# HW11, Math 307. CSUF. Spring 2007.

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# 1 Section 5.6, problem 1

problem: If B is similar to A and C is similar to B, show that C is similar to A. What matrices are similar to *I*?

answer:

Since B is similar to A and C is similar to B, then we have the following

$$S_1^{-1}CS_1 = B (1)$$

$$S_2^{-1}BS_2 = A \tag{2}$$

From (1) and (2)

$$S_2^{-1}BS_2 = A$$
  

$$S_2^{-1} (S_1^{-1}CS_1) S_2 = A$$
  

$$(S_2^{-1}S_1^{-1}) C (S_1S_2) = A$$
  

$$(S_1S_2)^{-1} C (S_1S_2) = A$$

Let  $S_1S_2 = S_3$ , hence the above becomes

$$S_3^{-1}CS_3 = A$$

Hence C is similar to A. Now for the second part. We write

$$S^{-1}AS = I$$
$$S^{-1}A = S$$

So A must be I, hence only I is similar to I.

#### Section 5.6 problem 2 2

problem: Describe in words all the matrices that are similar to  $\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$ , and find 2 of them answer:

Let *A* be the above matrix. The above matrix represents a reflection across the x-axis. Hence Reflection across the y axis will be similar to it. Any multiple of this reflection matrix will also be similar to A.

Since reflection across the y-axis is  $B = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$  then this *B* matrix is similar to *A*. Then any multiple of *B* is also similar to *A*, such as  $\begin{pmatrix} -10 & 0 \\ 0 & 10 \end{pmatrix}$  and  $\begin{pmatrix} -20 & 0 \\ 0 & 20 \end{pmatrix}$ 

# 3 Section 5.6, problem 5

Problem: show (if *B* is invertible) then *BA* is similar to *AB* answer: we want to show that  $M^{-1}(BA)M = AB$ Let  $M^{-1}(BA)M = H$ , i.e. let  $BA^{\sim}H$ , and try to show that H = AB

$$M^{-1} (BA) M = H$$
  

$$(BA) M = MH$$
  

$$BA = MHM^{-1}$$
  

$$A = B^{-1}MHM^{-1}$$
  

$$AB = B^{-1}MHM^{-1}B$$
  

$$AB = (B^{-1}M) H (M^{-1}B)$$
  

$$AB = (M^{-1}B)^{-1} H (M^{-1}B)$$

Let  $M^{-1}B = Z$ , hence the above becomes

$$AB = Z^{-1}HZ$$

Then *H*<sup>~</sup>*AB* 

But we started by stating that  $H^{\tilde{}}BA$ , and since if  $r_1 r_2$  and  $r_2 r_3$  then  $r_1 r_3$  then we showed  $BA^{\tilde{}}AB$ .

# 4 Section 5.6 problem 18

problem: find normal matrix  $(NN^H = N^H N)$  that is not Hermitian, skew symmetric, unitary, or diagonal. Show that all permutation matrices are normal answer:

#### 5 Section 6.1, problem 1

problem: quadratic  $f = x^2 + 4xy + 2y^2$  has saddle point at origin, despite that its coefficients are positive. Write f as difference of 2 squares answer: Let  $f = (ax + by)^2 - (cx + dy)^2$ , hence

$$f = (ax + by)^{2} - (cx + dy)^{2}$$
  
=  $a^{2}x^{2} + b^{2}y^{2} + 2abxy - (c^{2}x^{2} + d^{2}y^{2} + 2cdxy)$   
=  $a^{2}x^{2} + b^{2}y^{2} + 2abxy - c^{2}x^{2} - d^{2}y^{2} - 2cdxy$   
=  $x^{2} (a^{2} - c^{2}) + y^{2} (b^{2} - d^{2}) + xy (2ab - 2cd)$ 

Hence, compare coefficients, we have  $a^2 - c^2 = 1$ ,  $b^2 - d^2 = 2$ , 2ab - 2cd = 4so ab - cd = 2.

Let c = 1, then we have

 $a^2 = 2, b^2 - d^2 = 2, 2ab - 2d = 4$ 

3 equations in 3 unknown. Solve with computer for speed (running out of time!) I get one of the solutions as *Г*-5

$$d = 0, a = -\sqrt{2}, b = -\sqrt{2}$$
  
So  $f = (ax + by)^2 - (cx + dy)^2 = \left(-\sqrt{2}x + -\sqrt{2}y\right)^2 - (x)^2$ 

# 6 Section 6.1, problem 8

problem: decide for or against PD for these matrices, write out corresponding  $f = x^T A x$ Answer: I use a > 0, and  $ac > b^2$  test where  $A = \begin{pmatrix} a & b \\ b & c \end{pmatrix}$ 

$$\begin{pmatrix} 1 & 3 \\ 3 & 5 \end{pmatrix} \rightarrow 1 > 0, 5 > 9 \text{ no}, \quad \text{Not PD} \rightarrow f = ax^2 = 2bxy + cy^2 \rightarrow f = x^2 + 6xy + 3y$$

$$\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \rightarrow a > 0, 1 > 1, \text{no}, \quad \text{Not PD} \rightarrow f = ax^2 = 2bxy + cy^2 \rightarrow f = x^2 - 2xy + y$$

$$\begin{pmatrix} 2 & 3 \\ 3 & 5 \end{pmatrix} \rightarrow a > 0, 10 > 9, \text{yes}, \quad \text{PD} \rightarrow f = ax^2 = 2bxy + cy^2 \rightarrow f = 2x^2 + 6xy + 5y$$

$$\begin{pmatrix} -1 & 2 \\ 2 & -8 \end{pmatrix} \rightarrow -1 > 0, \text{ no } \quad \text{Not PD} \rightarrow f = ax^2 + 2bxy + cy^2 \rightarrow f = -x^2 + 4xy - 8y$$
For (b) we have  $f = x^2 - 2xy + y$ , if  $y = \frac{x^2}{2x-1}$  then  $f = x^2 - 2x\frac{x^2}{2x-1} + \frac{x^2}{2x-1} = 0$ , hence I plot this:

And along the lines shown is f = 0

# 7 Section 6.1, problem 3

problem: if A is 2x2 symmetric matrix, passes test that a>0,  $ac > b^2$  solve equation det  $(A - \lambda I) = 0$  and show that eigenvalues are >0

answer:

Matrix is PD, then

$$det\left(\begin{pmatrix}a&b\\b&c\end{pmatrix}-\lambda\begin{pmatrix}1&0\\0&1\end{pmatrix}\right)=0$$
$$\begin{vmatrix}a-\lambda&b\\b&c-\lambda\end{vmatrix}=0$$
$$(a-\lambda)(c-\lambda)-b^{2}=0$$
$$ac-a\lambda-c\lambda+\lambda^{2}=0$$
$$\lambda^{2}+\lambda(-a-c)+ac=0$$

Hence  $\lambda_1 = a, \lambda_2 = c$ But a > 0, so  $\lambda_1 > 0$ , and given ac > positive quantity  $b^2$ , then  $\lambda_2 = c \rightarrow \lambda_2 > 0$ 

# 8 Section 6.1 problem 5

(a) For which numbers 
$$b$$
 is  $\begin{pmatrix} 1 & b \\ b & 9 \end{pmatrix}$  PD?  
 $\begin{pmatrix} a & b \\ b & c \end{pmatrix}$  is PD is  $a > 0$  and  $ac > b^2$   
for PD need  $ac > b^2$ , hence need  $9 > b^2$  is.  $b < 3$  and  $b > -3$ , so  $\boxed{-3 < b < 3}$   
(b)Factor  $A = LDL^T$  when  $b$  is in the range above  
 $\begin{pmatrix} 1 & b \\ b & 9 \end{pmatrix} \rightarrow l_{21} = b \rightarrow U = \begin{pmatrix} 1 & b \\ 0 & 9 - b^2 \end{pmatrix}$   
So  $L = \begin{pmatrix} 1 & 0 \\ b & 1 \end{pmatrix}, D = \begin{pmatrix} 1 & 0 \\ 0 & 9 - b^2 \end{pmatrix}, L^T = \begin{pmatrix} 1 & b \\ 0 & 1 \end{pmatrix}$   
(c) What is the minimum of  $f(x, y) = \frac{1}{2}(x^2 + 2bxy + 9y^2) - y$  when in this range when  $f(x, y) = \frac{1}{2}(x^2 + 2bxy + 9y^2) - y = \frac{1}{2}x^2 + bxy + \frac{9}{2}y^2 - y$   
 $\frac{\partial f}{\partial x} = x + by = 0, \frac{\partial f}{\partial y} = bx + 9y - 1 = 0$   
Hence  $\begin{pmatrix} 1 & b \\ b & 9 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & b \\ 0 & 9 - b^2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$   
Hence  $y = \frac{1}{9-b^2}$  and  $x + by = 0 \rightarrow \boxed{x = -\frac{b}{9-b^2}}$   
So  $f(x, y) = \frac{1}{2}(x^2 + 2bxy + 9y^2) - y$   
Hence  $f(x, y) \rightarrow \frac{1}{2}((-\frac{b}{9-b^2})^2 + 2b(-\frac{b}{9-b^2})(\frac{1}{9-b^2}) + 9(\frac{1}{9-b^2})^2) - (\frac{1}{9-b^2}) = \frac{1}{2(b^2-9)}$   
So minimum is  $\boxed{\frac{1}{2(b^2-9)}}$   
(d)When  $b = 3$ , we see that we get  $\frac{1}{0} = \infty$  so no minimum

#### 9 Section 6.1 problem 17

Problem: If A has independent columns then  $A^T A$  is square and symmetric and invertible. Rewrite  $\vec{x}^T A^T A \vec{x}$  to show why it is positive except when  $\vec{x} = 0$ , then  $A^T A$  is PD answer:  $\vec{x}^T (A^T A) \vec{x} = (A \vec{x})^T A \vec{x}$ 

Let  $A\vec{x} = \vec{b}$ , then the above is  $\vec{b}^T \vec{b} = \|\vec{b}\|^2$ , which is positive quantity except when  $\vec{b} = \vec{0}$ , which occurs when  $A\vec{x} = \vec{b} = \vec{0}$  which happens only when  $\vec{x} = \vec{0}$ , since *A* is invertible. Hence  $A^T A$  is positive definite except when  $\vec{x} = 0$ 

#### 10 Section 6.2, problem 7

**problem:** If  $A = QAQ^T$  is P.D. then  $R = Q\sqrt{A}Q^T$  is its S.P.D. square root. Why does R have positive eigenvalues? Compute R and verify  $R^2 = A$  for  $A = \begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix}$ ,  $A = \begin{pmatrix} 10 & -6 \\ -6 & 10 \end{pmatrix}$ 

**answer:** For  $A = \begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix}$ 

Given *R* is P.D. (problem said so), Hence  $\vec{x}^T R \vec{x} > 0$  for all  $\vec{x} \neq 0$ Now (assuming in all that follows that  $x \neq 0$ )

$$R\vec{x} = \lambda \vec{x}$$
$$\vec{x}^T R\vec{x} = x^T \lambda \vec{x}$$
$$\vec{x}^T R\vec{x} = \lambda \|\vec{x}\|^2$$

Since  $\vec{x}^T R \vec{x} > 0$  then  $\lambda \|\vec{x}\|^2 > 0$ , and since  $\|\vec{x}\|^2 > 0$  hence  $\lambda > 0$ To compute *R* we first need to find *Q*.

$$A = \begin{pmatrix} 10 & 6\\ 6 & 10 \end{pmatrix} \to l_{21} = \frac{6}{10} \to \begin{pmatrix} 10 & 6\\ 6 - \frac{6}{10} \times 10 & 10 - \frac{6}{10} \times 6 \end{pmatrix} \to \begin{pmatrix} 10 & 6\\ 0 & \frac{32}{5} \end{pmatrix}$$
  
Hence  $L = \begin{pmatrix} 1 & 0\\ \frac{6}{10} & 1 \end{pmatrix}, U = \begin{pmatrix} 10 & 6\\ 0 & \frac{32}{5} \end{pmatrix}$ 

Then

$$LDU = \begin{pmatrix} 1 & 0\\ \frac{6}{10} & 1 \end{pmatrix} \begin{pmatrix} 10 & 0\\ 0 & \frac{32}{5} \end{pmatrix} \begin{pmatrix} 1 & \frac{6}{10}\\ 0 & 1 \end{pmatrix}$$
$$= \begin{pmatrix} 1 & 0\\ \frac{6}{10} & 1 \end{pmatrix} \begin{pmatrix} 10 & 0\\ 0 & \frac{32}{5} \end{pmatrix} \begin{pmatrix} 1 & 0\\ \frac{6}{10} & 1 \end{pmatrix}^{T}$$

Hence we see that  $Q = L = \begin{pmatrix} 1 & 0 \\ \frac{6}{10} & 1 \end{pmatrix}$ ,  $\Lambda = D = \begin{pmatrix} 10 & 0 \\ 0 & \frac{32}{5} \end{pmatrix}$ ,  $Q^T = L^T$ Since *A* is SPD, then  $A = R^T R$  and  $A = Q\Lambda Q^T$ , hence we can take  $R = \sqrt{\Lambda}Q^T$ 

$$R = \sqrt{\Lambda}Q^{T} = \sqrt{\begin{pmatrix} 10 & 0\\ 0 & \frac{32}{5} \end{pmatrix}} \begin{pmatrix} 1 & \frac{6}{10}\\ 0 & 1 \end{pmatrix}$$
$$= \begin{pmatrix} \sqrt{10} & 0\\ 0 & \sqrt{\frac{32}{5}} \end{pmatrix} \begin{pmatrix} 1 & \frac{6}{10}\\ 0 & 1 \end{pmatrix}$$
$$= \begin{pmatrix} \sqrt{10} & \frac{3}{5}\sqrt{2}\sqrt{5}\\ 0 & \frac{4}{5}\sqrt{2}\sqrt{5} \end{pmatrix}$$

Verify that  $R^T R = A$ 

$$R^{T}R = \begin{pmatrix} \sqrt{10} & 0\\ \frac{3}{5}\sqrt{2}\sqrt{5} & \frac{4}{5}\sqrt{2}\sqrt{5} \end{pmatrix} \begin{pmatrix} \sqrt{10} & \frac{3}{5}\sqrt{2}\sqrt{5}\\ 0 & \frac{4}{5}\sqrt{2}\sqrt{5} \end{pmatrix}$$
$$= \begin{pmatrix} 10 & 6\\ 6 & 10 \end{pmatrix}$$

verified oK.

Now do the same for 
$$A = \begin{pmatrix} 10 & -6 \\ -6 & 10 \end{pmatrix}$$
  
 $A = \begin{pmatrix} 10 & -6 \\ -6 & 10 \end{pmatrix} \rightarrow l_{21} = \frac{-6}{10} \rightarrow U = \begin{pmatrix} 10 & -6 \\ -6 - \frac{-6}{10} \times 10 & 10 - \frac{-6}{10} \times -6 \end{pmatrix} \rightarrow \begin{pmatrix} 10 & -6 \\ 0 & \frac{32}{5} \end{pmatrix}$   
Hence  $L = \begin{pmatrix} 1 & 0 \\ -\frac{6}{10} & 1 \end{pmatrix}, U = \begin{pmatrix} 10 & -6 \\ 0 & \frac{32}{5} \end{pmatrix}$   
Then

$$LDU = \begin{pmatrix} 1 & 0 \\ -\frac{6}{10} & 1 \end{pmatrix} \begin{pmatrix} 10 & 0 \\ 0 & \frac{32}{5} \end{pmatrix} \begin{pmatrix} 1 & -\frac{6}{10} \\ 0 & 1 \end{pmatrix}$$
$$= \begin{pmatrix} 1 & 0 \\ -\frac{6}{10} & 1 \end{pmatrix} \begin{pmatrix} 10 & 0 \\ 0 & \frac{32}{5} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ -\frac{6}{10} & 1 \end{pmatrix}^{T}$$

Hence we see that  $Q = L = \begin{pmatrix} 1 & 0 \\ -\frac{6}{10} & 1 \end{pmatrix}$ ,  $\Lambda = D = \begin{pmatrix} 10 & 0 \\ 0 & \frac{32}{5} \end{pmatrix}$ ,  $Q^T = L^T$ Then now we find R

Since A is SPD, then  $A = R^T R$  and  $A = Q \Lambda Q^T$ , hence we can take  $R = \sqrt{\Lambda} Q^T$ 

$$R = \sqrt{\Lambda}Q^{T} = \sqrt{\left( \begin{matrix} 10 & 0 \\ 0 & \frac{32}{5} \end{matrix} \right)} \begin{pmatrix} 1 & -\frac{6}{10} \\ 0 & 1 \end{pmatrix}$$
$$= \left( \begin{matrix} \sqrt{10} & 0 \\ 0 & \sqrt{\frac{32}{5}} \end{matrix} \right) \begin{pmatrix} 1 & -\frac{6}{10} \\ 0 & 1 \end{pmatrix}$$
$$= \left( \begin{matrix} \sqrt{10} & -\frac{3}{5}\sqrt{2}\sqrt{5} \\ 0 & \frac{4}{5}\sqrt{2}\sqrt{5} \end{matrix} \right)$$

Verify that  $R^T R = A$ 

$$R^{T}R = \begin{pmatrix} \sqrt{10} & 0\\ -\frac{3}{5}\sqrt{2}\sqrt{5} & \frac{4}{5}\sqrt{2}\sqrt{5} \end{pmatrix} \begin{pmatrix} \sqrt{10} & -\frac{3}{5}\sqrt{2}\sqrt{5}\\ 0 & \frac{4}{5}\sqrt{2}\sqrt{5} \end{pmatrix}$$
$$= \begin{pmatrix} 10 & -6\\ -6 & 10 \end{pmatrix}$$

verified oK.

#### 11 Section 6.2, problem 4

Show from the eigenvalues that if A is P.D. so is  $A^2$  and so is  $A^{-1}$ answer: Given A is PD. Hence Eigenvalues of A are positive. Let eigenvalue of A be  $\lambda_A$ Let  $B = A^2$ Let eigenvalue of B be  $\lambda_B$ We need to show that  $\lambda_B > 0$ Now

$$Bx = \lambda_B x$$
$$A^2 x = \lambda_B x$$
$$AAx = \lambda_B x$$
$$A\lambda_A x = \lambda_B x$$
$$\lambda_A Ax = \lambda_B x$$
$$\lambda_A Ax = \lambda_B x$$
$$\lambda_A \lambda_A x = \lambda_B x$$

From the last statement above we can now say

$$\lambda_A^2 = \lambda_B$$

Hence  $\lambda_B > 0$ , hence by theorem 6B which says that if all eigenvalues are positive then the matrix is PD, then in this case the matrix *B* which is  $A^2$  is PD. QED Now for  $A^{-1}$ 

$$Ax = \lambda_A x$$

pre multiply both sides by 
$$A^{-1}$$

$$\overbrace{A^{-1}Ax}^{I} = A^{-1}\lambda_{A}x$$
$$x = A^{-1}\lambda_{A}x$$
$$\frac{1}{\lambda_{A}}x = A^{-1}x$$

i.e.

$$A^{-1}x = \frac{1}{\lambda_A}x$$

Hence eigenvalue of  $A^{-1}$  is  $\frac{1}{\lambda_A}$ . And since  $\lambda_A > 0$ , then so is  $\frac{1}{\lambda_A}$ , and by theorem 6B again, since all eigenvalues are positive then  $A^{-1}$  is P.D.

#### 12 Section 6.2, problem 6

From the pivots, eigenvalues, eigenvectors of  $A = \begin{pmatrix} 5 & 4 \\ 4 & 5 \end{pmatrix}$ , write A as  $R^T R$  in 3 ways

1.  $\left(L\sqrt{D}\right)\left(\sqrt{D}L^{T}\right)$ 2.  $\left(Q\sqrt{\Lambda}\right)\left(\sqrt{\Lambda}Q^{T}\right)$ 3.  $\left(Q\sqrt{\Lambda}Q^{T}\right)\left(Q\sqrt{\Lambda}Q^{T}\right)$ 

Answer:

First find if A is PD or not. Since this is a 2 by 2 matrix, a simple test is to look at the quantity  $a^2 - bc$  and if it is positive, and if *a* is also positive, then the matrix is PD

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$
$$a = 5 > 0$$
$$a^{2} - bc = 25 - 16$$
$$= 9 > 0$$

hence A is P.D.

Then it can be written as  $R^T R$  where *R* is full rank square matrix. 1) Since *A* is symmetric *P*.*D*., then it has choleskly decomposition  $CC^T$  where  $C = L\sqrt{D}$ , and  $C^T = \sqrt{D}L^T$  (the pivots are positive in the *D* matrix diagonal, so we can take their square root) The provide  $A = D^T R - (L\sqrt{D}) \left( \sqrt{D}L^T \right)$  is  $R = C - (\sqrt{D}L^T)$ .

Then we write 
$$A = R^T R = (L\sqrt{D}) (\sqrt{D}L^T)$$
 where  $R = (\sqrt{D}L^T)$   
 $\begin{pmatrix} 5 & 4 \\ 4 & 5 \end{pmatrix} \rightarrow l_{21} = \frac{4}{5} \rightarrow U = \begin{pmatrix} 5 & 4 \\ 0 & 5 - \frac{4}{5} \times 4 \end{pmatrix} = \begin{pmatrix} 5 & 4 \\ 0 & \frac{9}{5} \end{pmatrix}$   
Hence  $L = \begin{pmatrix} 1 & 0 \\ \frac{4}{5} & 1 \end{pmatrix}, U = \begin{pmatrix} 5 & 4 \\ 0 & \frac{9}{5} \end{pmatrix} \rightarrow LDU = \begin{pmatrix} 1 & 0 \\ \frac{4}{5} & 1 \end{pmatrix} \begin{pmatrix} 5 & 0 \\ 0 & \frac{9}{5} \end{pmatrix} \begin{pmatrix} 1 & \frac{4}{5} \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} \sqrt{5} & 0 \\ 0 & \sqrt{\frac{9}{5}} \end{pmatrix} \begin{pmatrix} 1 & \frac{4}{5} \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} \sqrt{5} & \frac{4}{5}\sqrt{5} \\ 0 & \frac{3}{5}\sqrt{5} \end{pmatrix}$ 

Hence

$$A = \overbrace{\begin{pmatrix} \sqrt{5} & 0\\ \frac{4}{5}\sqrt{5} & \frac{3}{5}\sqrt{5} \end{pmatrix}}^{L\sqrt{D}} \overbrace{\begin{pmatrix} \sqrt{5} & \frac{4}{5}\sqrt{5}\\ 0 & \frac{3}{5}\sqrt{5} \end{pmatrix}}^{\sqrt{D}L^T}$$

2)From  $A = Q\Lambda Q^T$  where Q is the matrix which contains as its columns the normalized eigenvectors of A and  $\Lambda$  contains in its diagonal the eigenvalues of A. First start by finding eigenvalues and eigenvectors of A

$$\begin{pmatrix} 5 & 4 \\ 4 & 5 \end{pmatrix} \rightarrow \text{eigenvectors:} \left\{ \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\} \leftrightarrow 1, \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\} \leftrightarrow 9$$

$$\text{Hence } Q = \begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix} \rightarrow \text{normalize columns} \rightarrow Q = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix}, \Lambda = \begin{pmatrix} 1 & 0 \\ 0 & 9 \end{pmatrix}$$

So, verify first that the above is correct:

$$A = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0\\ 0 & 9 \end{pmatrix} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \frac{1}{\sqrt{2}} = \frac{1}{2} \times \begin{pmatrix} 10 & 8\\ 8 & 10 \end{pmatrix} = \begin{pmatrix} 5 & 4\\ 4 & 5 \end{pmatrix}$$
  
Correct. So we write  $R = \left(\sqrt{\Lambda}Q^T\right) = \sqrt{\begin{pmatrix} 1 & 0\\ 0 & 9 \end{pmatrix}} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \frac{1}{\sqrt{2}} = \begin{pmatrix} 1 & 0\\ 0 & 3 \end{pmatrix} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \frac{1}{\sqrt{2}} = \begin{pmatrix} -1 & 1\\ 3 & 3 \end{pmatrix} \frac{1}{\sqrt{2}}$   
Hence

$$A = R^{T}R$$

$$= \underbrace{\frac{R^{T} = Q\sqrt{\Lambda}}{\frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 3\\ 1 & 3 \end{pmatrix} \begin{pmatrix} -1 & 1\\ 3 & 3 \end{pmatrix} \frac{1}{\sqrt{2}}}}_{R = \sqrt{\Lambda}Q^{T}}$$

3) now find  $R = \left(Q\sqrt{\Lambda}Q^T\right)$   $R = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \sqrt{\begin{pmatrix} 1 & 0\\ 0 & 9 \end{pmatrix}} \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0\\ 0 & 3 \end{pmatrix} \begin{pmatrix} -1 & 1\\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1\\ 1 & 2 \end{pmatrix}$ Hence

$$A = R^{T}R$$

$$= \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}^{T} \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$$

$$R^{T} = Q\sqrt{\Lambda}Q^{T}R = Q\sqrt{\Lambda}Q^{T}$$

$$= \overbrace{\begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}}^{T} \overbrace{\begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}}^{T}$$

# 13 Section 6.2 problem 8

problem: if *A* is SPD and C is nonsignular, prove that  $B = C^T A C$  is also SPD solution: Since *A* is SPD, then it has positive eigenvalues.

Since *B* is similar to *A* (given), then *B* has the same eigenvalues as *A*, Hence *B* also has all its eigenvalues positive.

Hence by theorem 6B, *B* is symmetric positive definite.