## ST758, Homework 2

## Due Tuesday, Sep 23, 2014

- 1. Show the following facts about triangular matrices. A unit triangular matrix is a triangular matrix with all diagonal entries being 1.
  - (a) The product of two upper (lower) triangular matrices is upper (lower) triangular.
  - (b) The inverse of an upper (lower) triangular matrix is upper (lower) triangular.
  - (c) The product of two unit upper (lower) triangular matrices is unit upper (lower) triangular.
  - (d) The inverse of a unit upper (lower) triangular matrix is unit upper (lower) triangular.
  - (e) An orthogonal upper (lower) triangular matrix is diagonal.
- 2. Suppose matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  has entries  $a_{ij} = i(n j + 1)$  and matrix  $\mathbf{B} \in \mathbb{R}^{n \times n}$  has entries  $b_{ij} = \sum_{k=1}^{i} \sigma_k^2$  for  $j \geq i$  and  $\sigma_k^2 \geq 0$ . Show that  $\mathbf{A}$  and  $\mathbf{B}$  are positive semidefinite.
- 3. (a) Show the Sherman-Morrison formula

$$(A + uu^{\mathsf{T}})^{-1} = A^{-1} - \frac{1}{1 + u^{\mathsf{T}}A^{-1}u}A^{-1}uu^{\mathsf{T}}A^{-1},$$

where  $A \in \mathbb{R}^{n \times n}$  is a nonsingular matrix and  $u \in \mathbb{R}^n$ . This formula supplies the inverse of the symmetric, rank-one perturbation of A.

(b) Show the Woodbury formula

$$(A + UV^{\mathsf{T}})^{-1} = A^{-1} - A^{-1}U(I_m + V^{\mathsf{T}}A^{-1}U)^{-1}V^{\mathsf{T}}A^{-1}$$

where  $A \in \mathbb{R}^{n \times n}$  is nonsingular,  $U, V \in \mathbb{R}^{n \times m}$ , and  $I_m$  is the  $m \times m$  identity matrix. In many applications m is much smaller than n. Woodbury formula generalizes Sherman-Morrison and is valuable because the smaller matrix  $I_m + V^{\mathsf{T}} A^{-1} U$  is typically much easier to invert than the larger matrix  $A + UV^{\mathsf{T}}$ .

(c) Show the binomial inversion formula

$$(A + UBV^{\mathsf{T}})^{-1} = A^{-1} - A^{-1}U(B^{-1} + V^{\mathsf{T}}A^{-1}U)^{-1}V^{\mathsf{T}}A^{-1},$$

where  $\boldsymbol{A}$  and  $\boldsymbol{B}$  are nonsingular.

(d) Show the identity

$$\det(\boldsymbol{A} + \boldsymbol{U}\boldsymbol{V}^{\mathsf{T}}) = \det(\boldsymbol{A})\det(\boldsymbol{I}_m + \boldsymbol{V}^{\mathsf{T}}\boldsymbol{A}^{-1}\boldsymbol{U}).$$

This formula is useful for evaluating the density of a multivariate normal with covariance matrix  $A + UU^{\mathsf{T}}$ .

(e) Consider the  $n \times n$  matrix

$$\boldsymbol{M} = \begin{pmatrix} a & b & \cdots & b \\ b & a & \cdots & b \\ \vdots & \vdots & \vdots & \vdots \\ b & b & \cdots & a \end{pmatrix}.$$

Show that M has inverse and determinant

$$\boldsymbol{M}^{-1} = \frac{1}{a-b} \left[ \boldsymbol{I}_n - \frac{b}{a+(n-1)b} \boldsymbol{1}_n \boldsymbol{1}_n^T \right]$$
$$\det(\boldsymbol{M}) = (a-b)^{n-1} [a+(n-1)b].$$

- 4. Write an R function, with interface solve.power(A, k, b), to solve linear equation  $A^k x = b$ , where  $A \in \mathbb{R}^{n \times n}$  is non-singular and  $b \in \mathbb{R}^n$ . Test your function in simulated data.
- 5. Consider a mixed effects model

$$y_i = \boldsymbol{x}_i^t \boldsymbol{\beta} + \boldsymbol{z}_i^t \boldsymbol{\gamma} + \epsilon_i, \quad i = 1, \dots, n,$$

where  $\epsilon_i$  are independent normal errors  $N(0, \sigma_0^2)$ ,  $\boldsymbol{\beta} \in \mathbb{R}^p$  are fixed effects, and  $\boldsymbol{\gamma} \in \mathbb{R}^q$  are random effects assumed to be  $N(\mathbf{0}_q, \sigma_1^2 \boldsymbol{I}_q)$  independent of  $\epsilon_i$ . Show that  $\boldsymbol{y} \sim N(\boldsymbol{X}\boldsymbol{\beta}, \sigma_0^2 \boldsymbol{I}_n + \sigma_1^2 \boldsymbol{Z} \boldsymbol{Z}^t)$ , where  $\boldsymbol{y} \in \mathbb{R}^n$ ,  $\boldsymbol{X} \in \mathbb{R}^{n \times p}$ , and  $\boldsymbol{Z} \in \mathbb{R}^{n \times q}$ . Write an R function, with interface dmvnorm.lowrank(y, mu, Z, sigma0, sigma1, log = FALSE), that evaluates the (log)-density of a multivariate normal with mean  $\boldsymbol{\mu}$  and covariance  $\sigma_0^2 \boldsymbol{I} + \sigma_1^2 \boldsymbol{Z} \boldsymbol{Z}^t$  at  $\boldsymbol{y}$ . Test your function on simulated data.

6. In class we learnt about the BLAS and how it has become a de facto standard for basic linear algebra operations. R uses the BLAS and LAPACK libraries extensively to accelerate certain types of operations. Apart from \*, %\*%, eigen, and qr, what other common functions use the BLAS and LAPACK routines to speed up calculations?

Find out if R uses an optimized BLAS on your system. If yes, what version is used? If not, find out if any optimized BLAS is available for your system.