

1. Cholesky Decomposition.

- (a) Write a function with argument n to generate a random symmetric $n \times n$ -positive definite matrix. To do this:
- generate an $n \times n$ matrix C whose entries are independent normal random variables;
 - return CC^T .

Check your matrices are positive definite using the `eigen()` function.

- (b) Implement the recursive Cholesky decomposition algorithm from the lecture.
- (c) Test it using your function for generating positive definite matrices, and by comparing the answers to `chol()`.
- (d) Create a function which takes a vector μ and a symmetric positive definite matrix Σ and uses them to generate a multivariate normal vector $N_n(\mu, \Sigma)$. Your function should check that Σ is positive definite using `eigen()` and symmetric using `isSymmetric()`.

2. Sorting. Here is an algorithm called ‘Quicksort’ for sorting the objects in a vector.

Function: sort a vector x

Input: vector x of length n

Output: a vector $Q(x)$ containing entries of x arranged in ascending order

1. if $n \leq 1$ return x ;
 2. pick an arbitrary ‘pivot’ element $i \leq n$;
 3. let $z = (x_j \mid x_j < x_i)$ and $y = (x_j \mid x_j > x_i)$;
 4. let $z' = Q(z)$ and $y' = Q(y)$; [*i.e. call the algorithm on the smaller vectors*]
 5. let x' be the entries in x not used in y or z ; [*i.e. any entries equal to x_i*]
 6. return (z', x', y') .
- (a) Implement the algorithm in R, and test it on some random numbers.
- (b) What is the complexity if x_i is always the smallest element?
- (c) Show that, if the pivot x_i is the median element on each call, that the complexity is at most $O(n \log_2(n))$.

3. Back Solving. Here is a recursive algorithm to solve $Ax = b$ where A is an upper triangular matrix, using back substitution.

Function: solve $Ax = b$ for x by back-substitution
Input: $n \times n$ upper triangular matrix A and vector b of length n
Output: vector x of length n solving $Ax = b$

1. If $n = 1$ return $x = b/A$;
2. create a vector x of length n ;
3. set $x_n = b_n/A_{nn}$;
4. set $b' = b_{1:(n-1)} - A_{[1:(n-1),n]}x_n$;
5. set $A' = A_{[1:(n-1),1:(n-1)]}$;
6. solve $A'x' = b'$ for x' by back-substitution ;
7. set $x_{[1:(n-1)]} = x'$;
8. return x .

- (a) Implement this algorithm as a recursive function in R. Your function should take as input an upper triangular $n \times n$ matrix A and return a solution x satisfying $Ax = b$.
- (b) For $n = 10$, create an $n \times n$ upper triangular matrix A and a vector b of length n . Check the solution from your function against `backsolve()` and `solve()`.

4. Longest Increasing Subsequence.*

The object of this exercise is to write a function that, given a sequence of numbers $\mathbf{a} = (a_1, \dots, a_k)$, returns $Q(\mathbf{a}) = (a_{s_1}, \dots, a_{s_L})$, the longest subsequence of \mathbf{a} such that $a_{s_1} < \dots < a_{s_L}$. [Note that it is implicit in the idea of a subsequence that $s_1 < \dots < s_k$.]

- (a) Write a function that, for each i , recursively calculates the longest increasing subsequence of $(a_1, \dots, a_{i-1}, a_i)$ that ends with a_i . [Hint: remove the final element of \mathbf{a} and invoke the function on this shorter vector; then add a_k to the longest subsequence whose final element is less than a_k .]
- (b) Use this to return a function that solves the problem of finding $Q(\mathbf{a})$.
- (c) Calculate the computational complexity of this method.