Statistical Programming

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.

```
randPDmat <- function(n) {
    C <- matrix(rnorm(n^2), n, n)
    out <- C %*% t(C)
    out
}
A <- randPDmat(8)
all(eigen(A)$value > 0) # are all eigenvalues positive?
[1] TRUE
```

(b) Implement the recursive Cholesky decomposition algorithm from the lecture.

```
# Function to find Cholesky decomposition of symmetric
myChol <- function(A) {</pre>
    n <- dim(A)[1]
    if (dim(A)[2] != n)
        stop("A must be a square matrix") # check n x n
    if (n == 1)
        return(sqrt(A))
    L <- matrix(0, n, n)
    L[1, 1] <- sqrt(A[1, 1]) #count as 1 op
    L[2:n, 1] <- A[2:n, 1]/L[1, 1] #n-1 ops
    L[1, 2:n] <- rep(0, n - 1)
    A22 = A[2:n, 2:n, drop = FALSE]
    newA = A22 - L[2:n, 1] %*% t(L[2:n, 1]) #2(n-1)^2 here
    # but n(n-1) possible
    L[2:n, 2:n] = myChol(newA)
    return(L)
}
```

(c) Test it using your function for generating positive definite matrices, and by comparing the answers to chol().

```
A <- randPDmat(6)
## chol() gives an upper triangular matrix,
## but can compare transpose to our answer:
t(chol(A)) - myChol(A) # all numerically 0</pre>
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]
[1,]	0	0	0	0.00e+00	0	0
[2,]	0	0	0	0.00e+00	0	0
[3,]	0	0	0	0.00e+00	0	0
[4,]	0	0	0	2.22e-16	0	0
[5,]	0	0	0	-1.11e-16	0	0
[6,]	0	0	0	-2.78e-17	0	0

(d) Create a function which takes a vector **mu** and a symmetric positive definite matrix Sigma and uses them to generate a multivariate normal vector $N_n(\mu, \Sigma)$. Your function should check that Sigma is positive definite using **eigen()** and symmetric using **isSymmetric()**.

```
mvnorm <- function(mu, Sigma) {
    n <- nrow(Sigma)
    ## check matrix is valid
    if (min(eigen(Sigma)$values) <= 0)
        stop("Sigma must be positive definite")
    if (!isSymmetric(Sigma))
        stop("Sigma must be symmetric")

    ## use method from lecture
    L <- myChol(Sigma)
    z <- rnorm(n)
    x <- mu + L %*% z
    return(c(x))
}</pre>
```

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 ≤ 1 return x;
2. pick an arbitrary 'pivot' element i ≤ n;
3. let z = (x_j | x_j < x_i) and y = (x_j | 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.

```
quickSort <- function(x) {
    n <- length(x)
    if (n <= 1)
        return(x)
    i <- sample(n, 1)  # pick a pivot at random
    z <- x[x < x[i]]</pre>
```

```
y <- x[x > x[i]]
xis <- x[x == x[i]] # in case of ties
return(c(Recall(z), xis, Recall(y)))
}
x <- rnorm(10000)
out <- quickSort(x)</pre>
```

- (b) What is the complexity if x_i is always the smallest element? In this case we see g(n) = 2n + g(n-1) + g(0), so $g(n) = O(n^2)$. The algorithm relies on being able to divide the problem up to be efficient, so picking the smallest element doesn't work very well.
- (c) Show that, if the pivot x_i is the median element on each call, that the complexity is at most $O(n \log_2(n))$.

In this case we get a recursion of the form g(n) = 2n + 2g((n-1)/2). Now suppose that $g(k) \leq Mk \log_2 k$ for some M and all k < n. Then

$$g(n) \le 2n + 2M\frac{n}{2}\log_2\frac{n}{2}$$
$$= 2n + Mn\log_2n - Mn$$
$$\le Mn\log_2n$$

provided that $M \ge 2$. Hence $g(n) = O(n \log_2 n)$.

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.

```
backSolve <- function(A, b) {
    n <- length(b)
    if (nrow(A) != ncol(A))
        stop("A must be a square matrix")
    if (nrow(A) != n)
        stop("Dimensions of A and b must match")
    x = b[n]/A[n, n]
    if (n == 1)
        return(x)
    A2 <- A[-n, -n, drop = FALSE]
    b2 <- b[-n] - A[-n, n] * x
    x = c(backSolve(A2, b2), x)
    return(x)
}</pre>
```

(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().

```
> A <- matrix(0, 10, 10)
> A[upper.tri(A, diag = TRUE)] = rnorm(55)
> b <- rnorm(10)
> backSolve(A, b)
 [1]
      977.49
              95.75 -526.03 -44.26
                                      17.96
                                              35.75
                                                       4.99
                                                               1.40
 [9]
       0.67
               1.36
> solve(A, b)
 [1] 977.49
              95.75 -526.03 -44.26 17.96
                                              35.75
                                                       4.99
                                                               1.40
 [9] 0.67 1.36
```

4. Longest Increasing Subsequence.*

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

(a) Write a function that, for each *i*, recursively calculates the longest increasing subsequence of $(a_1, \ldots, a_{i-1}, a_i)$ that ends with a_i . [Hint: remove the final element of **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 .]

```
## Return longest increasing subsequences for first i entries input: x - a
## numeric vector output: a list of the same length as x, whose ith entry is
## the longest increasing subsequence of x[1], \ldots, x[i] that ends with x[i].
liseqs <- function(x) {</pre>
    ## check length of x and finish if <= 1.</pre>
    n <- length(x)</pre>
    if (n == 0)
        return(list()) else if (n == 1)
        return(list(x))
    ## remove last element and recall function
    x_s <- x[-n]
    tmp <- Recall(x_s)</pre>
    ## if last element is smallest, longest sequence is just that value
    if (\min(x) == x[n])
        return(c(tmp, list(x[n])))
    ## now get lengths of these subsequences
    len <- lengths(tmp, FALSE)</pre>
    ## attach x[n] to longest subsequence whose final element is smaller
    wh <- which.max(len * (x_s <= x[n])) # longest we can add x[n] to
    out <- c(tmp, list(c(tmp[[wh]], x[n])))</pre>
    out
}
liseqs(rnorm(10))
[[1]]
[1] 0.553
[[2]]
[1] 0.553 0.559
[[3]]
[1] -0.124
[[4]]
```

[1] -1.15

```
[[5]]
[1] -0.124 0.384
[[6]]
[1] -1.15 -1.06
[[7]]
[1] 0.553 0.559 0.757
[[8]]
[1] 0.553 0.559 0.757 1.817
[[9]]
[1] 0.553 0.559 0.581
[[10]]
[1] -1.153 -1.060 -0.161
```

(b) Use this to return a function that solves the problem of finding $Q(\mathbf{a})$. This is now rather trivial.

```
longIncSub <- function(x) {
    ## invoke the earlier function, and return the longest subsequence
    tmp <- liseqs(x)
    wh <- which.max(lengths(tmp))
    return(tmp[[wh]])
}
longIncSub(rnorm(10))
[1] -0.0537 0.1745 0.5944 0.8791</pre>
```

- (c) Calculate the computational complexity of this method.
 It is not hard to see that the code above just has to search through the list of vectors ending with a₁,..., a_{k-1} to find the longest, so this is an operation that is just linear in k. Since this is recursed we have the relation f(k) = O(k) + f(k − 1), and it is easy to check that this implies that f(k) = O(k²).