

## Key to Problem Set #5

1. (a) The 4 expected outputs are  $(c \ c \ c \ c)^T = (1 \ 1 \ 1 \ 1)^T (c) = \mathbf{Ax}$ . Equating these to the 4 actual outputs  $(b_1 \ b_2 \ b_3 \ b_4)^T = \mathbf{b}$  gives the linear system  $\mathbf{Ax} = \mathbf{b}$ .

$$(b) \mathbf{P} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T = \frac{1}{4} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} = \frac{1}{4} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix}.$$

- (c)  $\bar{c} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b} = \frac{1}{4}(b_1 + b_2 + b_3 + b_4)$ . That is, the best least-squares estimate of  $c$  is the simple average of the observations.

2. Let  $n$  be the size of  $\mathbf{I}_n$ . If  $n = 1$ ,  $\mathbf{I}_1 = (1)$   $\det(\mathbf{I}_1) = 1$  by definition of the determinant of a  $1 \times 1$  matrix. Now assume  $\det(\mathbf{I}_n) = 1$  and consider  $\det(\mathbf{I}_{n+1})$ . Expanding along the first row of  $\mathbf{I}_{n+1}$  gives  $\det(\mathbf{I}_{n+1}) = [\mathbf{I}_{n+1}]_{11}(-1)^{(1+1)} \det(\mathbf{M}_{11}) + [\mathbf{I}_{n+1}]_{12}(-1)^{(1+2)} \det(\mathbf{M}_{12}) + \dots + [\mathbf{I}_{n+1}]_{1(n+1)}(-1)^{(1+(n+1))} \det(\mathbf{M}_{1(n+1)}) = 1 \det(\mathbf{M}_{11})$  since the first element in row 1 of  $\mathbf{I}_{n+1}$  is unity and is the only nonzero element in that row. But  $\mathbf{M}_{11}$  is formed by deleting the first row and first column of  $\mathbf{I}_{n+1}$   $\mathbf{M}_{11} = \mathbf{I}_n$   $\det(\mathbf{M}_{11}) = \det(\mathbf{I}_n) = 1$   $\det(\mathbf{I}_{n+1}) = 1 \cdot 1 = 1$ . Thus, by induction, the determinant of any identity matrix equals one.

3.  $(1 - t^2)^3$

4. (a)  $0 = \det(\mathbf{A}^T - \mathbf{I}) = \det([\mathbf{A}^T - \mathbf{I}]^T) = \det(\mathbf{A} - \mathbf{I})$ . Thus  $(\mathbf{A}^T)$  iff  $(\mathbf{A})$ .

- (b)  $\mathbf{y}^* \mathbf{A} = \mu \mathbf{y}^*$   $\mathbf{y}^* (\mathbf{A} - \mu \mathbf{I}) = \mathbf{0}$   $\mathbf{0} = (\mathbf{A} - \mu \mathbf{I})^* \mathbf{y} = (\mathbf{A}^* - \bar{\mu} \mathbf{I}) \mathbf{y} = (\mathbf{A}^T - \bar{\mu} \mathbf{I}) \mathbf{y}$  since  $\mathbf{A}$  is real. Thus,  $\mu$  is a "left-hand" eigenvalue of  $\mathbf{A}$   $\bar{\mu}$   $(\mathbf{A}^T)$   $\bar{\mu}$   $(\mathbf{A})$  (using part a)  $\mu$   $(\mathbf{A})$  (using the hint).

- (c) Consider  $\langle \mathbf{y}, \mathbf{Ax} \rangle = \mathbf{y}^* (\mathbf{Ax}) = (\mathbf{y}^* \mathbf{A}) \mathbf{x} = {}_1 \mathbf{y}^* \mathbf{x} = {}_1 \langle \mathbf{y}, \mathbf{x} \rangle$ . But we also have  $\langle \mathbf{y}, \mathbf{Ax} \rangle = \langle \mathbf{y}, {}_2 \mathbf{x} \rangle = {}_2 \langle \mathbf{y}, \mathbf{x} \rangle$ . Thus  $\langle \mathbf{y}, \mathbf{Ax} \rangle = {}_1 \langle \mathbf{y}, \mathbf{x} \rangle = {}_2 \langle \mathbf{y}, \mathbf{x} \rangle$   $({}_1 - {}_2) \langle \mathbf{y}, \mathbf{x} \rangle = 0$   $\langle \mathbf{y}, \mathbf{x} \rangle = 0$   $\mathbf{y} \perp \mathbf{x}$ .