

**SOLUTION KEY TO THE LINEAR ALGEBRA
MAKE-UP FINAL EXAM**

(1) We find a least squares solution to

$$A\vec{x} = \vec{y} \quad \text{or} \quad \begin{bmatrix} 1 & \sin 0 & \cos 0 \\ 1 & \sin \pi/2 & \cos \pi/2 \\ 1 & \sin \pi & \cos \pi \\ 1 & \sin 3\pi/2 & \cos 3\pi/2 \\ 1 & \sin 2\pi & \cos 2\pi \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 7 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}.$$

The normal equation is

$$A^T A \vec{x}_* = A^T \vec{y} = \vec{y}_* \quad \text{or} \quad \begin{bmatrix} 5 & 0 & 1 \\ 0 & 2 & 0 \\ 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} a_* \\ b_* \\ c_* \end{bmatrix} = \begin{bmatrix} 7 \\ 0 \\ 7 \end{bmatrix}.$$

The least-squares solution is

$$\vec{x}_* = \begin{bmatrix} a_* \\ b_* \\ c_* \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

so the sought-after trigonometric polynomial is $p(t) = 1 + 2 \cos t$.

(2) (a) The hyperbola is $q(\vec{x}) = 1$ where $q(\vec{x}) = \vec{x}^T A \vec{x}$ and

$$A = \frac{1}{8} \begin{bmatrix} 7 & -3 \\ -3 & -1 \end{bmatrix}.$$

We have $p_A(\lambda) = \lambda^2 - (3/4)\lambda - 1/4 = (\lambda - 1)(\lambda + 1/4)$ so the eigenvalues of A are $\lambda_1 = +1, \lambda_2 = -1/4$. The principal axes are

$$c_1 \text{ axis: } E_{+1} = \text{Ker}(1I - A) = \text{span } \vec{u}_1, \quad \vec{u}_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ -1 \end{bmatrix}$$

$$c_2 \text{ axis: } E_{-1/4} = \text{Ker}\left(-\frac{1}{4}I - A\right) = \text{span } \vec{u}_2, \quad \vec{u}_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ 3 \end{bmatrix}.$$

(b) In c_1 - c_2 coordinates: $q(\vec{x}) = \lambda_1 c_1^2 + \lambda_2 c_2^2$ so the equation of the hyperbola becomes

$$c_1^2 - \frac{1}{4}c_2^2 = 1.$$

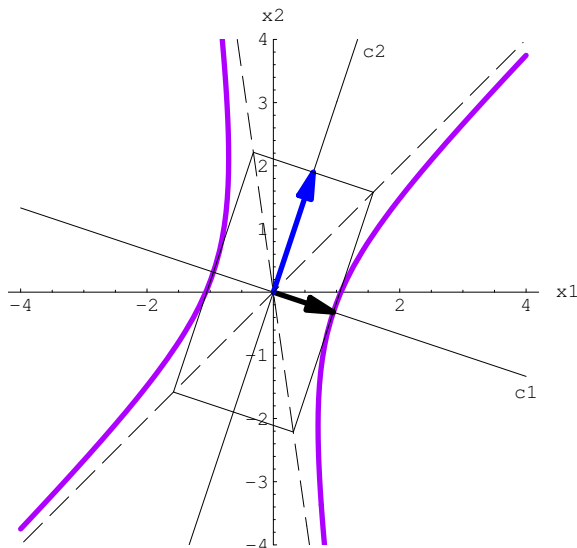


FIGURE 1. The hyperbola $7x_1^2 - 6x_1x_2 - x_2^2 = 8$ with its principal axes and the vectors \vec{u}_1 (black) and $2\vec{u}_2$ (blue).

- (c) The asymptote are spanned by the vectors $\vec{d}_\pm = |\lambda_1|^{-1/2}u_1 \pm |\lambda_2|^{-1/2}u_2$, namely

$$\vec{d}_+ = \vec{u}_1 + 2\vec{u}_2 = \sqrt{\frac{5}{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \vec{d}_- = \vec{u}_1 - 2\vec{u}_2 = \sqrt{\frac{1}{10}} \begin{bmatrix} 1 \\ -7 \end{bmatrix}.$$

- (3) (a) $p_A(\lambda) = \det(\lambda I - A) = \lambda^3 - 2\lambda^2 - 2\lambda = \lambda(\lambda^2 - 2\lambda - 2)$.
The roots of the quadratic factor are the complex numbers $1 \pm i$ so the eigenvalues are $\lambda_1 = 0, \lambda_2 = 1 + i, \lambda_3 = 1 - i$.
- (b)

$$(1) \quad E_0 = \text{Ker}(0I - A) = \text{span} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$(2) \quad E_{1 \pm i} = \text{Ker}((1 \pm i)I - A) = \text{span} \begin{bmatrix} 0 \\ 1 \\ \pm i \end{bmatrix},$$

so

$$S = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & i & -i \end{bmatrix}, \quad D = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1+i & 0 \\ 0 & 0 & 1-i \end{bmatrix}$$

and $S^{-1}AS = D$ is the desired diagonalization of A .

- (4) (a) Direct calculation shows $L(E_{11}) = E_{11}$, $L(E_{22}) = E_{22}$,
 $L(E_{12}) = L(E_{21}) = (E_{12} + E_{21})/2$, so

$$M = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1/2 & 1/2 & 0 \\ 0 & 1/2 & 1/2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

- (b) Since

$$\text{rref}(M) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

has pivots in the first, second and fourth columns it follows that $\text{Im}(M)$ has as basis the corresponding columns of M , so we may take $\mathfrak{B} = \{E_{11}, (E_{12} + E_{21})/2, E_{22}\}$. In fact, changing this basis just a bit we obtain an orthonormal basis $\{E_{11}, (E_{12} + E_{21})/\sqrt{2}, E_{22}\}$.

- (c) Since L is an orthogonal projection, $\mathcal{S}^\perp = \text{Ker}(L)$, so we start by finding a basis for $\text{Ker}(M)$. From $\text{rref}(M)$ it is easy to read off the basis consisting of the single vector $[0, -1, 1, 0]$, hence $\text{Ker } L$ has a basis consisting of the matrix $N = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$. Since $\|N\|^2 = \text{Trace}(N^T N) = 2$, the desired orthonormal basis is

$$\mathfrak{U} = \{U\}, \quad U = \frac{1}{\sqrt{2}} \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}.$$

- (d)

$$\begin{aligned} P(A) &= \langle N, A \rangle N = \frac{1}{2} \text{Trace}(N^T A) N = \frac{1}{2} \begin{bmatrix} 0 & a_{12} - a_{21} \\ a_{21} - a_{12} & 0 \end{bmatrix} \\ &= \frac{1}{2}(A - A^T). \end{aligned}$$

Another, more direct, way of deducing this is as follows. It is geometrically obvious that $A = \text{proj}_{\mathcal{S}} A + \text{proj}_{\mathcal{S}^\perp} A$ for any matrix $A \in \mathbb{R}^{2 \times 2}$ and subspace $\mathcal{S} \subset \mathbb{R}^{2 \times 2}$ (think of adding the projections of a vector onto two mutually orthogonal lines in \mathbb{R}^2), which in our case can be read to say $L(A) + P(A) = A$, so $P(A) = A - (1/2)(A + A^T) = (1/2)(A - A^T)$.

- (5) (a) *False*. $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ is a counterexample.

- (b) *False.* $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ is a counterexample.
- (c) *False.* $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ are symmetric, yet $AB = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ is not.
- (d) *True.* If A is diagonalizable, it is similar to a diagonal matrix, hence *a fortiori* to a symmetric matrix.
- (e) *True.* Since $A^2 = 0$ then $A(A\vec{x}) = \vec{0}$, so any vector $A\vec{x} \in \text{Im}(A)$ is in $\text{Ker}(A)$, therefore $\text{Im}(A) \subset \text{Ker}(A)$. Hence, $\dim(\text{Im}(A)) \leq \dim(\text{Ker}(A))$ so the rank r and the nullity n of A satisfy $r \leq n$. Since $r + n = 10$ by the rank-nullity theorem, $r \leq 5$.
- (f) *True.* By the fundamental theorem of linear algebra, $\text{Ker}(A)$ and $\text{Im}(A^T)$ are mutually orthogonally complementary subspaces. Since $A = A^T$, $\text{Im}(A^T) = \text{Im}(A)$ and any two vectors $\vec{x} \in \text{Ker}(A)$ and $\vec{y} \in \text{Im}(A)$ must be orthogonal.
- (g) *False.* $S^T AS = D$ for some orthogonal S and diagonal D . Squaring the equation and recalling that S and S^T are the inverses of one another we obtain

$$D^2 = (S^T AS)^2 = S^T ASS^T AS = S^T A I A S = S^T A^2 S = S^T 0 S = 0$$

since $A^2 = 0$. By looking at the diagonal of D we conclude that $\lambda^2 = 0$ for any eigenvalue λ of A , hence D itself is the zero matrix. Therefore $A = SDS^T = S0S^T = 0$.