

LINER EQUATION AND LU DECOMPOSITION

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Assignment

Given Matrix equation

$$\mathbf{MX} = \mathbf{b}$$

Get Augmented matrix $[\mathbf{M}|\mathbf{b}]$

convert it into Echelon Form

0.1 Echelon Form

Algorithm for converting any $m \times n$ matrix

- for i the iteration -
 - consider a submatrix considering (ith row to all row)*(ith col to all col)

Listing 1: Python Code for the Plot

```
1 def calc_ecl(m):
2     temp_mat = np.copy(m)
3     for i in range(len(temp_mat)):
4         temp_mat[i:,i:] = update_sub_mat(temp_mat[i:,i:])
5     return(temp_mat)
```

- Using 1st row of this submatrix, for all the subsequent rows of this :

```
1 def update_sub_mat(mat):
2     m_temp = []
3     m_temp.append(mat[0])
4     for i in range(1,np.shape(mat)[0]):
5         m_temp.append(update_row(mat[i],mat[0]))
6     return (np.asarray(m_temp) )
```

- * Make 1st element 0

```
1 def update_row(r2,r1):
2     coeff = r2[0]/r1[0]
3     r_temp = r2 - coeff*r1
4     return(r_temp)
```

Listing 2: Python Code for the Plot

- do this starting from largest submatrix, the given matrix itself.

0.2 Linear Equation

- Convert given equation into augmented matrix form
- Convert this augmented matrix to Echelon Form
- use back substitution to solve the given matrix

0.2.1 Back-substitution

Algorithm:

- for $i:[n-1,0]$
 - Consider i^{th} row of the augmented matrix and starting from i th element of this row
 - this will be $[a_{ii}, \dots, a_{in} | b_n]$
 - assume at i th iteration we have solution till $[x_{i+1}, \dots, x_n]$, and we need to calculate x_i we are starting from calculating x_n and moving towards x_o
 - pass this to `calc_row_sol()`
 - * extract coefficient vector from $(i+1)$ th element of this row upto 2nd last term
 $coeff_vec = [a_{i,i+1}, \dots, a_{i,n}]$
 - * $const = b_i$
 - * now, x_i will be simply
 $x_i = (b_i - coeff_vec \cdot sol_vec) / a_{ii}$
 - * Augment this x_i to the starting of sol_vec

implementation:

```

1 def calc_row_sol(mat_vec, sol_vec):
2     sol = (mat_vec[-1] - np.dot(mat_vec[1:-1], sol_vec)) / mat_vec[0]
3     return(sol)
4
5 def solve_mat_eqn(m, b_vec):
6     if(np.shape(m)[0] != len(b_vec)):
7         raise ValueError('Matrix and vector dim mismatch')
8     else:
9         mat_given = np.column_stack((m, b_vec))
10        mat = calc_ecl(mat_given)
11        sol_vec = []
12        for i in reversed(range(np.shape(mat)[0])):
```

```

13         sol_vec.insert(0, calc_row_sol(mat[i,i:], sol_vec))
14     return(sol_vec)

```

0.3 LU decomposition

Factorizing a given matrix in Lower and Upper triangle matrices :

0.3.1 Using Equation solution

Here I will use the linear equation solution developed in previous section

$$\mathbf{M} = \mathbf{L} \times \mathbf{U}$$

Algorithm

- Convert \mathbf{M} in echelon form , and assign it to \mathbf{U}
- transpose \mathbf{U} say , \mathbf{U}^T
- Now i th row of \mathbf{L} will be simply the solution of the matrix equation

$$\mathbf{U}^T \times \mathbf{x} = m$$

where m is the i^{th} row of \mathbf{M}

implementation

```

1  from ecln import calc_ecl
2  from linear_solver_v2 import solve_mat_eqn
3
4  def lu_decomp(m):
5      u = calc_ecl(m)
6      u_t = (np.copy(u)).transpose()
7      l = [solve_mat_eqn(u_t, m_vec) for m_vec in m]
8      return(np.asarray(l), u)

```

0.3.2 Doolittle's Algorithm

:

- Initialize $l_{n \times n}$ and $u_{n \times n}$ as Zero matrices
- make diagonal elements of $l = 1$
- For $k = 1 : n$

– for $m = k, k + 1, \dots, n$

$$u_{k,m} = a_{k,m} - \sum_{j=1}^{k-1} l_{kj} u_{jm}$$

– for $i = k + 1, \dots, m$

$$l_{ik} = \frac{a_{ik} - \sum_{j=1}^{k-1} l_{ij} u_{jk}}{u_{kk}}$$

```

1  import numpy as np
2
3  '''
4  lower and upper triangle decomposition
5  using Doolittle's Algorithm
6  '''
7
8  def lu_decomp_doolittle(m):
9      m = np.asarray(m)
10     n = np.shape(m)[0]
11     l = np.zeros((3,3))
12     u = np.zeros((3,3))
13
14     for k in range(n):
15         for m in range(k,n):
16             s = 0
17             for j in range(k):
18                 s += l[k][j]*u[j][m]
19             u[k][m] = a[k][m] - s
20
21         for i in range(k,n):
22             if(i==k):l[i][k] = 0
23             else:
24                 s = 0
25                 for j in range(k):
26                     s = l[i][j]*u[j][k]
27                 l[i][k] = (a[i][k]-s)/u[k][k]
28
29     return(np.asarray(l),np.asarray(u))
30
31
32 a = [[2.0,3.0,-4.0],[1.0,5.0,-1.0],[3.0,7.0,-3.0]]
33 l , u = lu_decomp_doolittle(a)
34
35 print('lower:\n' , l)
36 print('upper:\n' , u)

```
