

## Pivoting

The starting point of this tutorial is the  $LU$  decomposition code (`LU_scratch.py`) that we wrote in class. We found that it fails when the input matrix has a singular sub-matrix, even when the input matrix itself is invertible. For instance

$$A = \begin{pmatrix} 2 & 2 & 1 \\ 2 & 2 & -1 \\ 1 & -1 & 0 \end{pmatrix}$$

has the sub-matrix

$$\begin{pmatrix} 2 & 2 \\ 2 & 2 \end{pmatrix},$$

which is singular and therefore our  $LU$  decomposition code encounters a division by zero.

The solution to this problem is “pivoting”, described in Lecture 6. In particular, we can do  $LU$  decomposition with partial pivoting, which involves finding a permutation matrix  $P$  such that  $PA = LU$ . The pseudo-code for  $LU$  decomposition with partial pivoting can be found in the Lecture 6 Slides. In this tutorial, we will implement this algorithm following the steps below.

- Write a function that swaps two rows of a matrix. Inputs should be a  $n \times n$  array  $M$  and two indices  $i$  and  $j$ , where  $0 \leq i, j \leq n - 1$  (following the Python convention that indices start from 0). Output should be the array  $M$  with rows  $i$  and  $j$  swapped.
- The function you have created will be good for the swapping rows as required for  $P$  and  $U$  (i.e. all elements of the rows get moved). For  $L$  you must only swap the elements of the appropriate rows that are *before* the column with the pivot. Modify your row swapping function so that only the first  $k$  elements of the rows get swapped, where  $k$  will be an additional input to your function. Note, you can also use this function on  $P$  and  $U$ , with the appropriate value of  $k$ .
- There is a line in the pseudo-code that says “Select  $k \geq j$  to maximize  $|U_{k,j}|$ .” Read the documentation on the `argmax` function on [numpy.org](https://numpy.org/doc/stable/reference/generated/numpy.argmax.html) to figure out how to find the index of the largest element in an array. Work out how to use this function to implement the line in the pseudo-code mentioned above.
- Now implement the pseudo-code for  $LU$  decomposition with pivoting. Input should be a  $n \times n$  array  $A$ , output should be  $n \times n$  arrays  $P$ ,  $L$  and  $U$  so that  $P$  is a permutation matrix and  $PA = LU$ . Note that the last loop in the pseudo-code with pivoting is the same as without. So you may find the `LU_scratch.py` code helpful.
- Test it by decomposing the matrix  $A$  above and verifying that  $PA = LU$ . Also, test the matrix from class (see `LU_scratch.py`) that didn’t work when we didn’t use pivoting.
- **Bonus:** Memory and time savers: (a) When you swap whole rows of  $U$ , you’re actually swapping lots of zeros. The first  $j - 1$  elements of both of the rows you’re swapping are zeros, and so it’s better to avoid swapping these elements (note this is already written into the pseudo-code). Modify your row swapping function to incorporate this efficiency. (b) Modify your code so that you store the permutation matrix as a vector; see last part of Lecture 6 slides. This memory saver can be very significant when the size of your matrices is large.