

# COMPULSORY EXERCISE 1

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## 1. GIVENS ROTATIONS TO SOLVE $A\mathbf{x} = \mathbf{b}$

General idea, using roughly the same notation as in the book:

$$\begin{pmatrix} x & x & x & x & | & b_1 \\ x & x & x & x & | & b_2 \\ x & x & x & x & | & b_3 \\ x & x & x & x & | & b_4 \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{r} & x^* & x^* & x^* & | & b_1 \\ x & x & x & x & | & b_2 \\ x & x & x & x & | & b_3^* \\ \mathbf{0} & x^* & x^* & x^* & | & b_4^* \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{r}^* & x^{**} & x^{**} & x^{**} & | & b_1 \\ x & x & x & x & | & b_2^{**} \\ \mathbf{0} & x^* & x^* & x^* & | & b_3^* \\ 0 & x^* & x^* & x^* & | & b_4^* \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} r_{1,1} & r_{1,2} & r_{1,3} & r_{1,4} & | & c_1 \\ \mathbf{0} & x^{**} & x^{**} & x^{**} & | & b_2^{**} \\ 0 & x^{**} & x^{**} & x^{**} & | & b_3^{**} \\ 0 & x^* & x^* & x^* & | & b_4^* \end{pmatrix} \rightarrow \dots$$

with the asterix superscripts denoting the number of times that element has been changed via matrix multiplication.

In Python, this becomes:

```

1 def rotsolve(A, b):
2     """
3     Solve the system Ax=b where A any N x N matrix, again using
4     Givens rotations. Modified from rothesstri(A, b).
5
6     A: A matrix. Nonsingular ones only.
7     b: The right hand side.
8     """
9     n = shape(A)[0]
10    A = hstack([A, b])
11    for k in xrange(n-1): # columns
12        for j in xrange(n-2, k-1, -1): # rows
13            r = linalg.norm([A[j, k], A[j+1, k]])
14            if r>0:
15                c=A[j, k]/r; s=A[j+1, k]/r
16                A[[j, j+1],(k+1):(n+1)] = \
17                    [[c, s],[-s, c]]*A[[j, j+1],(k+1):(n+1)]
18            A[j, k] = r; A[j+1,k] = 0
19    z = A[:, n].copy()
20    rbacksolve(A[:, :n], z, n)
21    return z

```

The program itself is enclosed, obviously.

**An aside of arguable interest.** Comparing errors (problem 3) with a friend gave very different results; there are obviously many ways to implement this solver. The following is a perhaps more intuitive approach:

$$\begin{pmatrix} x & x & x & x & | & b_1 \\ x & x & x & x & | & b_2 \\ x & x & x & x & | & b_3 \\ x & x & x & x & | & b_4 \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{r} & x^* & x^* & x^* & | & b_1^* \\ x & x & x & x & | & b_2 \\ x & x & x & x & | & b_3 \\ \mathbf{0} & x^* & x^* & x^* & | & b_4^* \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{r} & x^* & x^* & x^* & | & b_1^{**} \\ x & x & x & x & | & b_2 \\ \mathbf{0} & x^* & x^* & x^* & | & b_3^* \\ 0 & x^* & x^* & x^* & | & b_4^* \end{pmatrix} \\ \rightarrow \begin{pmatrix} r_{1,1} & r_{1,2} & r_{1,3} & r_{1,4} & | & c_1 \\ \mathbf{0} & x^* & x^* & x^* & | & b_2^{**} \\ 0 & x^* & x^* & x^* & | & b_3^{**} \\ 0 & x^* & x^* & x^* & | & b_4^* \end{pmatrix} \rightarrow \dots$$

with code:

```

def rotsolveAlt(A, b):
    """
    Solve the system Ax=b where A any N x N matrix, again using
    Givens
    rotations. Modified from rothesstri(A, b).

    A: A matrix. Nonsingular ones preferred, obviously.
    b: The right hand side.
    """
    n = shape(A)[0]
    A = hstack([A, b])
    for k in xrange(n-1): # columns
        for j in xrange(k+1, n): # rows
            r = linalg.norm([A[k, k], A[j, k]], 2)
            if r>0:
                c=A[k, k]/r; s=A[j, k]/r
                A[[k, j],(k+1):(n+1)] = \
                    mat([ [c, s], [-s, c] ]) * \
                    A[[k, j],(k+1):(n+1)]
                A[k, k] = r; A[j, k] = 0
    z = A[:, n].copy()
    rbacksolve(A[:, :n], z, n)
    return z

```

## 2. COMPLEXITY OF THE ALGORITHM

Taking, for simplicity, the computation of  $r$  to be four operations (squaring, squaring, adding, taking the square root), and having that computing  $c$  and  $s$  each takes one operation, we have besides this each inner step in the algorithm (i.e. for each  $j$ ) containing a multiplication of a  $(2 \times 2)$  matrix with a  $(2 \times (n - k + 1))$  matrix, for a total of  $6(n - k + 1) + 6 = 6(n - k + 2)$  operations.

There are  $n - k$   $j$ -steps, making for a total of  $6(n - k + 2)(n - k)$  arithmetic operations per  $k$ -step. Thus, the total number of operations are

$$(2.1) \quad \sum_{k=1}^{n-1} 6(n - k + 2)(n - k) = 6 \sum_{l=1}^{n-1} l(l + 2)$$

$$(2.2) \quad \approx 6 \int_0^{n-1} l^2 + 2l dl = 2(n-1)^3 + 2(n-1) = 2n^3 - 6n^2 + 8n - 1 \approx \underline{2n^3}.$$

as  $n$  becomes large.

Finally, we end up with a system  $U\mathbf{x} = \mathbf{c}$ , where  $U$  is upper triangular, at which point the backsolver can be applied, but that algorithm has takes  $n^2$  operations, thus not changing the result.  $\square$

### 3. ALSO

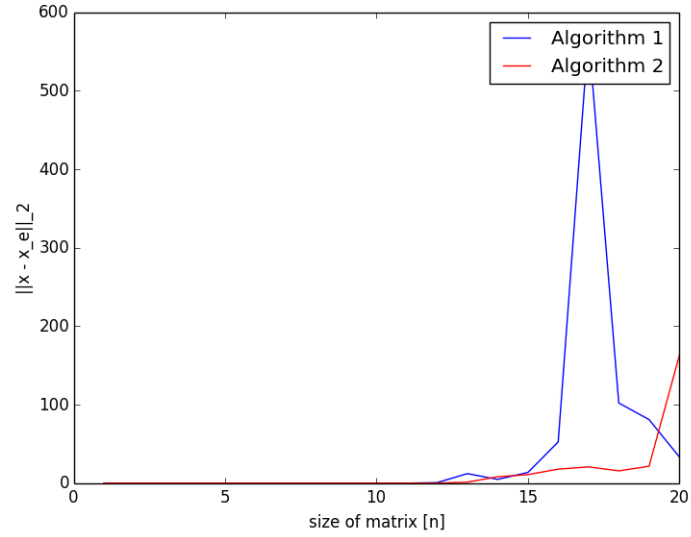
The code is enclosed, but:

```

1 if __name__ == '__main__':
2
3     N = 20
4
5     H_ = hilbert(N)
6     xe_ = mat(ones(N)).T
7     err = empty(N)
8     errAlt = empty(N)
9     iterations = arange(N)
10    for n in iterations:
11        # H = H_[:n+1, :n+1]; xe = xe_[:n+1]
12        H = hilbert(n+1); xe = mat(ones(n+1)).T
13        b = H*xe
14        x = rotsolve(H, b)
15        err[n] = linalg.norm(x - xe, 2)
16        y = rotsolveAlt(H, b)
17        errAlt[n] = linalg.norm(y - xe, 2)
18
19    plt.plot(1+iterations, err, 'b')
20    plt.plot(1+iterations, errAlt, 'r')
21    plt.legend(['Algorithm 1', 'Algorithm 2'])
22    plt.xlabel('size of matrix [n]')
23    plt.ylabel('||x - x_e||_2')
24    plt.show()

```

As noted, there were two implementations of `rotsolve`; I have plotted the er-



ror of both.