

Part A Simulation and Statistical Programming

Practical 5

Working with Matrices, Solving Equations, Regression

Q1. Using the `matrix()`, `seq()` and `rep()` functions to construct the following 4x4 Hankel matrix

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \\ 4 & 5 & 6 & 7 \end{bmatrix}$$

```
A = matrix(rep(0:3, rep(4, 4))+rep(1:4, 4), 4, 4);  
A = cbind(1:4, 2:5, 3:6, 4:7)
```

Use this matrix to create the following matrix

$$B = \begin{bmatrix} 1 & 1/2 & 1/3 & 1/4 \\ 1/2 & 1/3 & 1/4 & 1/5 \\ 1/3 & 1/4 & 1/5 & 1/6 \\ 1/4 & 1/5 & 1/6 & 1/7 \end{bmatrix}$$

```
B = 1/A
```

Q2. (Refer to the reference slides) Calculate the matrix

$$H = C(C^T C)^{-1} C^T,$$

where C is the 4x3 matrix composed of the first three columns of B . Create a diagonal matrix which has the diagonal elements of H as its diagonal entries. Calculate the eigenvalues and eigenvectors of H . Compare the trace of H to the sum of its eigenvalues. Compare the determinant of H with the product of the eigenvalues.

```
> A=cbind(1:4,2:5,3:6,4:7)  
> B=1/A  
> C=B[,1:3,drop=F]  
> H = C%%solve(t(C)%*%C)%*%t(C) #crossprod(C,C) is better than t(C)%*%C  
> round(H,2)  
      [,1] [,2] [,3] [,4]  
[1,]  1.00  0.01 -0.02  0.01  
[2,]  0.01  0.90  0.25 -0.17  
[3,] -0.02  0.25  0.38  0.42  
[4,]  0.01 -0.17  0.42  0.72  
> (e = eigen(H))
```

```

$values
[1] 1.000000e+00 1.000000e+00 1.000000e+00 -4.320137e-16

$eigenvectors
      [,1]      [,2]      [,3]      [,4]
[1,] -0.4448746 0.24242393 -0.6543369 -0.02630668
[2,] -0.8051808 -0.80779931 0.3194446 0.31568019
[3,] -0.3776476 0.02656129 0.4774422 -0.78920046
[4,] -0.1056067 0.53664272 0.4917797 0.52613364

> sum(diag(H))
[1] 3
> sum(e$values)
[1] 3
> det(H)
[1] -4.089843e-16
> prod(e$values)
[1] -4.320137e-16

```

Q3. Let $[x_1, x_2, x_3, x_4, x_5, x_6]^T = [10, 11, 12, 13, 14, 15]^T$. Find the coefficients of the quintic polynomial $f(x)$ for which $[f(x_1), f(x_2), f(x_3), f(x_4), f(x_5), f(x_6)]^T = [25, 16, 26, 19, 21, 20]^T$.

```

> y = c(25, 16, 26, 19, 21, 20)
> x = matrix(c(10:15)^rep(0:5, rep(6,6))), nrow = 6)
> solve(x,y)
[1] 2.536100e+05 -1.025510e+05 1.650092e+04 -1.320667e+03 5.258333e+01
[6] -8.333333e-01

```

Q4. The data set `speed.txt` give the times in seconds recorded by the winners in the finals of the men's sprint events (100, 200, 400, 800 and 1500 metres) at each of the 21 Olympic Games from 1900 to 2004 along with the heights above sea level of the different venues.

Fit the following models to the dataset

(i) speed against $\log(\text{distance})$

(ii) speed against distance, distance^2 , year, year^2 and $\log(\text{altitude})$.

Fit model (i) twice (A) using the `lm()` and `summary()` commands, and (B) using `qr.solve()`.

For each model produce normal qq-plots and comment on these plots.

```

> s = read.table("speed.txt", header = T)
> (n<-dim(s)[1])
[1] 92
> #either (just the intercept and slope)
> X<-cbind(rep(1,n),log(s$Distance.100))
> (theta<-qr.solve(X,s$Distance.100/s$Time))
[1] 0.10719340 -0.01526622
> #or (the full model fit)
> l1 = lm(Distance.100/Time~log(Distance.100),data=s)
> par(mfrow = c(1,2))
> qqnorm(residuals(l1),main=''); qqline(residuals(l1))
>
> #poor fit, try more elaborate model
> s$Distance.sqr = s$Distance.100^2
> s$Year.sqr = s$Year^2
> #(full model fit)
> l2=lm(Distance.100/Time~Distance.100+Distance.sqr+Year+Year.sqr+log(Altitude), data=s)

```

```

> summary(l2)
...
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
(Intercept) -1.835e+00  5.002e-01  -3.669 0.000421 ***
Distance.100 -5.893e-03  1.090e-04 -54.058 < 2e-16 ***
Distance.sqr  2.120e-04  6.141e-06  34.529 < 2e-16 ***
Year          1.887e-03  5.131e-04   3.679 0.000408 ***
Year.sqr     -4.573e-07  1.316e-07  -3.476 0.000800 ***
log(Altitude) 2.709e-04  5.037e-05   5.378 6.39e-07 ***
...
> par(mfrow = c(1,2))
> qqnorm(residuals(l2),main=''); qqline(residuals(l2))

Better.

```

Q5. Here is an algorithm to solve $Ax=b$ for A 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 $n \times 1$ b
Output: $n \times 1$ vector x solving $Ax=b$

Step 1: set n equal the number of rows in A
Step 2: If n equals 1 then we are done; return $x=b/A$.
Step 3: Otherwise (if $n>1$)

Step 3.1: create an $n \times 1$ vector x
Step 3.2: set the last entry of x to be $x[n]=b[n]/A[n,n]$
Step 3.3: set $b' = b[1:(n-1)] - A[1:(n-1),n]*x[n]$.
Step 3.4: set $A' = A[1:(n-1),1:(n-1)]$

Step 3.5: solve $A'x' = b'$ for x' by back-substitution
Step 3.6: set $x[1:(n-1)] = x'$

Step 3.7: 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 $Ax=b$.

```

#back substitution for UPPER triangular
my.back<-function(A,b) {
  n=dim(A)[1] #assume nxn
  if (n==1) return(b/A)
  i=1:(n-1)

  x=matrix(0,n,1)
  x[n]=b[n]/A[n,n]
  b2=b[i]-A[i,n]*x[n]
  A22=A[i,i,drop=FALSE]

  x[i]=my.back(A22,b2)

  return(x)
}

```

(b). Create an $n \times n$ upper triangular matrix A and an $n \times 1$ vector b and check your solution against `backsolve()` and `solve()`.

```
> A=matrix(runif(9),3,3); A[lower.tri(A)]<-0; round(A,2)
      [,1] [,2] [,3]
[1,]  0.1 0.74 0.99
[2,]  0.0 0.26 0.41
[3,]  0.0 0.00 0.97
> b=matrix(runif(9),3,1); round(b,2)
      [,1]
[1,] 0.16
[2,] 0.21
[3,] 0.48
> my.back(A,b)
      [,1]
[1,] -3.45932839
[2,]  0.02887721
[3,]  0.49810310
> backsolve(A,b)
      [,1]
[1,] -3.45932839
[2,]  0.02887721
[3,]  0.49810310
> solve(A,b)
      [,1]
[1,] -3.45932839
[2,]  0.02887721
[3,]  0.49810310
```

Q6. Solving $Ax=b$ using `solve(A)%*%b` is about 4 times slower than `solve(A,b)`. We saw `solve(A,b)` was about $(2/3)n^3$ multiplications. Show that `solve(A)` is about $(8/3)n^3$ (Hint: if $AB=I$ then the i th column of B (B_i say) is the solution of $AB_i=I_i$ where I_i is the i th column of I ; we need the LU factorization of A and then solve $AB_i=I_i$ for $i=1,2,\dots,n$).

The initial LU is $2n^3/3$ and then we solve $AB_i=I_i$ n times using elimination, at $2n^2$ multiplications per solve. Total cost is $2n^3/3+2n^3=8n^3/3$.