1. Find the LU factorization of the following matrices A.

(a)
$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -1 \\ -3 & 1 & 1 \end{bmatrix}$$
.
(b) $A = \begin{bmatrix} 1 & 1 \\ 3 & -4 \end{bmatrix}$.

2. Use the previous LU factorization of the matrices A to solve the systems below. No credit will be given for any other method.

$$x + 2y - z = 3$$

$$2x + y - 2z = 3$$

$$-3x + y + z = -6.$$

(b)

$$x_1 + x_2 = 3 3x_1 - 4x_2 = 2$$

- 3. Write code to eliminate in the first column of a 100×100 matrix without pivoting and using one loop.
- 4. Use PA = LU factorization to solve the following systems. No credit will be given for any other method. (a)

$$3x + y + 2z = 0$$

$$6x + 3y + 4z = 1$$

$$3x + y + 5z = 3.$$

(b)

$$3x_1 + 7x_2 = 1 6x_1 + x_2 = -11$$

- 5. (a) Find the PA = LU factorization of $A = \begin{bmatrix} 2 & 1 & 5 \\ 4 & 4 & -4 \\ 1 & 3 & 1 \end{bmatrix}$. (b) Solve $Ax = b = \begin{bmatrix} 5 \\ 0 \\ 6 \end{bmatrix}$ using the above PA = LU factorization.

6. (a) Show that $A = \begin{bmatrix} 1 & 2 \\ 3 & 10 \end{bmatrix}$ is not symmetric.

- (b) Show that $A = \begin{bmatrix} 1 & 2 \\ 2 & 2 \end{bmatrix}$ is symmetric but is not positive definite. (c) Show that $A = \begin{bmatrix} 1 & 3 \\ 3 & 10 \end{bmatrix}$ is symmetric but is not positive definite.
- 7. Find the Cholesky factorization, $A = GG^T$, of each of the following symmetric, positive definite matrices using the algorithm on p. 116.

(a)
$$A = \begin{bmatrix} 1 & 2 \\ 2 & 8 \end{bmatrix}$$

(b) $A = \begin{bmatrix} 25 & 5 \\ 5 & 26 \end{bmatrix}$

8. Use the previous Cholesky factorization and then forward and back substitution to solve

$$x_1 + 2x_2 = 4 2x_1 + 8x_2 = 2$$

(No credit will be given for any other method.)

9. Find the Cholesky factorization, $A = GG^T$, of the symmetric, positive definite matrices $A = \begin{bmatrix} 9 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 8 \end{bmatrix}$

using $G = LD^{\frac{1}{2}}$.

- 10. Given $A = \begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix}$ and $B = \begin{bmatrix} .0001 & 1 \\ 1 & 1 \end{bmatrix}$.
 - (a) Roughly describe why matrix A is ill-conditioned and matrix B is well-conditioned.
 - (b) Use the numpy command *cond* to approximate the norm 2 condition numbers $\kappa_2(A)$ and $\kappa_2(B)$. Does the numpy approximate fit your rough idea above?
- 11. Find the norms $||A||_2 = \lambda_1$ and the condition numbers $\kappa(A) = \frac{\lambda_1}{\lambda_2}$ of the following positive definite matrices (by hand):

(a)
$$A = \begin{bmatrix} 100 & 0 \\ 0 & 2 \end{bmatrix}$$
.
(b) $A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$.
(c) $A = \begin{bmatrix} 3 & 1 \\ 1 & 1 \end{bmatrix}$.

12. Find the norms $||A||_2$ and the condition numbers $\kappa(A)$ of the following positive definite matrices (by hand) from $\sqrt{\lambda_1(A^T A)}$ and $\sqrt{\lambda_n(A^T A)}$:

(a)
$$A = \begin{bmatrix} -2 & 0 \\ 0 & 2 \end{bmatrix}$$
.
(b) $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$.
(c) $A = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$.

13. Given the ill-conditioned matrix $A = \begin{bmatrix} 1 & 1 \\ 1 & 1.0001 \end{bmatrix}$.

- (a) Solve Ax = b for $b = \begin{vmatrix} 2 \\ 2 \end{vmatrix}$.
- (b) Solve Ax = b for $b = \begin{bmatrix} 2\\ 2.0001 \end{bmatrix}$.
- (c) You should see that the solutions, x, to this ill-conditioned problem are very sensitive to small changes in the input b. There is no robust algorithm.

14. Even well conditioned problems can give errors. Given $B = \begin{bmatrix} .0001 & 1 \\ 1 & 1 \end{bmatrix}$.

(a) Solve
$$Bx = b$$
 for $b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ using no pivots with three digits of accuracy.

(b) Solve Bx = b for $b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ using partial pivoting with three digits of accuracy.

- 15. The system $\begin{bmatrix} 1 & 2 \\ 1.0001 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 3.0001 \end{bmatrix}$ has one solution $x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$. Suppose $\hat{x} = \begin{bmatrix} 3 \\ -0.0001 \end{bmatrix}$ is the approximate solution after running some algorithm.
 - (a) Compute the residual $\hat{r} = b A\hat{x}$, $\|\hat{r}\|_{\infty}$, and $\|\hat{r}\|_{2}$.
 - (b) Compute the relative error in the residual, $\frac{\|\hat{r}\|_2}{\|b\|_2}$. You may use numpy's LA.norm(r) command.
 - (c) Compute the relative forward error $\frac{\|x-\hat{x}\|_2}{\|x\|_2}$.
 - (d) Verify the bound $\frac{||x-\hat{x}||}{||x||} \le \kappa(A) \frac{||\hat{r}||}{||b||}$. (you may use numpy's LA.cond(A) command.)
- 16. TEXTBOOK EXERCISES: 1, 3, 4, 5.