Answers to 2009 STA 437/1005 mid-term test

1a)
$$\bar{x} = [100 \ 200]'$$
.

1b)

$x_{j1} -$	\bar{x}_1	$(x_{j1} - \bar{x}_1)^2$	$x_{j2} - \bar{x}_2$	$](x_{j2}-\bar{x}_2)^2$	$(x_{j1} - \bar{x}_{j1})(x_{j2} - \bar{x}_2)$
-5		25	10	100	-50
0		0	0	0	0
5		25	-10	100	-50
-5		25	-5	25	25
5		25	5	25	25

$$s_{11} = (1/4) \sum_{j} (x_{j1} - \bar{x}_{1})^{2} = 100/4$$

$$s_{22} = (1/4) \sum_{j} (x_{j2} - \bar{x}_{j})^{2} = 250/4$$

$$s_{12} = s_{21} = (1/4) \sum_{j} (x_{j1} - \bar{x}_{j1}) (x_{j2} - \bar{x}_{2}) = -50/4$$

$$S = \begin{bmatrix} 100/4 & -50/4 \\ -50/4 & 250/4 \end{bmatrix} = (50/4) \begin{bmatrix} 2 & -1 \\ -1 & 5 \end{bmatrix}$$

$$1c) S^{-1} = (4/50) \begin{bmatrix} 5/9 & 1/9 \\ 1/9 & 2/9 \end{bmatrix} = (4/450) \begin{bmatrix} 5 & 1 \\ 1 & 2 \end{bmatrix}$$

$$d_{1}^{2} = (x_{1} - \bar{x})'S^{-1}(x_{1} - \bar{x}) = [-5\ 10](4/450) \begin{bmatrix} 5 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} -5 \\ 10 \end{bmatrix} = 2$$

2a) By the spectral decomposition theorem, we can write any symmetric real matrix, A, as $A = \lambda_1 e_1 e'_1 + \cdots + \lambda_k e_k e'_k$. Since the e_i are orthogonal (hence linearly independent), we can write any non-zero vector, c, as $c = c_1 e_1 + \cdots + c_k e_k$, for some scalars c_i , not all of which are zero. We can now write c'Ac as follows:

$$c'Ac = (c_1e_1 + \dots + c_ke_k)' (\lambda_1e_1e'_1 + \dots + \lambda_ke_ke'_k) (c_1e_1 + \dots + c_ke_k)$$

Multiplying this out, and noting that $e'_i e_i = 1$, and also that $e'_i e_j = 0$ when $i \neq j$, we see that

$$c'Ac = c_1^2\lambda_1 + \dots + c_k^2\lambda_k$$

If all the λ_k are positive, this must be greater than zero, since at least one of the c_i is non-zero. The matrix A is therefore positive definite.

2b) First, note that the converse of part (a) is true: if a matrix A is positive definite, all its eigenvalues are positive. To see this, suppose e is an eigenvalue of A (which we can take to be of length one), with eigenvalue λ . Then e'Ae > 0 by the definition of positive definiteness. But $e'Ae = e'\lambda e = \lambda e'e = \lambda$, so $\lambda > 0$.

Next, write A and B as follows, using the spectral decomposition theorem:

$$A = \alpha_1 e_1 e'_1 + \dots + \alpha_k e_k e'_k$$
$$B = \beta_1 e_1 e'_1 + \dots + \beta_k e_k e'_k$$

where e_1, \ldots, e_k are the eigenvectors of both A and B, $\alpha_1, \ldots, \alpha_k$ are the eigenvalues of A, and β_1, \ldots, β_k are the eigenvalues of B.

When we now multiply A and B, and note that $e'_i e_i = 1$, and also that $e'_i e_j = 0$ when $i \neq j$, we get

$$AB = \alpha_1 \beta_1 e_1 e'_1 + \dots + \alpha_k \beta_k e_k e'_k$$

Since $(e_i e'_i)' = e''_i e'_i = e_i e'_i$, each term above is symmetric, so AB is also symmetric. We can easily see that the eigenvectors of AB are the same e_i , and that the eigenvalues are the $\alpha_i \beta_i$ — for instance

$$ABe_1 = \alpha_1\beta_1e_1e'_1e_1 + \alpha_2\beta_2e_2e'_2e_1 + \cdots = \alpha_1\beta_1e_1$$

If A and B are positive definite, then all the α_i and β_i are positive, and therefore all the eigenvalues of AB are positive. By part (a), AB is therefore positive definite.

- 3a) A
- 3b) B
- 3c) B

3d) C

The T^2 statistic for testing H_0 : $\mu = 0$ versus H_1 : $\mu \neq 0$ is

$$T^2 = n\bar{x}'S_X^{-1}\bar{x}$$

Let A be any non-singular $p \times p$ matrix. Transforming to y = Ax, we will find that the sample mean is $\bar{y} = A\bar{x}$ and the sample covariance matrix is $S_Y = AS_X A'$. The new T^2 statistic will then be

$$T^{2} = n\bar{y}'S_{Y}^{-1}\bar{y}$$

$$= n(A\bar{x})'(AS_{X}A')^{-1}(A\bar{x})$$

$$= n\bar{x}'A'A'^{-1}S_{X}^{-1}A^{-1}A\bar{x}$$

$$= n\bar{x}'S_{X}^{-1}\bar{x}$$

which is the same as the old T^2 statistic.

- 4a) D, no outliers, not normal (heavy-tailed).
- 4b) E, no outliers, not normal (light-tailed).
- 4c) A, no outliers, no reason to think it's not normal (note the small sample size).
- 4d) B, one outlier (rightmost point, about 7), no reason to think it's not normal (after ignoring the outlier).
- 5a) $\Sigma = \begin{bmatrix} 5 & 1 & 1 & 1 \\ 1 & 6 & 2 & 2 \\ 1 & 2 & 7 & 3 \\ 1 & 2 & 3 & 8 \end{bmatrix}$

5b)

$$\operatorname{Var}(\overline{Y}) = [1/4 \ 1/4 \ 1/4 \ 1/4] \Sigma \begin{bmatrix} 1/4 \\ 1/4 \\ 1/4 \\ 1/4 \end{bmatrix} = 23/8$$

5c) Normal, with mean $0 + 1(8)^{-1}(y_4 - 0) = y_4/8$ and variance $5 - 1(8)^{-1}1 = 39/8$.