

42/40.

MATH 309 HW8

Sample solution based on that of Fangming Xu

5 I.10 #4 $\det(S - \lambda M) = 0$, so $(25 - 5\lambda) - 16 = 0$. So $\lambda_1 = \frac{9}{5}$. The corresponding eigen-vector is $\begin{bmatrix} 1 \\ -\frac{4}{5} \end{bmatrix}$. In the other case, $\lambda_2 = \infty$. Thus, the eigen-vector should be $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

5 I.10 #6 (a) Since S is symmetric, $S = S^T$, so $Z^T S Z = Z^T S^T Z = (Z^T S Z)^T$. So it preserves the symmetry.
(b) Since S is positive definite, we can write $Z^T S Z = Z^T U D U^T Z = Y D Y^T$. So $x^T Z^T S Z x = x^T Y D Y^T x = Y^T (\lambda_1 y_1^2 + \dots + \lambda_n y_n^2) x > 0$. So $Z^T S Z$ is also positive definite. Zx is not the zero vector because Z is invertible, so it must have rank n.

you can do it in a easier way: congruence preserves signs of eigenvalues, and $\lambda > 0$ means P.D.

5 I.10 #7 Positive definite matrices are congruent to the identity matrix. $x^T A x = x^T Z^T I Z x = x^T z^T Z x = Y^T Y > 0$, so A has positive eigen values and A is positive definite.

3 I.10 #8 $S = uu^T$ so S is positive definite. Thus we can write S as $S = S^{\frac{1}{2}} S^{\frac{1}{2}}$ where $S^{\frac{1}{2}} = U \Sigma^{\frac{1}{2}} U^T$. Since we need to solve $M^{-1} S x = \lambda x$, now we can let $v = S^{\frac{1}{2}} x$ and solve $S^{\frac{1}{2}} M^{-1} S^{\frac{1}{2}} v = \lambda S^{\frac{1}{2}} x = \lambda v$, which is basically the eigenvalue and eigen vector of $S^{\frac{1}{2}} M^{-1} S^{\frac{1}{2}}$. Then, just choose the minimum eigenvalue and the corresponding eigenvector.

→ consider I.10 #3...

5 I.11 #1 $\|v\|_2^2 = |v_1|^2 + \dots + |v_n|^2 \leq (|v_1| + \dots + |v_n|) \cdot \max(|v_j|) = \|v\|_1 \|v\|_\infty$.

5 I.11 #3 Since $(\sum |v_j|^2) \leq n \cdot \max\{|v_j|\}^2$, $\|v\|_2 \leq \sqrt{n} \|v\|_\infty$. For the second question, we know that $(\sum |v_j|^2) \leq n \sum (|v_j|^2)$, ($w = (1,1,1,\dots, 1)$.) So $\|v\|_1 \leq \sqrt{n} \|v\|_2$.

4 I.11 #10 I have no idea what I am supposed to prove... Is it just $\|A\|_F^2 = (\text{tr}(A^T A))^{\frac{1}{2}} = \text{tr}(A^T A) = (A, A)$. **Yes, we do this by direct computations.** Let $A = [A_{11} \dots A_{1n}]$. Then $\|A\|_F^2 = \sum_{i,j} |a_{ij}|^2 = \sum_{j=1}^n \sum_{i=1}^m |a_{ij}|^2 = \sum_j A_j^T A_j = \text{tr}(A^T A)$.

5 I.11 #12 For any a_{ij} , we have $a_{ij} = \sum_{k=1}^n a_{ik} \cdot b_{kj} \leq n \cdot \max\{\text{entry in } A\} \cdot \max\{\text{entry in } B\} = n \|A\|_\infty \|B\|_\infty$. Thus, $\|AB\|_\infty \leq n \|A\|_\infty \|B\|_\infty$. The next step is just multiplying both sides by \sqrt{mp} and rearrange. $\sqrt{mp} \|AB\|_\infty \leq \sqrt{mp} \cdot n \|A\|_\infty \|B\|_\infty$, so $\sqrt{mp} \|AB\|_\infty \leq \sqrt{mn} \|A\|_\infty \sqrt{np} \|B\|_\infty$.

5 Bonus

(Forward:) $\text{trace}(A^T B) = \sum A_{ij} B_{ij} = \|A\| \|B\| = \sqrt{\sum |A_{ij}|^2} \cdot \sqrt{\sum |B_{ij}|^2}$, so $\sum |A_{ij} B_{ij}| \cdot \sum |A_{ij} B_{ij}| = \sum |A_{ij}|^2 \cdot \sum |B_{ij}|^2$. Expanding both sides, we can see that while (A,B) is nonzero, B has to be a multiple of A for this equation to hold.

(Backward:) Write $B = k A$, while k is a non-zero constant. Thus, we can write $(A,B) = \text{trace}(A^T B) = \sum A_{ij} B_{ij} = \sum A_{ij} k A_{ij} = \sqrt{\sum |A_{ij}|^2} \cdot \sqrt{\sum k |A_{ij}|^2} = \sqrt{\sum |A_{ij}|^2} \cdot \sqrt{\sum |B_{ij}|^2}$.

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