate the determinant of A
det_A = np.linalg.det(original_A)

# Construct the augmented matrix [A|b]
augmented_matrix = np.hstack((original_A, np.expand_dims(original_b, axis=1)))

# Print all intermediate and final results
print("\nAugmented matrix [A|b]:")
print(augmented_matrix)
print("\nSolution vector x:")
print(x)
print("\nPermutation matrix P:")
print(P)
print("\nLower triangular matrix L:")
print(L)
print("\nUpper triangular matrix U:")
print(U)
print("\nIs A equal to LU:", is_A_LU)

```

```
print("\nDeterminant of A:", det_A)
```

Augmented matrix [A|b]:

```
[[ 2. -0.5  0.  2. ]
 [-0.5  2. -0.5  3. ]
 [ 0. -0.5  2. 14. ]]
```

Solution vector x:

```
[2. 4. 8.]
```

Permutation matrix P:

```
[[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]]
```

Lower triangular matrix L:

```
[[ 1.          0.          0.          ]
 [-0.25        1.          0.          ]
 [ 0.          -0.26666667  1.          ]]
```

Upper triangular matrix U:

```
[[ 2.          -0.5          0.          ]
 [ 0.           1.875        -0.5          ]
 [ 0.           0.           1.86666667]]
```

Is A equal to LU: True

Determinant of A: 6.999999999999998

```
[4]: # !sudo apt-get install texlive-xetex texlive-fonts-recommended
      ↪ texlive-plain-generic
```

```
[5]: # !jupyter nbconvert --to pdf /content/Math548_hw7_Lazizbek.ipynb
```