

- 7 Invertible 3 by 3 matrix A : row space basis = column space basis = $(1, 0, 0)$, $(0, 1, 0)$, $(0, 0, 1)$; nullspace basis and left nullspace basis are *empty*. Matrix $B = \begin{bmatrix} A & A \end{bmatrix}$: row space basis $(1, 0, 0, 1, 0, 0)$, $(0, 1, 0, 0, 1, 0)$ and $(0, 0, 1, 0, 0, 1)$; column space basis $(1, 0, 0)$, $(0, 1, 0)$, $(0, 0, 1)$; nullspace basis $(-1, 0, 0, 1, 0, 0)$ and $(0, -1, 0, 0, 1, 0)$ and $(0, 0, -1, 0, 0, 1)$; left nullspace basis is empty.
- 8 $\begin{bmatrix} I & 0 \end{bmatrix}$ and $\begin{bmatrix} I & I; & 0 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 \end{bmatrix}$ = 3 by 2 have row space dimensions = 3, 3, 0 = column space dimensions; nullspace dimensions 2, 3, 2; left nullspace dimensions 0, 2, 3.
- 9 (a) Same row space and nullspace. So rank (dimension of row space) is the same
(b) Same column space and left nullspace. Same rank (dimension of column space).
- 10 For **rand** (3), almost surely rank = 3, nullspace and left nullspace contain only $(0, 0, 0)$.
For **rand** (3, 5) the rank is almost surely 3 and the dimension of the nullspace is 2.
- 11 (a) No solution means that $r < m$. Always $r \leq n$. Can't compare m and n here.
(b) Since $m - r > 0$, the left nullspace must contain a nonzero vector.
- 12 A neat choice is $\begin{bmatrix} 1 & 1 \\ 0 & 2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 1 & 2 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 2 & 1 \\ 2 & 4 & 0 \\ 1 & 0 & 1 \end{bmatrix}$; $r + (n - r) = n = 3$ does not match $2 + 2 = 4$. Only $\mathbf{v} = \mathbf{0}$ is in both $N(A)$ and $C(A^T)$.
- 13 (a) *False*: Usually row space \neq column space (same dimension!) (b) *True*: A and $-A$ have the same four subspaces (c) *False* (choose A and B same size and invertible: then they have the same four subspaces)
- 14 Row space basis can be the nonzero rows of U : $(1, 2, 3, 4)$, $(0, 1, 2, 3)$, $(0, 0, 1, 2)$; nullspace basis $(0, 1, -2, 1)$ as for U ; column space basis $(1, 0, 0)$, $(0, 1, 0)$, $(0, 0, 1)$ (happen to have $C(A) = C(U) = \mathbf{R}^3$); left nullspace has empty basis.
- 15 After a row exchange, the row space and nullspace stay the same; $(2, 1, 3, 4)$ is in the new left nullspace after the row exchange.
- 16 If $A\mathbf{v} = \mathbf{0}$ and \mathbf{v} is a row of A then $\mathbf{v} \cdot \mathbf{v} = 0$.
- 17 Row space = yz plane; column space = xy plane; nullspace = x axis; left nullspace = z axis. For $I + A$: Row space = column space = \mathbf{R}^3 , both nullspaces contain only the zero vector.
- 18 Row $3 - 2$ row $2 +$ row $1 =$ zero row so the vectors $c(1, -2, 1)$ are in the left nullspace. The same vectors happen to be in the nullspace (an accident for this matrix).
- 19 (a) Elimination on $A\mathbf{x} = \mathbf{0}$ leads to $0 = b_3 - b_2 - b_1$ so $(-1, -1, 1)$ is in the left nullspace. (b) 4 by 3: Elimination leads to $b_3 - 2b_1 = 0$ and $b_4 + b_2 - 4b_1 = 0$, so $(-2, 0, 1, 0)$ and $(-4, 1, 0, 1)$ are in the left nullspace. *Why?* Those vectors multiply the matrix to give *zero rows*. Section 4.1 will show another approach: $A\mathbf{x} = \mathbf{b}$ is solvable (\mathbf{b} is in $C(A)$) when \mathbf{b} is orthogonal to the left nullspace.
- 20 (a) Special solutions $(-1, 2, 0, 0)$ and $(-\frac{1}{4}, 0, -3, 1)$ are perpendicular to the rows of R (and then ER). (b) $A^T\mathbf{y} = \mathbf{0}$ has 1 independent solution = last row of E^{-1} . ($E^{-1}A = R$ has a zero row, which is just the transpose of $A^T\mathbf{y} = \mathbf{0}$).
- 21 (a) \mathbf{u} and \mathbf{w} (b) \mathbf{v} and \mathbf{z} (c) rank < 2 if \mathbf{u} and \mathbf{w} are dependent or if \mathbf{v} and \mathbf{z} are dependent (d) The rank of $\mathbf{u}\mathbf{v}^T + \mathbf{w}\mathbf{z}^T$ is 2.
- 22 $A = \begin{bmatrix} \mathbf{u} & \mathbf{w} \end{bmatrix} \begin{bmatrix} \mathbf{v}^T & \mathbf{z}^T \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 2 & 2 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 4 & 2 \\ 5 & 1 \end{bmatrix}$ has column space spanned by \mathbf{u} and \mathbf{w} , row space spanned by \mathbf{v} and \mathbf{z} .

- 23** As in Problem 22: Row space basis $(3, 0, 3), (1, 1, 2)$; column space basis $(1, 4, 2), (2, 5, 7)$; the rank of $(3 \text{ by } 2) \text{ times } (2 \text{ by } 3)$ cannot be larger than the rank of either factor, so $\text{rank} \leq 2$ and the $3 \text{ by } 3$ product is not invertible.
- 24** $A^T y = d$ puts d in the row space of A ; unique solution if the left nullspace (nullspace of A^T) contains only $y = 0$.
- 25** (a) True (A and A^T have the same rank) (b) False $A = \begin{bmatrix} 1 & 0 \end{bmatrix}$ and A^T have very different left nullspaces (c) False (A can be invertible and unsymmetric even if $C(A) = C(A^T)$) (d) True (The subspaces for A and $-A$ are always the same. If $A^T = A$ or $A^T = -A$ they are also the same for A^T)
- 26** The rows of $C = AB$ are combinations of the rows of B . So $\text{rank } C \leq \text{rank } B$. Also $\text{rank } C \leq \text{rank } A$, because the columns of C are combinations of the columns of A .
- 27** Choose $d = bc/a$ to make $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ a rank-1 matrix. Then the row space has basis (a, b) and the nullspace has basis $(-b, a)$. Those two vectors are perpendicular!
- 28** B and C (checkers and chess) both have rank 2 if $p \neq 0$. Row 1 and 2 are a basis for the row space of C , $B^T y = 0$ has 6 special solutions with -1 and 1 separated by a zero; $N(C^T)$ has $(-1, 0, 0, 0, 0, 0, 1)$ and $(0, -1, 0, 0, 0, 0, 1, 0)$ and columns 3, 4, 5, 6 of I ; $N(C)$ is a challenge.
- 29** $a_{11} = 1, a_{12} = 0, a_{13} = 1, a_{22} = 0, a_{32} = 1, a_{31} = 0, a_{23} = 1, a_{33} = 0, a_{21} = 1$.
- 30** The subspaces for $A = uv^T$ are pairs of orthogonal lines (v and v^\perp , u and u^\perp). If B has those same four subspaces then $B = cA$ with $c \neq 0$.
- 31** (a) $AX = 0$ if each column of X is a multiple of $(1, 1, 1)$; $\dim(\text{nullspace}) = 3$.
 (b) If $AX = B$ then all columns of B add to zero; dimension of the B 's = 6.
 (c) $3 + 6 = \dim(M^{3 \times 3}) = 9$ entries in a $3 \text{ by } 3$ matrix.
- 32** The key is equal row spaces. First row of $A =$ combination of the rows of B : only possible combination (notice I) is 1 (row 1 of B). Same for each row so $F = G$.

Problem Set 4.1, page 202

- 1** Both nullspace vectors are orthogonal to the row space vector in \mathbf{R}^3 . The column space is perpendicular to the nullspace of A^T (two lines in \mathbf{R}^2 because $\text{rank} = 1$).
- 2** The nullspace of a $3 \text{ by } 2$ matrix with rank 2 is \mathbf{Z} (only zero vector) so $x_n = 0$, and row space = \mathbf{R}^2 . Column space = plane perpendicular to left nullspace = line in \mathbf{R}^3 .
- 3** (a) $\begin{bmatrix} 1 & 2 & -3 \\ 2 & -3 & 1 \\ -3 & 5 & -2 \end{bmatrix}$ (b) Impossible, $\begin{bmatrix} 2 \\ -3 \\ 5 \end{bmatrix}$ not orthogonal to $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ (c) $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ in $C(A)$ and $N(A^T)$ is impossible: not perpendicular (d) Need $A^2 = 0$; take $A = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$ (e) $(1, 1, 1)$ in the nullspace (columns add to 0) and also row space; no such matrix.
- 4** If $AB = 0$, the columns of B are in the nullspace of A . The rows of A are in the left nullspace of B . If $\text{rank} = 2$, those four subspaces would have dimension 2 which is impossible for $3 \text{ by } 3$.
- 5** (a) If $Ax = b$ has a solution and $A^T y = 0$, then y is perpendicular to b . $b^T y = (Ax)^T y = x^T (A^T y) = 0$. (b) If $A^T y = (1, 1, 1)$ has a solution, $(1, 1, 1)$ is in the row space and is orthogonal to every x in the nullspace.

- 6** Multiply the equations by $y_1, y_2, y_3 = 1, 1, -1$. Equations add to $0 = 1$ so no solution: $\mathbf{y} = (1, 1, -1)$ is in the left nullspace. $A\mathbf{x} = \mathbf{b}$ would need $0 = (\mathbf{y}^T A)\mathbf{x} = \mathbf{y}^T \mathbf{b} = 1$.
- 7** Multiply the 3 equations by $\mathbf{y} = (1, 1, -1)$. Then $x_1 - x_2 = 1$ plus $x_2 - x_3 = 1$ minus $x_1 - x_3 = 1$ is $0 = 1$. Key point: This \mathbf{y} in $N(A^T)$ is not orthogonal to $\mathbf{b} = (1, 1, 1)$ so \mathbf{b} is not in the column space and $A\mathbf{x} = \mathbf{b}$ has *no solution*.
- 8** $\mathbf{x} = \mathbf{x}_r + \mathbf{x}_n$, where \mathbf{x}_r is in the row space and \mathbf{x}_n is in the nullspace. Then $A\mathbf{x}_n = \mathbf{0}$ and $A\mathbf{x} = A\mathbf{x}_r + A\mathbf{x}_n = A\mathbf{x}_r$. All $A\mathbf{x}$ are in $C(A)$.
- 9** $A\mathbf{x}$ is always in the *column space* of A . If $A^T A\mathbf{x} = \mathbf{0}$ then $A\mathbf{x}$ is also in the nullspace of A^T . So $A\mathbf{x}$ is perpendicular to itself. Conclusion: $A\mathbf{x} = \mathbf{0}$ if $A^T A\mathbf{x} = \mathbf{0}$.
- 10** (a) With $A^T = A$, the column and row spaces are the same (b) \mathbf{x} is in the nullspace and \mathbf{z} is in the column space = row space: so these “eigenvectors” have $\mathbf{x}^T \mathbf{z} = 0$.
- 11** **For A:** The nullspace is spanned by $(-2, 1)$, the row space is spanned by $(1, 2)$. The column space is the line through $(1, 3)$ and $N(A^T)$ is the perpendicular line through $(3, -1)$. **For B:** The nullspace of B is spanned by $(0, 1)$, the row space is spanned by $(1, 0)$. The column space and left nullspace are the same as for A .
- 12** \mathbf{x} splits into $\mathbf{x}_r + \mathbf{x}_n = (1, -1) + (1, 1) = (2, 0)$. Notice $N(A^T)$ is a plane $(1, 0) = (1, 1)/2 + (1, -1)/2 = \mathbf{x}_r + \mathbf{x}_n$.
- 13** $V^T W = \text{zero}$ makes each basis vector for V orthogonal to each basis vector for W . Then every \mathbf{v} in V is orthogonal to every \mathbf{w} in W (combinations of the basis vectors).
- 14** $A\mathbf{x} = B\hat{\mathbf{x}}$ means that $\begin{bmatrix} A & B \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ -\hat{\mathbf{x}} \end{bmatrix} = \mathbf{0}$. Three homogeneous equations in four unknowns always have a nonzero solution. Here $\mathbf{x} = (3, 1)$ and $\hat{\mathbf{x}} = (1, 0)$ and $A\mathbf{x} = B\hat{\mathbf{x}} = (5, 6, 5)$ is in both column spaces. Two planes in \mathbf{R}^3 must share a line.
- 15** A p -dimensional and a q -dimensional subspace of \mathbf{R}^n share at least a line if $p + q > n$. (The $p + q$ basis vectors of V and W cannot be independent.)
- 16** $A^T \mathbf{y} = \mathbf{0}$ leads to $(A\mathbf{x})^T \mathbf{y} = \mathbf{x}^T A^T \mathbf{y} = 0$. Then $\mathbf{y} \perp A\mathbf{x}$ and $N(A^T) \perp C(A)$.
- 17** If S is the subspace of \mathbf{R}^3 containing only the zero vector, then S^\perp is \mathbf{R}^3 . If S is spanned by $(1, 1, 1)$, then S^\perp is the plane spanned by $(1, -1, 0)$ and $(1, 0, -1)$. If S is spanned by $(2, 0, 0)$ and $(0, 0, 3)$, then S^\perp is the line spanned by $(0, 1, 0)$.
- 18** S^\perp is the nullspace of $A = \begin{bmatrix} 1 & 5 & 1 \\ 2 & 2 & 2 \end{bmatrix}$. Therefore S^\perp is a *subspace* even if S is not.
- 19** L^\perp is the 2-dimensional subspace (a plane) in \mathbf{R}^3 perpendicular to L . Then $(L^\perp)^\perp$ is a 1-dimensional subspace (a line) perpendicular to L^\perp . In fact $(L^\perp)^\perp$ is L .
- 20** If V is the whole space \mathbf{R}^4 , then V^\perp contains only the zero vector. Then $(V^\perp)^\perp = \mathbf{R}^4 = V$.
- 21** For example $(-5, 0, 1, 1)$ and $(0, 1, -1, 0)$ span $S^\perp = \text{nullspace of } A = \begin{bmatrix} 1 & 2 & 2 & 3 \\ 1 & 3 & 3 & 2 \end{bmatrix}$.
- 22** $(1, 1, 1, 1)$ is a basis for P^\perp . $A = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$ has P as its nullspace and P^\perp as row space.
- 23** \mathbf{x} in V^\perp is perpendicular to any vector in V . Since V contains all the vectors in S , \mathbf{x} is also perpendicular to any vector in S . So every \mathbf{x} in V^\perp is also in S^\perp .

- 24 $AA^{-1} = I$: Column 1 of A^{-1} is orthogonal to the space spanned by the 2nd, 3rd, ..., n th rows of A .
- 25 If the columns of A are unit vectors, all mutually perpendicular, then $A^T A = I$.
- 26 $A = \begin{bmatrix} 2 & 2 & -1 \\ -1 & 2 & 2 \\ 2 & -1 & 2 \end{bmatrix}$, This example shows a matrix with perpendicular columns. $A^T A = 9I$ is *diagonal*: $(A^T A)_{ij} = (\text{column } i \text{ of } A) \cdot (\text{column } j \text{ of } A)$. When the columns are *unit vectors*, then $A^T A = I$.
- 27 The lines $3x + y = b_1$ and $6x + 2y = b_2$ are **parallel**. They are the same line if $b_2 = 2b_1$. In that case (b_1, b_2) is perpendicular to $(-2, 1)$. The nullspace of the 2 by 2 matrix is the line $3x + y = 0$. One particular vector in the nullspace is $(-1, 3)$.
- 28 (a) $(1, -1, 0)$ is in both planes. Normal vectors are perpendicular, but planes still intersect! (b) Need *three* orthogonal vectors to span the whole orthogonal complement. (c) Lines can meet at the zero vector without being orthogonal.
- 29 $A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 1 & -1 \\ 2 & -1 & 0 \\ 3 & 0 & -1 \end{bmatrix}$; A has $\mathbf{v} = (1, 2, 3)$ in row space and column space. B has \mathbf{v} in its column space and nullspace. \mathbf{v} **can not** be in the nullspace and row space, or in the left nullspace and column space. These spaces are orthogonal and $\mathbf{v}^T \mathbf{v} \neq 0$.
- 30 When $AB = 0$, the column space of B is contained in the nullspace of A . Therefore the dimension of $C(B) \leq \text{dimension of } N(A)$. This means $\text{rank}(B) \leq 4 - \text{rank}(A)$.
- 31 $\text{null}(N')$ produces a basis for the *row space* of A (perpendicular to $N(A)$).
- 32 We need $\mathbf{r}^T \mathbf{n} = 0$ and $\mathbf{c}^T \boldsymbol{\ell} = 0$. All possible examples have the form $a\mathbf{c}\mathbf{r}^T$ with $a \neq 0$.
- 33 Both \mathbf{r} 's orthogonal to both \mathbf{n} 's, both \mathbf{c} 's orthogonal to both $\boldsymbol{\ell}$'s, each pair independent. All A 's with these subspaces have the form $[\mathbf{c}_1 \ \mathbf{c}_2]M[\mathbf{r}_1 \ \mathbf{r}_2]^T$ for a 2 by 2 invertible M .

Problem Set 4.2, page 214

- 1 (a) $\mathbf{a}^T \mathbf{b} / \mathbf{a}^T \mathbf{a} = 5/3$; $\mathbf{p} = 5\mathbf{a}/3$; $\mathbf{e} = (-2, 1, 1)/3$ (b) $\mathbf{a}^T \mathbf{b} / \mathbf{a}^T \mathbf{a} = -1$; $\mathbf{p} = \mathbf{a}$; $\mathbf{e} = \mathbf{0}$.
- 2 (a) The projection of $\mathbf{b} = (\cos \theta, \sin \theta)$ onto $\mathbf{a} = (1, 0)$ is $\mathbf{p} = (\cos \theta, 0)$
 (b) The projection of $\mathbf{b} = (1, 1)$ onto $\mathbf{a} = (1, -1)$ is $\mathbf{p} = (0, 0)$ since $\mathbf{a}^T \mathbf{b} = 0$.
- 3 $P_1 = \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ and $P_1 \mathbf{b} = \frac{1}{3} \begin{bmatrix} 5 \\ 5 \\ 5 \end{bmatrix}$. $P_2 = \frac{1}{11} \begin{bmatrix} 1 & 3 & 1 \\ 3 & 9 & 3 \\ 1 & 3 & 1 \end{bmatrix}$ and $P_2 \mathbf{b} = \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$.
- 4 $P_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $P_2 = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$. P_1 projects onto $(1, 0)$, P_2 projects onto $(1, -1)$
 $P_1 P_2 \neq 0$ and $P_1 + P_2$ is not a projection matrix.
- 5 $P_1 = \frac{1}{9} \begin{bmatrix} 1 & -2 & -2 \\ -2 & 4 & 4 \\ -2 & 4 & 4 \end{bmatrix}$, $P_2 = \frac{1}{9} \begin{bmatrix} 4 & 4 & -2 \\ 4 & 4 & -2 \\ -2 & -2 & 1 \end{bmatrix}$. P_1 and P_2 are the projection matrices onto the lines through $\mathbf{a}_1 = (-1, 2, 2)$ and $\mathbf{a}_2 = (2, 2, -1)$ $P_1 P_2 = \text{zero matrix}$ because $\mathbf{a}_1 \perp \mathbf{a}_2$.
- XXX Above solution does not fit in 3 lines.
- 6 $\mathbf{p}_1 = (\frac{1}{9}, -\frac{2}{9}, -\frac{2}{9})$ and $\mathbf{p}_2 = (\frac{4}{9}, \frac{4}{9}, -\frac{2}{9})$ and $\mathbf{p}_3 = (\frac{4}{9}, -\frac{2}{9}, \frac{4}{9})$. So $\mathbf{p}_1 + \mathbf{p}_2 + \mathbf{p}_3 = \mathbf{b}$.

$$7 \quad P_1 + P_2 + P_3 = \frac{1}{9} \begin{bmatrix} 1 & -2 & -2 \\ -2 & 4 & 4 \\ -2 & 4 & 4 \end{bmatrix} + \frac{1}{9} \begin{bmatrix} 4 & 4 & -2 \\ 4 & 4 & -2 \\ -2 & -2 & 1 \end{bmatrix} + \frac{1}{9} \begin{bmatrix} 4 & -2 & 4 \\ -2 & 1 & -2 \\ 4 & -2 & 4 \end{bmatrix} = I.$$

We can add projections onto *orthogonal* vectors. This is important.

8 The projections of $(1, 1)$ onto $(1, 0)$ and $(1, 2)$ are $\mathbf{p}_1 = (1, 0)$ and $\mathbf{p}_2 = (0.6, 1.2)$. Then $\mathbf{p}_1 + \mathbf{p}_2 \neq \mathbf{b}$.

9 Since A is invertible, $P = A(A^T A)^{-1} A^T = A A^{-1} (A^T)^{-1} A^T = I$: project on all of \mathbf{R}^2 .

10 $P_2 = \begin{bmatrix} 0.2 & 0.4 \\ 0.4 & 0.8 \end{bmatrix}$, $P_2 \mathbf{a}_1 = \begin{bmatrix} 0.2 \\ 0.4 \end{bmatrix}$, $P_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $P_1 P_2 \mathbf{a}_1 = \begin{bmatrix} 0.2 \\ 0 \end{bmatrix}$. This is not $\mathbf{a}_1 = (1, 0)$. No, $P_1 P_2 \neq (P_1 P_2)^2$.

11 (a) $\mathbf{p} = A(A^T A)^{-1} A^T \mathbf{b} = (2, 3, 0)$, $\mathbf{e} = (0, 0, 4)$, $A^T \mathbf{e} = \mathbf{0}$ (b) $\mathbf{p} = (4, 4, 6)$, $\mathbf{e} = \mathbf{0}$.

12 $P_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ = projection matrix onto the column space of A (the xy plane)

$P_2 = \begin{bmatrix} 0.5 & 0.5 & 0 \\ 0.5 & 0.5 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ = Projection matrix onto the second column space. Certainly $(P_2)^2 = P_2$.

13 $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$, P = square matrix = $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$, $\mathbf{p} = P \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 0 \end{bmatrix}$.

14 The projection of this \mathbf{b} onto the column space of A is \mathbf{b} itself when \mathbf{b} is in that space.

But P is not necessarily I . $P = \frac{1}{21} \begin{bmatrix} 5 & 8 & -4 \\ 8 & 17 & 2 \\ -4 & 2 & 20 \end{bmatrix}$ and $\mathbf{b} = P\mathbf{b} = \mathbf{p} = \begin{bmatrix} 0 \\ 2 \\ 4 \end{bmatrix}$.

15 $2A$ has the same column space as A . $\hat{\mathbf{x}}$ for $2A$ is *half* of $\hat{\mathbf{x}}$ for A .

16 $\frac{1}{2}(1, 2, -1) + \frac{3}{2}(1, 0, 1) = (2, 1, 1)$. So \mathbf{b} is in the plane. Projection shows $P\mathbf{b} = \mathbf{b}$.

17 If $P^2 = P$ then $(I - P)^2 = (I - P)(I - P) = I - PI - IP + P^2 = I - P$. When P projects onto the column space, $I - P$ projects onto the *left nullspace*.

18 (a) $I - P$ is the projection matrix onto $(1, -1)$ in the perpendicular direction to $(1, 1)$
(b) $I - P$ projects onto the plane $x + y + z = 0$ perpendicular to $(1, 1, 1)$.

19 For any basis vectors in the plane $x - y - 2z = 0$, say $(1, 1, 0)$ and $(2, 0, 1)$, the matrix P is $\begin{bmatrix} 5/6 & 1/6 & 1/3 \\ 1/6 & 5/6 & -1/3 \\ 1/3 & -1/3 & 1/3 \end{bmatrix}$.

20 $\mathbf{e} = \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}$, $Q = \frac{\mathbf{e}\mathbf{e}^T}{\mathbf{e}^T \mathbf{e}} = \begin{bmatrix} 1/6 & -1/6 & -1/3 \\ -1/6 & 1/6 & 1/3 \\ -1/3 & 1/3 & 2/3 \end{bmatrix}$, $I - Q = \begin{bmatrix} 5/6 & 1/6 & 1/3 \\ 1/6 & 5/6 & -1/3 \\ 1/3 & -1/3 & 1/3 \end{bmatrix}$.

21 $(A(A^T A)^{-1} A^T)^2 = A(A^T A)^{-1} (A^T A) (A^T A)^{-1} A^T = A(A^T A)^{-1} A^T$. So $P^2 = P$. $P\mathbf{b}$ is in the column space (where P projects). Then its projection $P(P\mathbf{b})$ is $P\mathbf{b}$.

22 $P^T = (A(A^T A)^{-1} A^T)^T = A((A^T A)^{-1})^T A^T = A(A^T A)^{-1} A^T = P$. ($A^T A$ is symmetric!)

23 If A is invertible then its column space is all of \mathbf{R}^n . So $P = I$ and $\mathbf{e} = \mathbf{0}$.

24 The nullspace of A^T is *orthogonal* to the column space $C(A)$. So if $A^T \mathbf{b} = \mathbf{0}$, the projection of \mathbf{b} onto $C(A)$ should be $\mathbf{p} = \mathbf{0}$. Check $P\mathbf{b} = A(A^T A)^{-1} A^T \mathbf{b} = A(A^T A)^{-1} \mathbf{0}$.

- 25** The column space of P will be S . Then $r = \text{dimension of } S = n$.
- 26** A^{-1} exists since the rank is $r = m$. Multiply $A^2 = A$ by A^{-1} to get $A = I$.
- 27** If $A^T A \mathbf{x} = \mathbf{0}$ then $A \mathbf{x}$ is in the nullspace of A^T . But $A \mathbf{x}$ is always in the column space of A . To be in both of those perpendicular spaces, $A \mathbf{x}$ must be zero. So A and $A^T A$ have the *same nullspace*.
- 28** $P^2 = P = P^T$ give $P^T P = P$. Then the $(2, 2)$ entry of P equals the $(2, 2)$ entry of $P^T P$ which is the length squared of column 2.
- 29** $A = B^T$ has independent columns, so $A^T A$ (which is BB^T) must be invertible.
- 30** (a) The column space is the line through $\mathbf{a} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$ so $P_C = \frac{\mathbf{a}\mathbf{a}^T}{\mathbf{a}^T \mathbf{a}} = \frac{1}{25} \begin{bmatrix} 9 & 12 \\ 12 & 25 \end{bmatrix}$.
 (b) The row space is the line through $\mathbf{v} = (1, 2, 2)$ and $P_R = \mathbf{v}\mathbf{v}^T / \mathbf{v}^T \mathbf{v}$. Always $P_C A = A$ (columns of A project to themselves) and $A P_R = A$. Then $P_C A P_R = A$!
- 31** The error $\mathbf{e} = \mathbf{b} - \mathbf{p}$ must be perpendicular to all the \mathbf{a} 's.
- 32** Since $P_1 \mathbf{b}$ is in $C(A)$, $P_2(P_1 \mathbf{b})$ equals $P_1 \mathbf{b}$. So $P_2 P_1 = P_1 = \mathbf{a}\mathbf{a}^T / \mathbf{a}^T \mathbf{a}$ where $\mathbf{a} = (1, 2, 0)$.
- 33** If $P_1 P_2 = P_2 P_1$ then S is contained in T or T is contained in S .
- 34** BB^T is invertible as in Problem 29. Then $(A^T A)(BB^T) = \text{product of } r \text{ by } r \text{ invertible matrices, so rank } r$. AB can't have rank $< r$, since A^T and B^T cannot increase the rank.
Conclusion: A (m by r of rank r) times B (r by n of rank r) produces AB of rank r .

Problem Set 4.3, page 226

- 1** $A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$ and $\mathbf{b} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$ give $A^T A = \begin{bmatrix} 4 & 8 \\ 8 & 26 \end{bmatrix}$ and $A^T \mathbf{b} = \begin{bmatrix} 36 \\ 112 \end{bmatrix}$.
- $A^T A \hat{\mathbf{x}} = A^T \mathbf{b}$ gives $\hat{\mathbf{x}} = \begin{bmatrix} 1 \\ 4 \end{bmatrix}$ and $\mathbf{p} = A \hat{\mathbf{x}} = \begin{bmatrix} 1 \\ 5 \\ 13 \\ 17 \end{bmatrix}$ and $\mathbf{e} = \mathbf{b} - \mathbf{p} = \begin{bmatrix} -1 \\ 3 \\ -5 \\ 3 \end{bmatrix}$
 $E = \|\mathbf{e}\|^2 = 44$
- 2** $\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$. This $A\mathbf{x} = \mathbf{b}$ is unsolvable. Change \mathbf{b} to $\mathbf{p} = P\mathbf{b} = \begin{bmatrix} 1 \\ 5 \\ 13 \\ 17 \end{bmatrix}$; $\hat{\mathbf{x}} = \begin{bmatrix} 1 \\ 4 \end{bmatrix}$ exactly solves $A\hat{\mathbf{x}} = \mathbf{p}$.
- 3** In Problem 2, $\mathbf{p} = A(A^T A)^{-1} A^T \mathbf{b} = (1, 5, 13, 17)$ and $\mathbf{e} = \mathbf{b} - \mathbf{p} = (-1, 3, -5, 3)$. \mathbf{e} is perpendicular to both columns of A . This shortest distance $\|\mathbf{e}\|$ is $\sqrt{44}$.
- 4** $E = (C + 0D)^2 + (C + 1D - 8)^2 + (C + 3D - 8)^2 + (C + 4D - 20)^2$. Then $\partial E / \partial C = 2C + 2(C + D - 8) + 2(C + 3D - 8) + 2(C + 4D - 20) = 0$ and $\partial E / \partial D = 1 \cdot 2(C + D - 8) + 3 \cdot 2(C + 3D - 8) + 4 \cdot 2(C + 4D - 20) = 0$. These normal equations are again $\begin{bmatrix} 4 & 8 \\ 8 & 26 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 36 \\ 112 \end{bmatrix}$.

- 5 $E = (C-0)^2 + (C-8)^2 + (C-8)^2 + (C-20)^2$. $A^T = [1 \ 1 \ 1 \ 1]$ and $A^T A = [4]$. $A^T \mathbf{b} = [36]$ and $(A^T A)^{-1} A^T \mathbf{b} = 9 = \text{best height } C$. Errors $\mathbf{e} = (-9, -1, -1, 11)$.
- 6 $\mathbf{a} = (1, 1, 1, 1)$ and $\mathbf{b} = (0, 8, 8, 20)$ give $\hat{x} = \mathbf{a}^T \mathbf{b} / \mathbf{a}^T \mathbf{a} = 9$ and the projection is $\hat{\mathbf{x}} \mathbf{a} = \mathbf{p} = (9, 9, 9, 9)$. Then $\mathbf{e}^T \mathbf{a} = (-9, -1, -1, 11)^T (1, 1, 1, 1) = 0$ and $\|\mathbf{e}\| = \sqrt{204}$.
- 7 $A = [0 \ 1 \ 3 \ 4]^T$, $A^T A = [26]$ and $A^T \mathbf{b} = [112]$. Best $D = 112/26 = 56/13$.
- 8 $\hat{\mathbf{x}} = 56/13$, $\mathbf{p} = (56/13)(0, 1, 3, 4)$. $(C, D) = (9, 56/13)$ don't match $(C, D) = (1, 4)$. Columns of A were not perpendicular so we can't project separately to find C and D .
- 9 Parabola
Project \mathbf{b}
4D to 3D $\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 3 & 9 \\ 1 & 4 & 16 \end{bmatrix} \begin{bmatrix} C \\ D \\ E \end{bmatrix} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$. $A^T A \hat{\mathbf{x}} = \begin{bmatrix} 4 & 8 & 26 \\ 8 & 26 & 92 \\ 26 & 92 & 338 \end{bmatrix} \begin{bmatrix} C \\ D \\ E \end{bmatrix} = \begin{bmatrix} 36 \\ 112 \\ 400 \end{bmatrix}$.
- 10 $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} \begin{bmatrix} C \\ D \\ E \\ F \end{bmatrix} = \begin{bmatrix} 0 \\ 8 \\ 8 \\ 20 \end{bmatrix}$. Then $\begin{bmatrix} C \\ D \\ E \\ F \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 0 \\ 47 \\ -28 \\ 5 \end{bmatrix}$. Exact cubic so $\mathbf{p} = \mathbf{b}$, $\mathbf{e} = \mathbf{0}$. This Vandermonde matrix gives exact interpolation by a cubic at 0, 1, 3, 4.
- 11 (a) The best line $x = 1 + 4t$ gives the center point $\hat{\mathbf{b}} = 9$ when $\hat{t} = 2$.
(b) The first equation $Cm + D \sum t_i = \sum b_i$ divided by m gives $C + D\hat{t} = \hat{\mathbf{b}}$.
- 12 (a) $\mathbf{a} = (1, \dots, 1)$ has $\mathbf{a}^T \mathbf{a} = m$, $\mathbf{a}^T \mathbf{b} = b_1 + \dots + b_m$. Therefore $\hat{x} = \mathbf{a}^T \mathbf{b} / m$ is the mean of the b 's. (b) $\mathbf{e} = \mathbf{b} - \hat{x} \mathbf{a}$, $\mathbf{b} = (1, 2, b)$, $\|\mathbf{e}\|^2 = \sum_{i=1}^m (b_i - \hat{x})^2 = \text{variance}$.
(c) $\mathbf{p} = (3, 3, 3)$, $\mathbf{e} = (-2, -1, 3)$, $\mathbf{p}^T \mathbf{e} = 0$. $P = \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$.
- 13 $(A^T A)^{-1} A^T (\mathbf{b} - A\mathbf{x}) = \hat{\mathbf{x}} - \mathbf{x}$. When $\mathbf{e} = \mathbf{b} - A\mathbf{x}$ averages to $\mathbf{0}$, so does $\hat{\mathbf{x}} - \mathbf{x}$.
- 14 The matrix $(\hat{\mathbf{x}} - \mathbf{x})(\hat{\mathbf{x}} - \mathbf{x})^T$ is $(A^T A)^{-1} A^T (\mathbf{b} - A\mathbf{x})(\mathbf{b} - A\mathbf{x})^T A (A^T A)^{-1}$. When the average of $(\mathbf{b} - A\mathbf{x})(\mathbf{b} - A\mathbf{x})^T$ is $\sigma^2 I$, the average of $(\hat{\mathbf{x}} - \mathbf{x})(\hat{\mathbf{x}} - \mathbf{x})^T$ will be the output covariance matrix $(A^T A)^{-1} A^T \sigma^2 A (A^T A)^{-1}$ which simplifies to $\sigma^2 (A^T A)^{-1}$.
- 15 When A has 1 column of ones, Problem 14 gives the expected error $(\hat{x} - x)^2$ as $\sigma^2 (A^T A)^{-1} = \sigma^2 / m$. By taking m measurements, the variance drops from σ^2 to σ^2 / m .
- 16 $\frac{1}{10} b_{10} + \frac{9}{10} \hat{x}_9 = \frac{1}{10} (b_1 + \dots + b_{10})$. Knowing \hat{x}_9 avoids adding all b 's.
- 17 $\begin{bmatrix} 1 & -1 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 7 \\ 7 \\ 21 \end{bmatrix}$. The solution $\hat{\mathbf{x}} = \begin{bmatrix} 9 \\ 4 \end{bmatrix}$ comes from $\begin{bmatrix} 3 & 2 \\ 2 & 6 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 35 \\ 42 \end{bmatrix}$.
- 18 $\mathbf{p} = A\hat{\mathbf{x}} = (5, 13, 17)$ gives the heights of the closest line. The error is $\mathbf{b} - \mathbf{p} = (2, -6, 4)$. This error \mathbf{e} has $P\mathbf{e} = P\mathbf{b} - P\mathbf{p} = \mathbf{p} - \mathbf{p} = \mathbf{0}$.
- 19 If $\mathbf{b} = \text{error } \mathbf{e}$ then \mathbf{b} is perpendicular to the column space of A . Projection $\mathbf{p} = \mathbf{0}$.
- 20 If $\mathbf{b} = A\hat{\mathbf{x}} = (5, 13, 17)$ then $\hat{\mathbf{x}} = (9, 4)$ and $\mathbf{e} = \mathbf{0}$ since \mathbf{b} is in the column space of A .
- 21 \mathbf{e} is in $N(A^T)$; \mathbf{p} is in $C(A)$; $\hat{\mathbf{x}}$ is in $C(A^T)$; $N(A) = \{\mathbf{0}\} = \text{zero vector only}$.

- 22** The least squares equation is $\begin{bmatrix} 5 & \mathbf{0} \\ \mathbf{0} & 10 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 5 \\ -10 \end{bmatrix}$. Solution: $C = 1, D = -1$.
Line $1 - t$. Symmetric t 's \Rightarrow diagonal $A^T A$
- 23** e is orthogonal to p ; then $\|e\|^2 = e^T(b - p) = e^T b = b^T b - b^T p$.
- 24** The derivatives of $\|Ax - b\|^2 = x^T A^T A x - 2b^T A x + b^T b$ (this term is constant) are zero when $2A^T A x = 2A^T b$, or $x = (A^T A)^{-1} A^T b$.
- 25** 3 points on a line: *Equal slopes* $(b_2 - b_1)/(t_2 - t_1) = (b_3 - b_2)/(t_3 - t_2)$. Linear algebra: Orthogonal to $(1, 1, 1)$ and (t_1, t_2, t_3) is $y = (t_2 - t_3, t_3 - t_1, t_1 - t_2)$ in the left nullspace. b is in the column space. Then $y^T b = 0$ is the same equal slopes condition written as $(b_2 - b_1)(t_3 - t_2) = (b_3 - b_2)(t_2 - t_1)$.
- 26** $\begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} C \\ D \\ E \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 3 \\ 4 \end{bmatrix}$ has $A^T A = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$, $A^T b = \begin{bmatrix} 8 \\ -2 \\ -3 \end{bmatrix}$, $\begin{bmatrix} C \\ D \\ E \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \\ -3/2 \end{bmatrix}$. At $x, y = 0, 0$ the best plane $2 - x - \frac{3}{2}y$ has height $C = 2 =$ average of $0, 1, 3, 4$.
- 27** The shortest link connecting two lines in space is *perpendicular to those lines*.
- 28** Only 1 plane contains $\mathbf{0}, a_1, a_2$ unless a_1, a_2 are *dependent*. Same test for a_1, \dots, a_n .
- 29** There is exactly one hyperplane containing the n points $\mathbf{0}, a_1, \dots, a_{n-1}$ *When the $n - 1$ vectors a_1, \dots, a_{n-1} are linearly independent*. (For $n = 3$, the vectors a_1 and a_2 must be independent. Then the three points $\mathbf{0}, a_1, a_2$ determine a plane.) The equation of the plane in \mathbf{R}^n will be $a_n^T x = 0$. Here a_n is any nonzero vector on the line (it is only a line!) perpendicular to a_1, \dots, a_{n-1} .

Problem Set 4.4, page 239

- 1** (a) *Independent* (b) *Independent and orthogonal* (c) *Independent and orthonormal*.
For orthonormal vectors, (a) becomes $(1, 0), (0, 1)$ and (b) is $(.6, .8), (.8, -.6)$.
- 2** Divide by length 3 to get $q_1 = (\frac{2}{3}, \frac{2}{3}, -\frac{1}{3})$. $q_2 = (-\frac{1}{3}, \frac{2}{3}, \frac{2}{3})$. $Q^T Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ but $Q Q^T = \begin{bmatrix} 5/9 & 2/9 & -4/9 \\ 2/9 & 8/9 & 2/9 \\ -4/9 & 2/9 & 5/9 \end{bmatrix}$.
- 3** (a) $A^T A$ will be $16I$ (b) $A^T A$ will be diagonal with entries 1, 4, 9.
- 4** (a) $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$, $Q Q^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \neq I$. Any Q with $n < m$ has $Q Q^T \neq I$. (b) $(1, 0)$ and $(0, 0)$ are *orthogonal*, not *independent*. Nonzero orthogonal vectors are independent. (c) Starting from $q_1 = (1, 1, 1)/\sqrt{3}$ my favorite is $q_2 = (1, -1, 0)/\sqrt{2}$ and $q_3 = (1, 1, -2)/\sqrt{6}$.
- 5** *Orthogonal* vectors are $(1, -1, 0)$ and $(1, 1, -1)$. *Orthonormal* are $(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0)$, $(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}})$.

- 6** $Q_1 Q_2$ is orthogonal because $(Q_1 Q_2)^T Q_1 Q_2 = Q_2^T Q_1^T Q_1 Q_2 = Q_2^T Q_2 = I$.
- 7** When Gram-Schmidt gives Q with orthonormal columns, $Q^T Q \hat{x} = Q^T b$ becomes $\hat{x} = Q^T b$.
- 8** If q_1 and q_2 are *orthonormal* vectors in \mathbf{R}^5 then $(q_1^T b)q_1 + (q_2^T b)q_2$ is closest to b .
- 9** (a) $Q = \begin{bmatrix} .8 & -.6 \\ .6 & .8 \\ 0 & 0 \end{bmatrix}$ has $P = Q Q^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ (b) $(Q Q^T)(Q Q^T) = Q(Q^T Q)Q^T = Q Q^T$.
- 10** (a) If q_1, q_2, q_3 are *orthonormal* then the dot product of q_1 with $c_1 q_1 + c_2 q_2 + c_3 q_3 = \mathbf{0}$ gives $c_1 = 0$. Similarly $c_2 = c_3 = 0$. *Independent* q 's (b) $Qx = \mathbf{0} \Rightarrow Q^T Qx = \mathbf{0} \Rightarrow x = \mathbf{0}$.
- 11** (a) Two *orthonormal* vectors are $q_1 = \frac{1}{10}(1, 3, 4, 5, 7)$ and $q_2 = \frac{1}{10}(-7, 3, 4, -5, 1)$
 (b) Closest in the plane: *project* $Q Q^T(1, 0, 0, 0, 0) = (0.5, -0.18, -0.24, 0.4, 0)$.
- 12** (a) Orthonormal a 's: $a_1^T b = a_1^T(x_1 a_1 + x_2 a_2 + x_3 a_3) = x_1(a_1^T a_1) = x_1$
 (b) Orthogonal a 's: $a_1^T b = a_1^T(x_1 a_1 + x_2 a_2 + x_3 a_3) = x_1(a_1^T a_1)$. Therefore $x_1 = a_1^T b / a_1^T a_1$
 (c) x_1 is the first component of A^{-1} times b .
- 13** The multiple to subtract is $\frac{a^T b}{a^T a}$. Then $B = b - \frac{a^T b}{a^T a} a = (4, 0) - 2 \cdot (1, 1) = (2, -2)$.
- 14** $\begin{bmatrix} 1 & 4 \\ 1 & 0 \end{bmatrix} = [q_1 \ q_2] \begin{bmatrix} \|a\| & q_1^T b \\ 0 & \|B\| \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{2} & 2\sqrt{2} \\ 0 & 2\sqrt{2} \end{bmatrix} = QR$.
- 15** (a) $q_1 = \frac{1}{3}(1, 2, -2)$, $q_2 = \frac{1}{3}(2, 1, 2)$, $q_3 = \frac{1}{3}(2, -2, -1)$ (b) The nullspace of A^T contains q_3 (c) $\hat{x} = (A^T A)^{-1} A^T(1, 2, 7) = (1, 2)$.
- 16** The projection $p = (a^T b / a^T a)a = 14a/49 = 2a/7$ is closest to b ; $q_1 = a/\|a\| = a/7$ is $(4, 5, 2, 2)/7$. $B = b - p = (-1, 4, -4, -4)/7$ has $\|B\| = 1$ so $q_2 = B$.
- 17** $p = (a^T b / a^T a)a = (3, 3, 3)$ and $e = (-2, 0, 2)$. $q_1 = (1, 1, 1)/\sqrt{3}$ and $q_2 = (-1, 0, 1)/\sqrt{2}$.
- 18** $A = a = (1, -1, 0, 0)$; $B = b - p = (\frac{1}{2}, \frac{1}{2}, -1, 0)$; $C = c - p_A - p_B = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3}, -1)$. Notice the pattern in those orthogonal A, B, C . In \mathbf{R}^5 , D would be $(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}, -1)$.
- 19** If $A = QR$ then $A^T A = R^T Q^T Q R = R^T R =$ lower triangular times upper triangular (this Cholesky factorization of $A^T A$ uses the same R as Gram-Schmidt!). The example has $A = \begin{bmatrix} -1 & 1 \\ 2 & 1 \\ 2 & 4 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} -1 & 2 \\ 2 & -1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 3 & 3 \\ 0 & 3 \end{bmatrix} = QR$ and the same R appears in $A^T A = \begin{bmatrix} 9 & 9 \\ 9 & 18 \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} 3 & 3 \\ 0 & 3 \end{bmatrix} = R^T R$.
- 20** (a) *True* (b) *True*. $Qx = x_1 q_1 + x_2 q_2$. $\|Qx\|^2 = x_1^2 + x_2^2$ because $q_1 \cdot q_2 = 0$.
- 21** The orthonormal vectors are $q_1 = (1, 1, 1, 1)/2$ and $q_2 = (-5, -1, 1, 5)/\sqrt{52}$. Then $b = (-4, -3, 3, 0)$ projects to $p = (-7, -3, -1, 3)/2$. And $b - p = (-1, -3, 7, -3)/2$ is orthogonal to both q_1 and q_2 .
- 22** $A = (1, 1, 2)$, $B = (1, -1, 0)$, $C = (-1, -1, 1)$. These are not yet unit vectors.

23 You can see why $\mathbf{q}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $\mathbf{q}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$, $\mathbf{q}_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$. $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & 4 \\ 0 & 3 & 6 \\ 0 & 0 & 5 \end{bmatrix} = QR$.

24 (a) One basis for the subspace S of solutions to $x_1 + x_2 + x_3 - x_4 = 0$ is $\mathbf{v}_1 = (1, -1, 0, 0)$, $\mathbf{v}_2 = (1, 0, -1, 0)$, $\mathbf{v}_3 = (1, 0, 0, 1)$ (b) Since S contains solutions to $(1, 1, 1, -1)^T \mathbf{x} = 0$, a basis for S^\perp is $(1, 1, 1, -1)$ (c) Split $(1, 1, 1, 1) = \mathbf{b}_1 + \mathbf{b}_2$ by projection on S^\perp and S : $\mathbf{b}_2 = (\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, -\frac{1}{2})$ and $\mathbf{b}_1 = (\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{3}{2})$.

25 This question shows 2 by 2 formulas for QR ; breakdown $R_{22} = 0$ when A is singular. $\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix} \cdot \frac{1}{\sqrt{5}} \begin{bmatrix} 5 & 3 \\ 0 & 1 \end{bmatrix}$. Singular $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \cdot \frac{1}{\sqrt{2}} \begin{bmatrix} 2 & 2 \\ 0 & 0 \end{bmatrix}$. The Gram-Schmidt process breaks down when $ad - bc = 0$.

26 $(\mathbf{q}_2^T \mathbf{C}^*) \mathbf{q}_2 = \frac{\mathbf{B}^T \mathbf{c}}{\mathbf{B}^T \mathbf{B}} \mathbf{B}$ because $\mathbf{q}_2 = \frac{\mathbf{B}}{\|\mathbf{B}\|}$ and the extra \mathbf{q}_1 in \mathbf{C}^* is orthogonal to \mathbf{q}_2 .

27 When a and b are not orthogonal, the projections onto these lines *do not add* to the projection onto the plane of a and b . We must use the orthogonal A and B (or orthonormal \mathbf{q}_1 and \mathbf{q}_2) to be allowed to add 1D projections.

28 There are mn multiplications in (11) and $\frac{1}{2}m^2n$ multiplications in each part of (12).

29 $\mathbf{q}_1 = \frac{1}{3}(2, 2, -1)$, $\mathbf{q}_2 = \frac{1}{3}(2, -1, 2)$, $\mathbf{q}_3 = \frac{1}{3}(1, -2, -2)$.

30 The columns of the wavelet matrix W are *orthonormal*. Then $W^{-1} = W^T$. See Section 7.2 for more about wavelets: a useful orthonormal basis with many zeros.

31 (a) $c = \frac{1}{2}$ normalizes all the orthogonal columns to have unit length (b) The projection $(\mathbf{a}^T \mathbf{b} / \mathbf{a}^T \mathbf{a}) \mathbf{a}$ of $\mathbf{b} = (1, 1, 1, 1)$ onto the first column is $\mathbf{p}_1 = \frac{1}{2}(-1, 1, 1, 1)$. (Check $\mathbf{e} = \mathbf{0}$.) To project onto the plane, add $\mathbf{p}_2 = \frac{1}{2}(1, -1, 1, 1)$ to get $(0, 0, 1, 1)$.

32 $Q_1 = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ reflects across x axis, $Q_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & -1 & 0 \end{bmatrix}$ across plane $y + z = 0$.

33 Orthogonal and lower triangular $\Rightarrow \pm 1$ on the main diagonal and zeros elsewhere.

34 (a) $Q\mathbf{u} = (I - 2\mathbf{u}\mathbf{u}^T)\mathbf{u} = \mathbf{u} - 2\mathbf{u}\mathbf{u}^T\mathbf{u}$. This is $-\mathbf{u}$, provided that $\mathbf{u}^T\mathbf{u}$ equals 1 (b) $Q\mathbf{v} = (I - 2\mathbf{u}\mathbf{u}^T)\mathbf{v} = \mathbf{v} - 2\mathbf{u}\mathbf{u}^T\mathbf{v} = \mathbf{v}$, provided that $\mathbf{u}^T\mathbf{v} = 0$.

35 Starting from $\mathbf{A} = (1, -1, 0, 0)$, the orthogonal (not orthonormal) vectors $\mathbf{B} = (1, 1, -2, 0)$ and $\mathbf{C} = (1, 1, 1, -3)$ and $\mathbf{D} = (1, 1, 1, 1)$ are in the directions of $\mathbf{q}_2, \mathbf{q}_3, \mathbf{q}_4$. The 4 by 4 and 5 by 5 matrices with *integer orthogonal columns* (not orthogonal rows,

since not orthonormal Q !) are $\begin{bmatrix} \mathbf{A} & \mathbf{B} & \mathbf{C} & \mathbf{D} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 1 & 1 & 1 \\ 0 & -2 & 1 & 1 \\ 0 & 0 & -3 & 1 \end{bmatrix}$ and

$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ -1 & 1 & 1 & 1 & 1 \\ 0 & -2 & 1 & 1 & 1 \\ 0 & 0 & -3 & 1 & 1 \\ 0 & 0 & 0 & -4 & 1 \end{bmatrix}$$

- 36** $[Q, R] = \mathbf{q} \mathbf{r}(A)$ produces from A (m by n of rank n) a “full-size” square $Q = [Q_1 \ Q_2]$ and $\begin{bmatrix} R \\ 0 \end{bmatrix}$. The columns of Q_1 are the orthonormal basis from Gram-Schmidt of the column space of A . The $m - n$ columns of Q_2 are an orthonormal basis for the left nullspace of A . Together the columns of $Q = [Q_1 \ Q_2]$ are an orthonormal basis for \mathbf{R}^m .
- 37** This question describes the next \mathbf{q}_{n+1} in Gram-Schmidt using the matrix Q with the columns $\mathbf{q}_1, \dots, \mathbf{q}_n$ (instead of using those \mathbf{q} ’s separately). Start from \mathbf{a} , subtract its projection $\mathbf{p} = Q^T \mathbf{a}$ onto the earlier \mathbf{q} ’s, divide by the length of $\mathbf{e} = \mathbf{a} - Q^T \mathbf{a}$ to get $\mathbf{q}_{n+1} = \mathbf{e} / \|\mathbf{e}\|$.

Problem Set 5.1, page 251

- 1** $\det(2A) = 8$; $\det(-A) = (-1)^4 \det A = \frac{1}{2}$; $\det(A^2) = \frac{1}{4}$; $\det(A^{-1}) = 2 = \det(A^T)^{-1}$.
- 2** $\det(\frac{1}{2}A) = (\frac{1}{2})^3 \det A = -\frac{1}{8}$ and $\det(-A) = (-1)^3 \det A = 1$; $\det(A^2) = 1$; $\det(A^{-1}) = -1$.
- 3** (a) *False*: $\det(I + I)$ is not $1 + 1$ (b) *True*: The product rule extends to ABC (use it twice) (c) *False*: $\det(4A)$ is $4^n \det A$ (d) *False*: $A = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$, $B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$, $AB - BA = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ is invertible.
- 4** Exchange rows 1 and 3 to show $|J_3| = -1$. Exchange rows 1 and 4, then 2 and 3 to show $|J_4| = 1$.
- 5** $|J_5| = 1$, $|J_6| = -1$, $|J_7| = -1$. Determinants 1, 1, -1, -1 repeat so $|J_{101}| = 1$.
- 6** To prove Rule 6, multiply the zero row by $t = 2$. The determinant is multiplied by 2 (Rule 3) but the matrix is the same. So $2 \det(A) = \det(A)$ and $\det(A) = 0$.
- 7** $\det(Q) = 1$ for rotation and $\det(Q) = -1$ for reflection ($1 - 2 \sin^2 \theta - 2 \cos^2 \theta = -1$).
- 8** $Q^T Q = I \Rightarrow |Q|^2 = 1 \Rightarrow |Q| = \pm 1$; Q^n stays orthogonal so \det can’t blow up.
- 9** $\det A = 1$ from two row exchanges. $\det B = 2$ (subtract rows 1 and 2 from row 3, then columns 1 and 2 from column 3). $\det C = 0$ (equal rows) even though $C = A + B$!
- 10** If the entries in every row add to zero, then $(1, 1, \dots, 1)$ is in the nullspace: singular A has $\det = 0$. (The columns add to the zero column so they are linearly dependent.) If every row adds to one, then rows of $A - I$ add to zero (not necessarily $\det A = 1$).
- 11** $CD = -DC \Rightarrow \det CD = (-1)^n \det DC$ and *not* $-\det DC$. If n is even we can have an invertible CD .
- 12** $\det(A^{-1})$ divides twice by $ad - bc$ (once for each row). This gives $\frac{ad-bc}{(ad-bc)^2} = \frac{1}{ad-bc}$.
- 13** Pivots 1, 1, 1 give determinant = 1; pivots 1, -2, -3/2 give determinant = 3.
- 14** $\det(A) = 36$ and the 4 by 4 second difference matrix has $\det = 5$.
- 15** The first determinant is 0, the second is $1 - 2t^2 + t^4 = (1 - t^2)^2$.

- 16** A singular rank one matrix has determinant = 0. The skew-symmetric K also $\det K = 0$ (see #17).
- 17** Any 3 by 3 skew-symmetric K has $\det(K^T) = \det(-K) = (-1)^3 \det(K)$. This is $-\det(K)$. But always $\det(K^T) = \det(K)$. So we must have $\det(K) = 0$ for 3 by 3.
- 18**
$$\begin{vmatrix} 1 & a & a^2 \\ 1 & b & b^2 \\ 1 & c & c^2 \end{vmatrix} = \begin{vmatrix} 1 & a & a^2 \\ 0 & b-a & b^2-a^2 \\ 0 & c-a & c^2-a^2 \end{vmatrix} = \begin{vmatrix} b-a & b^2-a^2 \\ c-a & c^2-a^2 \end{vmatrix} \quad (\text{to reach 2 by 2, eliminate } a \text{ and } a^2 \text{ in row 1 by column operations}).$$
 Factor out $b-a$ and $c-a$ from the 2 by 2: $(b-a)(c-a) \begin{vmatrix} 1 & b+a \\ 1 & c+a \end{vmatrix} = (b-a)(c-a)(c-b)$.
- 19** For triangular matrices, just multiply the diagonal entries: $\det(U) = 6$, $\det(U^{-1}) = \frac{1}{6}$, and $\det(U^2) = 36$. 2 by 2 matrix: $\det(U) = ad$, $\det(U^2) = a^2 d^2$. If $ad \neq 0$ then $\det(U^{-1}) = 1/ad$.
- 20** $\det \begin{bmatrix} a-Lc & b-Ld \\ c-\ell a & d-\ell b \end{bmatrix}$ reduces to $(ad-bc)(1-L\ell)$. The determinant changes if you do two row operations at once.
- 21** Rules 5 and 3 give Rule 2. (Since Rules 4 and 3 give 5, they also give Rule 2.)
- 22** $\det(A) = 3$, $\det(A^{-1}) = \frac{1}{3}$, $\det(A - \lambda I) = \lambda^2 - 4\lambda + 3$. The numbers $\lambda = 1$ and $\lambda = 3$ give $\det(A - \lambda I) = 0$. *Note to instructor:* If you discuss this exercise, you can explain that this is the reason determinants come before eigenvalues. Identify $\lambda = 1$ and $\lambda = 3$ as the eigenvalues of A .
- 23** $\det(A) = 10$, $A^2 = \begin{bmatrix} 18 & 7 \\ 14 & 11 \end{bmatrix}$, $\det(A^2) = 100$, $A^{-1} = \frac{1}{10} \begin{bmatrix} 3 & -1 \\ -2 & 4 \end{bmatrix}$ has $\det \frac{1}{10}$. $\det(A - \lambda I) = \lambda^2 - 7\lambda + 10 = 0$ when $\lambda = 2$ or $\lambda = 5$; those are eigenvalues.
- 24** Here $A = LU$ with $\det(L) = 1$ and $\det(U) = -6$ product of pivots, so also $\det(A) = -6$. $\det(U^{-1}L^{-1}) = -\frac{1}{6} = 1/\det(A)$ and $\det(U^{-1}L^{-1}A)$ is $\det I = 1$.
- 25** When the i, j entry is ij , row 2 = 2 times row 1 so $\det A = 0$.
- 26** When the ij entry is $i + j$, row 3 - row 2 = row2 - row 1 so A is singular: $\det A = 0$.
- 27** $\det A = abc$, $\det B = -abcd$, $\det C = a(b-a)(c-b)$ by doing elimination.
- 28** (a) *True*: $\det(AB) = \det(A)\det(B) = 0$ (b) *False*: A row exchange gives $-\det =$ product of pivots. (c) *False*: $A = 2I$ and $B = I$ have $A - B = I$ but the determinants have $2^n - 1 \neq 1$ (d) *True*: $\det(AB) = \det(A)\det(B) = \det(BA)$.
- 29** A is rectangular so $\det(A^T A) \neq (\det A^T)(\det A)$: these determinants are not defined.
- 30** Derivatives of $f = \ln(ad - bc)$:
$$\begin{bmatrix} \partial f / \partial a & \partial f / \partial c \\ \partial f / \partial b & \partial f / \partial d \end{bmatrix} = \begin{bmatrix} \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ \frac{-c}{ad-bc} & \frac{a}{ad-bc} \end{bmatrix} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = A^{-1}.$$
- 31** The Hilbert determinants are $1, 8 \times 10^{-2}, 4.6 \times 10^{-4}, 1.6 \times 10^{-7}, 3.7 \times 10^{-12}, 5.4 \times 10^{-18}, 4.8 \times 10^{-25}, 2.7 \times 10^{-33}, 9.7 \times 10^{-43}, 2.2 \times 10^{-53}$. Pivots are ratios of determinants so the 10th pivot is near 10^{-10} . The Hilbert matrix is numerically difficult (*ill-conditioned*).

- 32** Typical determinants of $\text{rand}(n)$ are $10^6, 10^{25}, 10^{79}, 10^{218}$ for $n = 50, 100, 200, 400$. $\text{randn}(n)$ with normal distribution gives $10^{31}, 10^{78}, 10^{186}, \text{Inf}$ which means $\geq 2^{1024}$. MATLAB allows $1.999999999999999 \times 2^{1023} \approx 1.8 \times 10^{308}$ but one more 9 gives Inf!
- 33** I now know that maximizing the determinant for 1, -1 matrices is **Hadamard's problem** (1893): see Brenner in American Math. Monthly volume 79 (1972) 626-630. Neil Sloane's wonderful On-Line Encyclopedia of Integer Sequences (research.att.com/~njas) includes the solution for small n (and more references) when the problem is changed to 0, 1 matrices. That sequence A003432 starts from $n = 0$ with 1, 1, 1, 2, 3, 5, 9. Then the 1, -1 maximum for size n is 2^{n-1} times the 0, 1 maximum for size $n - 1$ (so $(32)(5) = 160$ for $n = 6$ in sequence **A003433**).

To reduce the 1, -1 problem from 6 by 6 to the 0, 1 problem for 5 by 5, multiply the six rows by ± 1 to put $+1$ in column 1. Then subtract row 1 from rows 2 to 6 to get a 5 by 5 submatrix S of $-2, 0$ and divide S by -2 .

Here is an advanced MATLAB code and a 1, -1 matrix with largest $\det A = 48$ for $n = 5$:

```
n = 5; p = (n - 1)^2; A0 = ones(n); maxdet = 0;
for k = 0 : 2^p - 1
    Asub = rem(floor(k * 2.^(-p + 1 : 0)), 2); A = A0; A(2 : n, 2 : n) = 1 - 2 *
    reshape(Asub, n - 1, n - 1);
    if abs(det(A)) > maxdet, maxdet = abs(det(A)); maxA = A;
end
end
```

Output: maxA =

1	1	1	1	1
1	1	1	-1	-1
1	1	-1	1	-1
1	-1	1	1	-1
1	-1	-1	-1	1

maxdet = 48.

- 34** Reduce B by row operations to [row 3; row 2; row 1]. Then $\det B = -6$ (odd permutation).

Problem Set 5.2, page 263

- 1** $\det A = 1 + 18 + 12 - 9 - 4 - 6 = 12$, rows are independent; $\det B = 0$, row 1 + row 2 = row 3; $\det C = -1$, independent rows ($\det C$ has one term, odd permutation)
- 2** $\det A = -2$, independent; $\det B = 0$, dependent; $\det C = -1$, independent.
- 3** All cofactors of row 1 are zero. A has rank ≤ 2 . Each of the 6 terms in $\det A$ is zero. Column 2 has no pivot.
- 4** $a_{11}a_{23}a_{32}a_{44}$ gives -1 , because $2 \leftrightarrow 3$, $a_{14}a_{23}a_{32}a_{41}$ gives $+1$, $\det A = 1 - 1 = 0$; $\det B = 2 \cdot 4 \cdot 4 \cdot 2 - 1 \cdot 4 \cdot 4 \cdot 1 = 64 - 16 = 48$.
- 5** Four zeros in the same row guarantee $\det = 0$. $A = I$ has 12 zeros (maximum with $\det \neq 0$).
- 6** (a) If $a_{11} = a_{22} = a_{33} = 0$ then 4 terms are sure zeros (b) 15 terms must be zero.

- 7 $5!/2 = 60$ permutation matrices have $\det = +1$. Move row 5 of I to the top; starting from $(5, 1, 2, 3, 4)$ elimination will do four row exchanges.
- 8 Some term $a_{1\alpha}a_{2\beta}\cdots a_{n\omega}$ in the big formula is not zero! Move rows $1, 2, \dots, n$ into rows $\alpha, \beta, \dots, \omega$. Then these nonzero a 's will be on the main diagonal.
- 9 To get $+1$ for the even permutations, the matrix needs an *even* number of -1 's. To get $+1$ for the odd P 's, the matrix needs an *odd* number of -1 's. So all six terms $= +1$ in the big formula and $\det = 6$ are impossible: $\max(\det) = 4$.
- 10 The $4!/2 = 12$ even permutations are $(1, 2, 3, 4), (2, 1, 4, 3), (3, 1, 4, 2), (4, 3, 2, 1)$, and 8 P 's with one number in place and even permutation of the other three numbers. $\det(I + P_{\text{even}}) = 16$ or 4 or 0 (16 comes from $I + I$).
- 11 $C = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$. $D = \begin{bmatrix} 0 & 42 & -35 \\ 0 & -21 & 14 \\ -3 & 6 & -3 \end{bmatrix}$. $\det B = 1(0) + 2(42) + 3(-35) = -21$. Puzzle: $\det D = 441 = (-21)^2$. Why?
- 12 $C = \begin{bmatrix} 3 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 3 \end{bmatrix}$ and $AC^T = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 4 \end{bmatrix}$. Therefore $A^{-1} = \frac{1}{4}C^T = C^T / \det A$.
- 13 (a) $C_1 = 0, C_2 = -1, C_3 = 0, C_4 = 1$ (b) $C_n = -C_{n-2}$ by cofactors of row 1 then cofactors of column 1. Therefore $C_{10} = -C_8 = C_6 = -C_4 = C_2 = -1$.
- 14 We must choose 1's from column 2 then column 1, column 4 then column 3, and so on. Therefore n must be even to have $\det A_n \neq 0$. The number of row exchanges is $n/2$ so $C_n = (-1)^{n/2}$.
- 15 The 1, 1 cofactor of the n by n matrix is E_{n-1} . The 1, 2 cofactor has a single 1 in its first column, with cofactor E_{n-2} : sign gives $-E_{n-2}$. So $E_n = E_{n-1} - E_{n-2}$. Then E_1 to E_6 is $1, 0, -1, -1, 0, 1$ and this cycle of six will repeat: $E_{100} = E_4 = -1$.
- 16 The 1, 1 cofactor of the n by n matrix is F_{n-1} . The 1, 2 cofactor has a 1 in column 1, with cofactor F_{n-2} . Multiply by $(-1)^{1+2}$ and also (-1) from the 1, 2 entry to find $F_n = F_{n-1} + F_{n-2}$ (so these determinants are Fibonacci numbers).
- 17 $|B_4| = 2 \det \begin{bmatrix} 1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} + \det \begin{bmatrix} 1 & -1 \\ -1 & 2 \\ -1 & -1 \end{bmatrix} = 2|B_3| - \det \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} = 2|B_3| - |B_2|$. $|B_3|$ and $-|B_2|$ are cofactors of row 4 of B_4 .
- 18 Rule 3 (linearity in row 1) gives $|B_n| = |A_n| - |A_{n-1}| = (n+1) - n = 1$.
- 19 Since x, x^2, x^3 are all in the same row, they are never multiplied in $\det V_4$. The determinant is zero at $x = a$ or b or c , so $\det V$ has factors $(x-a)(x-b)(x-c)$. Multiply by the cofactor V_3 . The Vandermonde matrix $V_{ij} = (x_i)^{j-1}$ is for fitting a polynomial $p(x) = b$ at the points x_i . It has $\det V = \text{product of all } x_k - x_m \text{ for } k > m$.
- 20 $G_2 = -1, G_3 = 2, G_4 = -3$, and $G_n = (-1)^{n-1}(n-1) = (\text{product of the } \lambda\text{'s})$.
- 21 $S_1 = 3, S_2 = 8, S_3 = 21$. The rule looks like every second number in Fibonacci's sequence $\dots 3, 5, 8, 13, 21, 34, 55, \dots$ so the guess is $S_4 = 55$. Following the solution to Problem 30 with 3's instead of 2's confirms $S_4 = 81 + 1 - 9 - 9 - 9 = 55$. Problem 33 directly proves $S_n = F_{2n+2}$.
- 22 Changing 3 to 2 in the corner reduces the determinant F_{2n+2} by 1 times the cofactor of that corner entry. This cofactor is the determinant of S_{n-1} (one size smaller) which is F_{2n} . Therefore changing 3 to 2 changes the determinant to $F_{2n+2} - F_{2n}$ which is F_{2n+1} .

- 23** (a) If we choose an entry from B we must choose an entry from the zero block; result zero. This leaves entries from A times entries from D leading to $(\det A)(\det D)$
 (b) and (c) Take $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$, $C = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, $D = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$. See #25.
- 24** (a) All L 's have $\det = 1$; $\det U_k = \det A_k = 2, 6, -6$ for $k = 1, 2, 3$ (b) Pivots $2, \frac{3}{2}, \frac{-1}{3}$.
- 25** Problem 23 gives $\det \begin{bmatrix} I & 0 \\ -CA^{-1} & I \end{bmatrix} = 1$ and $\det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = |A|$ times $|D - CA^{-1}B|$ which is $|AD - ACA^{-1}B|$. If $AC = CA$ this is $|AD - CAA^{-1}B| = \det(AD - CB)$.
- 26** If A is a row and B is a column then $\det M = \det AB = \text{dot product of } A \text{ and } B$. If A is a column and B is a row then AB has rank 1 and $\det M = \det AB = 0$ (unless $m = n = 1$). This block matrix is invertible when AB is invertible which certainly requires $m \leq n$.
- 27** (a) $\det A = a_{11}C_{11} + \cdots + a_{1n}C_{1n}$. Derivative with respect to $a_{11} = \text{cofactor } C_{11}$.
- 28** Row 1 $- 2$ row 2 $+ \text{row } 3 = 0$ so this matrix is singular.
- 29** There are five nonzero products, all 1's with a plus or minus sign. Here are the (row, column) numbers and the signs: $+(1, 1)(2, 2)(3, 3)(4, 4) + (1, 2)(2, 1)(3, 4)(4, 3) - (1, 2)(2, 1)(3, 3)(4, 4) - (1, 1)(2, 2)(3, 4)(4, 3) - (1, 1)(2, 3)(3, 2)(4, 4)$. Total -1 .
- 30** The 5 products in solution 29 change to $16 + 1 - 4 - 4 - 4$ since A has 2's and -1 's:
 $(2)(2)(2)(2) + (-1)(-1)(-1)(-1) - (-1)(-1)(2)(2) - (2)(2)(-1)(-1) - (2)(-1)(-1)(2)$.
- 31** $\det P = -1$ because the cofactor of $P_{14} = 1$ in row one has sign $(-1)^{1+4}$. The big formula for $\det P$ has only one term $(1 \cdot 1 \cdot 1 \cdot 1)$ with minus sign because three exchanges take 4, 1, 2, 3 into 1, 2, 3, 4; $\det(P^2) = (\det P)(\det P) = +1$ so $\det \begin{bmatrix} 0 & I \\ I & 0 \end{bmatrix} = \det \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ is *not right*.
- 32** The problem is to show that $F_{2n+2} = 3F_{2n} - F_{2n-2}$. Keep using Fibonacci's rule:
 $F_{2n+2} = F_{2n+1} + F_{2n} = F_{2n} + F_{2n-1} + F_{2n} = 2F_{2n} + (F_{2n} - F_{2n-2}) = 3F_{2n} - F_{2n-2}$.
- 33** The difference from 20 to 19 multiplies its 3 by 3 cofactor $= 1$: then \det drops by 1.
- 34** (a) The last three rows must be dependent (b) In each of the 120 terms: Choices from the last 3 rows must use 3 columns; at least one of those choices will be zero.
- 35** Subtracting 1 from the n, n entry subtracts its cofactor C_{nn} from the determinant. That cofactor is $C_{nn} = 1$ (smaller Pascal matrix). Subtracting 1 from 1 leaves 0.

Problem Set 5.3, page 279

- 1** (a) $\begin{vmatrix} 2 & 5 \\ 1 & 4 \end{vmatrix} = 3$, $\begin{vmatrix} 1 & 5 \\ 2 & 4 \end{vmatrix} = 6$, $\begin{vmatrix} 2 & 1 \\ 1 & 2 \end{vmatrix} = 3$ so $x_1 = -6/3 = -2$ and $x_2 = 3/3 = 1$ (b) $|A| = 4$, $|B_1| = 3$, $|B_2| = 2$, $|B_3| = 1$. Therefore $x_1 = 3/4$ and $x_2 = -1/2$ and $x_3 = 1/4$.

- 2 (a) $y = \begin{vmatrix} a & 1 \\ c & 0 \end{vmatrix} / \begin{vmatrix} a & b \\ c & d \end{vmatrix} = c/(ad - bc)$ (b) $y = \det B_2 / \det A = (fg - id)/D$.
- 3 (a) $x_1 = 3/0$ and $x_2 = -2/0$: no solution (b) $x_1 = x_2 = 0/0$: undetermined.
- 4 (a) $x_1 = \det(\begin{bmatrix} \mathbf{b} & \mathbf{a}_2 & \mathbf{a}_3 \end{bmatrix}) / \det A$, if $\det A \neq 0$ (b) The determinant is linear in its first column so $x_1|\mathbf{a}_1 \mathbf{a}_2 \mathbf{a}_3| + x_2|\mathbf{a}_2 \mathbf{a}_2 \mathbf{a}_3| + x_3|\mathbf{a}_3 \mathbf{a}_2 \mathbf{a}_3|$. The last two determinants are zero because of repeated columns, leaving $x_1|\mathbf{a}_1 \mathbf{a}_2 \mathbf{a}_3|$ which is $x_1 \det A$.
- 5 If the first column in A is also the right side \mathbf{b} then $\det A = \det B_1$. Both B_2 and B_3 are singular since a column is repeated. Therefore $x_1 = |B_1|/|A| = 1$ and $x_2 = x_3 = 0$.
- 6 (a) $\begin{bmatrix} 1 & -\frac{2}{3} & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & -\frac{7}{3} & 1 \end{bmatrix}$ (b) $\frac{1}{4} \begin{bmatrix} 3 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 3 \end{bmatrix}$. An invertible symmetric matrix has a symmetric inverse.
- 7 If all cofactors = 0 then A^{-1} would be the zero matrix if it existed; cannot exist. (And the cofactor formula gives $\det A = 0$.) $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ has no zero cofactors but it is not invertible.
- 8 $C = \begin{bmatrix} 6 & -3 & 0 \\ 3 & 1 & -1 \\ -6 & 2 & 1 \end{bmatrix}$ and $AC^T = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$. This is $(\det A)I$ and $\det A = 3$.
The 1, 3 cofactor of A is 0.
Multiplying by 4 or 100: no change.
- 9 If we know the cofactors and $\det A = 1$, then $C^T = A^{-1}$ and also $\det A^{-1} = 1$. Now A is the inverse of C^T , so A can be found from the cofactor matrix for C .
- 10 Take the determinant of $AC^T = (\det A)I$. The left side gives $\det AC^T = (\det A)(\det C)$ while the right side gives $(\det A)^n$. Divide by $\det A$ to reach $\det C = (\det A)^{n-1}$.
- 11 The cofactors of A are integers. Division by $\det A = \pm 1$ gives integer entries in A^{-1} .
- 12 Both $\det A$ and $\det A^{-1}$ are integers since the matrices contain only integers. But $\det A^{-1} = 1/\det A$ so $\det A$ must be 1 or -1.
- 13 $A = \begin{bmatrix} 0 & 1 & 3 \\ 1 & 0 & 1 \\ 2 & 1 & 0 \end{bmatrix}$ has cofactor matrix $C = \begin{bmatrix} -1 & 2 & 1 \\ 3 & -6 & 2 \\ 1 & 3 & -1 \end{bmatrix}$ and $A^{-1} = \frac{1}{5}C^T$.
- 14 (a) Lower triangular L has cofactors $C_{21} = C_{31} = C_{32} = 0$ (b) $C_{12} = C_{21}$, $C_{31} = C_{13}$, $C_{32} = C_{23}$ make S^{-1} symmetric. (c) Orthogonal Q has cofactor matrix $C = (\det Q)(Q^{-1})^T = \pm Q$ also orthogonal. Note $\det Q = 1$ or -1 .
- 15 For $n = 5$, C contains 25 cofactors and each 4 by 4 cofactor has 24 terms. Each term needs 3 multiplications: total 1800 multiplications vs. 125 for Gauss-Jordan.
- 16 (a) Area $\begin{vmatrix} 3 & 2 \\ 1 & 4 \end{vmatrix} = 10$ (b) and (c) Area $10/2 = 5$, these triangles are half of the parallelogram in (a).
- 17 Volume = $\begin{vmatrix} 3 & 1 & 1 \\ 1 & 3 & 1 \\ 1 & 1 & 3 \end{vmatrix} = 20$. Area of faces = length of cross product = $\begin{vmatrix} i & j & k \\ 3 & 1 & 1 \\ 1 & 3 & 1 \end{vmatrix} = -2i - 2j + 8k$ length = $6\sqrt{2}$
- 18 (a) Area $\frac{1}{2} \begin{vmatrix} 2 & 1 & 1 \\ 3 & 4 & 1 \\ 0 & 5 & 1 \end{vmatrix} = 5$ (b) 5 + new triangle area $\frac{1}{2} \begin{vmatrix} 2 & 1 & 1 \\ 0 & 5 & 1 \\ -1 & 0 & 1 \end{vmatrix} = 5 + 7 = 12$.
- 19 $\begin{vmatrix} 2 & 1 \\ 2 & 3 \end{vmatrix} = 4 = \begin{vmatrix} 2 & 2 \\ 1 & 3 \end{vmatrix}$ because the transpose has the same determinant. See #22.

- 20** The edges of the hypercube have length $\sqrt{1+1+1+1} = 2$. The volume $\det H$ is $2^4 = 16$. ($H/2$ has orthonormal columns. Then $\det(H/2) = 1$ leads again to $\det H = 16$.)
- 21** The maximum volume $L_1 L_2 L_3 L_4$ is reached when the edges are orthogonal in \mathbf{R}^4 . With entries 1 and -1 all lengths are $\sqrt{4} = 2$. The maximum determinant is $2^4 = 16$, achieved in Problem 20. For a 3 by 3 matrix, $\det A = (\sqrt{3})^3$ can't be achieved by ± 1 .
- 22** This question is still waiting for a solution! An 18.06 student showed me how to transform the parallelogram for A to the parallelogram for A^T , without changing its area. (Edges slide along themselves, so no change in baselength or height or area.)
- 23** $A^T A = \begin{bmatrix} \mathbf{a}^T \\ \mathbf{b}^T \\ \mathbf{c}^T \end{bmatrix} \begin{bmatrix} \mathbf{a} & \mathbf{b} & \mathbf{c} \end{bmatrix} = \begin{bmatrix} \mathbf{a}^T \mathbf{a} & 0 & 0 \\ 0 & \mathbf{b}^T \mathbf{b} & 0 \\ 0 & 0 & \mathbf{c}^T \mathbf{c} \end{bmatrix}$ has $\frac{\det A^T A}{\det A} = \frac{(\|\mathbf{a}\| \|\mathbf{b}\| \|\mathbf{c}\|)^2}{\pm \|\mathbf{a}\| \|\mathbf{b}\| \|\mathbf{c}\|}$
- 24** The box has height 4 and volume $= \det \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 3 & 4 \end{bmatrix} = 4$. $\mathbf{i} \times \mathbf{j} = \mathbf{k}$ and $(\mathbf{k} \cdot \mathbf{w}) = 4$.
- 25** The n -dimensional cube has 2^n corners, $n2^{n-1}$ edges and $2n(n-1)$ -dimensional faces. Coefficients from $(2+x)^n$ in Worked Example 2.4A. Cube from $2I$ has volume 2^n .
- 26** The pyramid has volume $\frac{1}{6}$. The 4-dimensional pyramid has volume $\frac{1}{24}$ (and $\frac{1}{n!}$ in \mathbf{R}^n)
- 27** $x = r \cos \theta$, $y = r \sin \theta$ give $J = r$. The columns are orthogonal and their lengths are 1 and r .
- 28** $J = \begin{vmatrix} \sin \varphi \cos \theta & \rho \cos \varphi \sin \theta & -\rho \sin \varphi \sin \theta \\ \sin \varphi \sin \theta & \rho \cos \varphi \cos \theta & \rho \sin \varphi \cos \theta \\ \cos \varphi & -\rho \sin \varphi & \theta \end{vmatrix} = \rho^2 \sin \varphi$. This Jacobian is needed for triple integrals inside spheres.
- 29** From x, y to r, θ : $\begin{vmatrix} \partial r / \partial x & \partial r / \partial y \\ \partial \theta / \partial x & \partial \theta / \partial y \end{vmatrix} = \begin{vmatrix} x/r & y/r \\ -y/r^2 & x/r^2 \end{vmatrix} = \begin{vmatrix} \cos \theta & \sin \theta \\ (-\sin \theta)/r & (\cos \theta)/r \end{vmatrix} = \frac{1}{r} = \frac{1}{\text{Jacobian in 27}}$.
- 30** The triangle with corners $(0, 0)$, $(6, 0)$, $(1, 4)$ has area 24. Rotated by $\theta = 60^\circ$ the area is *unchanged*. The determinant of the rotation matrix is $J = \begin{vmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{vmatrix} = \begin{vmatrix} 1/2 & -\sqrt{3}/2 \\ \sqrt{3}/2 & 1/2 \end{vmatrix} = 1$.
- 31** Base area 10, height 2, volume 20.
- 32** The volume of the box is $\det \begin{bmatrix} 2 & 4 & 0 \\ -1 & 3 & 0 \\ 1 & 2 & 2 \end{bmatrix} = 20$.
- 33** $\begin{vmatrix} u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \\ w_1 & w_2 & w_3 \end{vmatrix} = u_1 \begin{vmatrix} v_2 & v_3 \\ w_2 & w_3 \end{vmatrix} - u_2 \begin{vmatrix} v_1 & v_3 \\ w_1 & w_3 \end{vmatrix} + u_3 \begin{vmatrix} v_1 & v_2 \\ w_1 & w_2 \end{vmatrix}$. This is $\mathbf{u} \cdot (\mathbf{v} \times \mathbf{w})$.
- 34** $(\mathbf{w} \times \mathbf{u}) \cdot \mathbf{v} = (\mathbf{v} \times \mathbf{w}) \cdot \mathbf{u} = (\mathbf{u} \times \mathbf{v}) \cdot \mathbf{w}$: Even permutation of $(\mathbf{u}, \mathbf{v}, \mathbf{w})$ keeps the same determinant. Odd permutations reverse the sign.

- 35** $S = (2, 1, -1)$, area $\|PQ \times PS\| = \|(-2, -2, -1)\| = 3$. The other four corners can be $(0, 0, 0)$, $(0, 0, 2)$, $(1, 2, 2)$, $(1, 1, 0)$. The volume of the tilted box is $|\det| = 1$.
- 36** If $(1, 1, 0)$, $(1, 2, 1)$, (x, y, z) are in a plane the volume is $\det \begin{bmatrix} x & y & z \\ 1 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix} = x - y + z = 0$. The “box” with those edges is flattened to zero height.
- 37** $\det \begin{bmatrix} x & y & z \\ 2 & 3 & 1 \\ 1 & 2 & 3 \end{bmatrix} = 7x - 5y + z$ will be *zero* when (x, y, z) is a combination of $(2, 3, 1)$ and $(1, 2, 3)$. The plane containing those two vectors has equation $7x - 5y + z = 0$.
- 38** Doubling each row multiplies the volume by 2^n . Then $2 \det A = \det(2A)$ only if $n = 1$.
- 39** $AC^T = (\det A)I$ gives $(\det A)(\det C) = (\det A)^n$. Then $\det A = (\det C)^{1/3}$ with $n = 4$. With $\det A^{-1} = 1/\det A$, construct A^{-1} using the cofactors. *Invert to find A.*
- 40** The cofactor formula adds 1 by 1 determinants (which are just entries) *times* their cofactors of size $n - 1$. Jacobi discovered that this formula can be generalized. For $n = 5$, Jacobi multiplied each 2 by 2 determinant from rows 1-2 (with columns $a < b$) times a 3 by 3 determinant from rows 3-5 (using the remaining columns $c < d < e$).
The key question is $+$ or $-$ sign (as for cofactors). The product is given a $+$ sign when a, b, c, d, e is an even permutation of $1, 2, 3, 4, 5$. This gives the correct determinant $+1$ for that permutation matrix. More than that, all other P that permute a, b and separately c, d, e will come out with the correct sign when the 2 by 2 determinant for columns a, b multiplies the 3 by 3 determinant for columns c, d, e .
- 41** The Cauchy-Binet formula gives the determinant of a square matrix AB (and AA^T in particular) when the factors A, B are rectangular. For (2 by 3) times (3 by 2) there are 3 products of 2 by 2 determinants from A and B (printed in boldface):

$$\begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix} \begin{bmatrix} g & j \\ h & k \\ i & \ell \end{bmatrix} \quad \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix} \begin{bmatrix} g & j \\ h & k \\ i & \ell \end{bmatrix} \quad \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix} \begin{bmatrix} g & j \\ h & k \\ i & \ell \end{bmatrix}$$

$$\text{Check } A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 4 & 7 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 1 \\ 2 & 4 \\ 3 & 7 \end{bmatrix} \quad AB = \begin{bmatrix} 14 & 30 \\ 30 & 66 \end{bmatrix}$$

$$\text{Cauchy-Binet: } (4 - 2)(4 - 2) + (7 - 3)(7 - 3) + (14 - 12)(14 - 12) = \mathbf{24} \\ (14)(66) - (30)(30) = \mathbf{24}$$

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- 1** The eigenvalues are 1 and 0.5 for A , 1 and 0.25 for A^2 , 1 and 0 for A^∞ . Exchanging the rows of A changes the eigenvalues to 1 and -0.5 (the trace is now $0.2 + 0.3$). Singular matrices stay singular during elimination, so $\lambda = 0$ does not change.
- 2** A has $\lambda_1 = -1$ and $\lambda_2 = 5$ with eigenvectors $x_1 = (-2, 1)$ and $x_2 = (1, 1)$. The matrix $A + I$ has the same eigenvectors, with eigenvalues increased by 1 to **0** and **6**. That zero eigenvalue correctly indicates that $A + I$ is singular.
- 3** A has $\lambda_1 = 2$ and $\lambda_2 = -1$ (check trace and determinant) with $x_1 = (1, 1)$ and $x_2 = (2, -1)$. A^{-1} has the same eigenvectors, with eigenvalues $1/\lambda = \frac{1}{2}$ and -1 .

- 4** A has $\lambda_1 = -3$ and $\lambda_2 = 2$ (check trace $= -1$ and determinant $= -6$) with $x_1 = (3, -2)$ and $x_2 = (1, 1)$. A^2 has the *same eigenvectors* as A , with eigenvalues $\lambda_1^2 = 9$ and $\lambda_2^2 = 4$.
- 5** A and B have eigenvalues 1 and 3. $A + B$ has $\lambda_1 = 3, \lambda_2 = 5$. Eigenvalues of $A + B$ are *not equal* to eigenvalues of A plus eigenvalues of B .
- 6** A and B have $\lambda_1 = 1$ and $\lambda_2 = 1$. AB and BA have $\lambda = 2 \pm \sqrt{3}$. Eigenvalues of AB are *not equal* to eigenvalues of A times eigenvalues of B . Eigenvalues of AB and BA *are equal* (this is proved in section 6.6, Problems 18-19).
- 7** The eigenvalues of U (on its diagonal) are the *pivots* of A . The eigenvalues of L (on its diagonal) are all 1's. The eigenvalues of A are *not* the same as the pivots.
- 8** (a) Multiply Ax to see λx which reveals λ (b) Solve $(A - \lambda I)x = 0$ to find x .
- 9** (a) Multiply by A : $A(Ax) = A(\lambda x) = \lambda Ax$ gives $A^2x = \lambda^2x$ (b) Multiply by A^{-1} : $x = A^{-1}Ax = A^{-1}\lambda x = \lambda A^{-1}x$ gives $A^{-1}x = \frac{1}{\lambda}x$ (c) Add $Ix = x$: $(A + I)x = (\lambda + 1)x$.
- 10** A has $\lambda_1 = 1$ and $\lambda_2 = .4$ with $x_1 = (1, 2)$ and $x_2 = (1, -1)$. A^∞ has $\lambda_1 = 1$ and $\lambda_2 = 0$ (same eigenvectors). A^{100} has $\lambda_1 = 1$ and $\lambda_2 = (.4)^{100}$ which is near zero. So A^{100} is very near A^∞ : same eigenvectors and close eigenvalues.
- 11** Columns of $A - \lambda_1 I$ are in the nullspace of $A - \lambda_2 I$ because $M = (A - \lambda_2 I)(A - \lambda_1 I) = \text{zero matrix}$ [this is the *Cayley-Hamilton Theorem* in Problem 6.2.32]. Notice that M has *zero eigenvalues* $(\lambda_1 - \lambda_2)(\lambda_1 - \lambda_1) = 0$ and $(\lambda_2 - \lambda_2)(\lambda_2 - \lambda_1) = 0$.
- 12** The projection matrix P has $\lambda = 1, 0, 1$ with eigenvectors $(1, 2, 0), (2, -1, 0), (0, 0, 1)$. Add the first and last vectors: $(1, 2, 1)$ also has $\lambda = 1$. Note $P^2 = P$ leads to $\lambda^2 = \lambda$ so $\lambda = 0$ or 1 .
- 13** (a) $Pu = (uu^T)u = u(u^Tu) = u$ so $\lambda = 1$ (b) $Pv = (uu^T)v = u(u^Tv) = 0$ (c) $x_1 = (-1, 1, 0, 0), x_2 = (-3, 0, 1, 0), x_3 = (-5, 0, 0, 1)$ all have $Px = 0x = 0$.
- 14** Two eigenvectors of this rotation matrix are $x_1 = (1, i)$ and $x_2 = (1, -i)$ (more generally cx_1 , and dx_2 with $cd \neq 0$).
- 15** The other two eigenvalues are $\lambda = \frac{1}{2}(-1 \pm i\sqrt{3})$; the three eigenvalues are $1, 1, -1$.
- 16** Set $\lambda = 0$ in $\det(A - \lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda)$ to find $\det A = (\lambda_1)(\lambda_2) \cdots (\lambda_n)$.
- 17** $\lambda_1 = \frac{1}{2}(a + d + \sqrt{(a-d)^2 + 4bc})$ and $\lambda_2 = \frac{1}{2}(a + d - \sqrt{(a-d)^2 + 4bc})$ add to $a + d$. If A has $\lambda_1 = 3$ and $\lambda_2 = 4$ then $\det(A - \lambda I) = (\lambda - 3)(\lambda - 4) = \lambda^2 - 7\lambda + 12$.
- 18** These 3 matrices have $\lambda = 4$ and 5, trace 9, det 20: $\begin{bmatrix} 4 & 0 \\ 0 & 5 \end{bmatrix}, \begin{bmatrix} 3 & 2 \\ -1 & 6 \end{bmatrix}, \begin{bmatrix} 2 & 2 \\ -3 & 7 \end{bmatrix}$.
- 19** (a) rank = 2 (b) $\det(B^T B) = 0$ (d) eigenvalues of $(B^2 + I)^{-1}$ are $1, \frac{1}{2}, \frac{1}{5}$.
- 20** $A = \begin{bmatrix} 0 & 1 \\ -28 & 11 \end{bmatrix}$ has trace 11 and determinant 28, so $\lambda = 4$ and 7. Moving to a 3 by 3 companion matrix, $C = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 6 & -11 & 6 \end{bmatrix}$ has $\det(C - \lambda I) = -\lambda^3 + 6\lambda^2 - 11\lambda + 6 = (1 - \lambda)(2 - \lambda)(3 - \lambda)$. Notice the trace $6 = 1 + 2 + 3$, determinant $6 = (1)(2)(3)$, and also $11 = (1)(2) + (1)(3) + (2)(3)$.

- 21** $(A - \lambda I)$ has the same determinant as $(A - \lambda I)^T$. $\begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$ and $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ have *different* because every square matrix has $\det M = \det M^T$.
- 22** $\lambda = 1$ (for Markov), 0 (for singular), $-\frac{1}{2}$ (so sum of eigenvalues = trace = $\frac{1}{2}$).
- 23** $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$, $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, $\begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}$. Always A^2 is the zero matrix if $\lambda = 0$ and 0 , by the Cayley-Hamilton Theorem in Problem 6.2.32.
- 24** $\lambda = 0, 0, 6$ (notice rank 1 and trace 6) with $\mathbf{x}_1 = (0, -2, 1)$, $\mathbf{x}_2 = (1, -2, 0)$, $\mathbf{x}_3 = (1, 2, 1)$.
- 25** With the same n λ 's and \mathbf{x} 's, $A\mathbf{x} = c_1\lambda_1\mathbf{x}_1 + \cdots + c_n\lambda_n\mathbf{x}_n$ equals $B\mathbf{x} = c_1\lambda_1\mathbf{x}_1 + \cdots + c_n\lambda_n\mathbf{x}_n$ for all vectors \mathbf{x} . So $A = B$.
- 26** The block matrix has $\lambda = 1, 2$ from B and $5, 7$ from D . All entries of C are multiplied by zeros in $\det(A - \lambda I)$, so C has no effect on the eigenvalues.
- 27** A has rank 1 with eigenvalues $0, 0, 0, 4$ (the 4 comes from the trace of A). C has rank 2 (ensuring two zero eigenvalues) and $(1, 1, 1, 1)$ is an eigenvector with $\lambda = 2$. With trace 4, the other eigenvalue is also $\lambda = 2$, and its eigenvector is $(1, -1, 1, -1)$.
- 28** B has $\lambda = -1, -1, -1, 3$ and C has $\lambda = 1, 1, 1, -3$. Both have $\det = -3$.
- 29** Triangular matrix: $\lambda(A) = 1, 4, 6$; $\lambda(B) = 2, \sqrt{3}, -\sqrt{3}$; Rank-1 matrix: $\lambda(C) = 0, 0, 6$.
- 30** $\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} a+b \\ c+d \end{bmatrix} = (a+b) \begin{bmatrix} 1 \\ 1 \end{bmatrix}$; $\lambda_2 = d - b$ to produce the correct trace $(a+b) + (d-b) = a+d$.
- 31** Eigenvector $(1, 3, 4)$ for A with $\lambda = 11$ and eigenvector $(3, 1, 4)$ for PAP^T . Eigenvectors with $\lambda \neq 0$ must be in the column space since $A\mathbf{x}$ is always in the column space, and $\mathbf{x} = A\mathbf{x}/\lambda$.
- 32** (a) \mathbf{u} is a basis for the nullspace, \mathbf{v} and \mathbf{w} give a basis for the column space
 (b) $\mathbf{x} = (0, \frac{1}{3}, \frac{1}{5})$ is a particular solution. Add any $c\mathbf{u}$ from the nullspace
 (c) If $A\mathbf{x} = \mathbf{u}$ had a solution, \mathbf{u} would be in the column space: wrong dimension 3.
- 33** If $\mathbf{v}^T\mathbf{u} = 0$ then $A^2 = \mathbf{u}(\mathbf{v}^T\mathbf{u})\mathbf{v}^T$ is the zero matrix and $\lambda^2 = 0, 0$ and $\lambda = 0, 0$ and trace $(A) = 0$. This zero trace also comes from adding the diagonal entries of $A = \mathbf{u}\mathbf{v}^T$:

$$A = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \begin{bmatrix} v_1 & v_2 \end{bmatrix} = \begin{bmatrix} u_1v_1 & u_1v_2 \\ u_2v_1 & u_2v_2 \end{bmatrix} \quad \text{has trace } u_1v_1 + u_2v_2 = \mathbf{v}^T\mathbf{u} = 0$$

- 34** $\det(P - \lambda I) = 0$ gives the equation $\lambda^4 = 1$. This reflects the fact that $P^4 = I$. The solutions of $\lambda^4 = 1$ are $\lambda = 1, i, -1, -i$. The real eigenvector $\mathbf{x}_1 = (1, 1, 1, 1)$ is not changed by the permutation P . Three more eigenvectors are (i, i^2, i^3, i^4) and $(1, -1, 1, -1)$ and $(-i, (-i)^2, (-i)^3, (-i)^4)$.
- 35** 3 by 3 permutation matrices: Since $P^T P = I$ gives $(\det P)^2 = 1$, the determinant is 1 or -1 . The pivots are always 1 (but there may be row exchanges). The trace of P can be 3 (for $P = I$) or 1 (for row exchange) or 0 (for double exchange). The possible eigenvalues are 1 and -1 and $e^{2\pi i/3}$ and $e^{-2\pi i/3}$.