18.335 Problem Set 3

Problem 1: QR and orthogonal bases

- (a) Trefethen, problem 10.4.
- (b) Prove that A = QR and B = RQ have the same eigenvalues, assuming A is a square matrix. Then do a little experiment: Construct a random 5×5 real-symmetric matrix in Julia via X=rand(5,5); A = X' + X. Use QR = qr(A) (do using LinearAlgebra first) to compute the QR factorization of A, and then compute B = QR.R * QR.Q. Then find the QR factorization B = Q'R', and compute R'Q'...repeat this process until the matrix converges (maybe writing a loop and/or a function). From what it converges to, suggest a procedure to compute the eigenvalues and eigenvectors of a realsymmetric matrix (no need to prove that it converges in general—we will discuss this in class).
- (c) Trefethen, problem 28.2,

Problem 2: Schur fine

In class, we will show that any square $m \times m$ matrix A can be factorized as $A = QTQ^*$ (the Schur factorization), where Q is unitary and T is an upper-triangular matrix (with the same eigenvalues as A, since the two matrices are similar).

- (a) A is called "normal" if $AA^* = A^*A$. Show that this implies $TT^* = T^*T$. From this, show that T must be diagonal. Hence, any normal matrix (e.g. unitary or Hermitian matrices) must be unitarily diagonalizable. Hint: consider the diagonal entries of TT^* and T^*T , starting from the (1,1) entries and proceeding diagonally downwards by induction.
- (b) Given the Schur factorization of an arbitrary A (not necessarily normal), describe an algorithm to find the eigenvalues and eigenvectors of A, assuming for simplicity that all the eigenvalues are distinct. The flop count should be asymptotically $Km^3 + O(m^2)$; give the constant K.

Problem 3: Caches and backsubstitution

In this problem, you will consider the impact of caches (again in the ideal-cache model from class) on the problem of *backsubstitution*: solving Rx = b for x, where R is an $m \times m$ upper-triangular matrix (such as might be obtained by Gaussian elimination). The simple algorithm you probably learned in previous linear-algebra classes (and reviewed in the book, lecture 17) is (processing the rows from bottom to top):

$$\begin{array}{l} x_m = b_m/r_{mm} \\ \text{for } j = m-1 \text{ down to } 1 \\ x_j = (b_j - \sum_{k=j+1}^m r_{jk} x_k)/r_{jj} \end{array}$$

Suppose that X and B are $m \times n$ matrices, and we want to solve RX = B for X—this is equivalent to solving Rx = b for n different right-hand sides b (the n columns of B). One way to solve the RX = B for X is to apply the standard backsubstitution algorithm, above, to each of the n columns in sequence.

- (a) Give the asymptotic cache complexity Q(m,n;Z) (in asymptotic Θ notation, ignoring constant factors) of this algorithm for solving RX = B.
- (b) Suppose m = n. Propose an algorithm for solving RX = B that achieves a better asymptotic cache complexity (by cache-aware/blocking or cache-oblivious algorithms, your choice). Can you gain the factor of $1/\sqrt{Z}$ savings that we showed is possible for square-matrix multiplication?

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