

COT 5615 Math for Intelligent Systems, Fall 2021

Midterm 2

1. (20 pts)

- (a) It can be expressed as n equations by squaring both sides and expand the norm-squared terms as

$$\mathbf{x}^T \mathbf{x} - 2\mathbf{y}_i^T \mathbf{x} + \mathbf{y}_i^T \mathbf{y}_i = \mathbf{x}^T \mathbf{x} - 2\mathbf{y}_{n+1}^T \mathbf{x} + \mathbf{y}_{n+1}^T \mathbf{y}_{n+1}, \quad i = 1, 2, \dots, n.$$

Then we have

$$\begin{bmatrix} (\mathbf{y}_1 - \mathbf{y}_{n+1})^T \\ (\mathbf{y}_2 - \mathbf{y}_{n+1})^T \\ \vdots \\ (\mathbf{y}_n - \mathbf{y}_{n+1})^T \end{bmatrix}_{n \times n} \mathbf{x} = \frac{1}{2} \begin{bmatrix} \|\mathbf{y}_1\|^2 - \|\mathbf{y}_{n+1}\|^2 \\ \|\mathbf{y}_2\|^2 - \|\mathbf{y}_{n+1}\|^2 \\ \vdots \\ \|\mathbf{y}_n\|^2 - \|\mathbf{y}_{n+1}\|^2 \end{bmatrix}_n.$$

Therefore, $\mathbf{A} = \begin{bmatrix} (\mathbf{y}_1 - \mathbf{y}_{n+1})^T \\ (\mathbf{y}_2 - \mathbf{y}_{n+1})^T \\ \vdots \\ (\mathbf{y}_n - \mathbf{y}_{n+1})^T \end{bmatrix}_{n \times n}$, $\mathbf{b} = \frac{1}{2} \begin{bmatrix} \|\mathbf{y}_1\|^2 - \|\mathbf{y}_{n+1}\|^2 \\ \|\mathbf{y}_2\|^2 - \|\mathbf{y}_{n+1}\|^2 \\ \vdots \\ \|\mathbf{y}_n\|^2 - \|\mathbf{y}_{n+1}\|^2 \end{bmatrix}_n$.

- (b) Define $\mathbf{Y} = [\mathbf{y}_1 \ \cdots \ \mathbf{y}_n]$, then $\mathbf{A}\mathbf{x} = \mathbf{b}$ can be expressed as

$$\mathbf{Y}^T \mathbf{x} - \mathbf{1} \mathbf{y}_{n+1}^T \mathbf{x} = \mathbf{b}.$$

If we introduce a new variable $z = -\mathbf{y}_{n+1}^T \mathbf{x}$, then $\mathbf{A}\mathbf{x} = \mathbf{b}$ can be expressed as

$$\mathbf{Y}^T \mathbf{x} + z \mathbf{1} = \mathbf{b}.$$

Together with $z = -\mathbf{y}_{n+1}^T \mathbf{x}$, they form a set of $n + 1$ equations with $n + 1$ unknowns:

$$\begin{bmatrix} \mathbf{Y}^T & \mathbf{1} \\ \mathbf{y}_{n+1}^T & 1 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ z \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ 0 \end{bmatrix}.$$

The solution to $\mathbf{A}\mathbf{x} = \mathbf{b}$ is unique if and only if the solution to the above equation is unique, in which case the $(n + 1) \times (n + 1)$ matrix

$$\begin{bmatrix} \mathbf{y}_1 & \mathbf{y}_2 & \cdots & \mathbf{y}_{n+1} \\ 1 & 1 & \cdots & 1 \end{bmatrix}$$

is invertible.

Sometimes it makes sense to increase the problem dimension. For example, if \mathbf{Y} is sparse, the resulting \mathbf{A} matrix may not be sparse. By solving the system with $n + 1$ variables, the sparsity pattern is kept, and the complexity could be a lot lower by using an appropriate algorithm.

2. (24 pts)

$$(a) \mathbf{A} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \sqrt{3} \\ 1 & -1 & 1 \\ -1 & -4 & 0 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ -1 \\ -2 \end{bmatrix}.$$

$$(b) \mathbf{A} = \begin{bmatrix} 0 & -6 \\ -4 & 3 \\ 1 & 8 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} -4 \\ 1 \\ 3 \end{bmatrix}.$$

$$(c) \mathbf{A} = \begin{bmatrix} -6\sqrt{2} & 0 \\ -4\sqrt{3} & 3\sqrt{3} \\ 2 & 16 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} -4\sqrt{2} \\ \sqrt{3} \\ 6 \end{bmatrix}.$$

$$(d) \mathbf{A} = \begin{bmatrix} \mathbf{B} \\ \mathbf{I}_{n \times n} \end{bmatrix}, \mathbf{b} = \begin{bmatrix} \mathbf{d} \\ \mathbf{0}_n \end{bmatrix}.$$

$$(e) \mathbf{A} = \begin{bmatrix} \mathbf{B} \\ \sqrt{2}\mathbf{F} \end{bmatrix}, \mathbf{b} = \begin{bmatrix} \mathbf{d} \\ \sqrt{2}\mathbf{g} \end{bmatrix}.$$

$$(f) \mathbf{A} = \begin{bmatrix} \mathbf{B} \\ \mathbf{D}^{\frac{1}{2}} \end{bmatrix}, \text{ where } \mathbf{D}^{\frac{1}{2}}\mathbf{D}^{\frac{1}{2}} = \mathbf{D}.$$

$$\mathbf{b} = \begin{bmatrix} \mathbf{d} \\ \mathbf{0}_n \end{bmatrix}.$$

3. (16 pts)

(a) False. In this case $\alpha = v < 0$.

(b) True. We will predict everything negative, so the false positive reate becomes zero.

(c) False. In this case $\alpha = v/2 < 0$.

(d) True. This is equivalent to $\hat{y} = \text{sign}(\mathbf{x}^T \mathbf{w} + 2v)$, i.e., $\alpha = -v > 0$.

4. (20 pts)

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}.$$

$$\mathbf{A}_{7 \times 4} = \begin{bmatrix} 1/2 & t_1/2 & t_1^2/2 & t_1^3/2 \\ 1/2 & t_2/2 & t_2^2/2 & t_2^3/2 \\ 1/2 & t_3/2 & t_3^2/2 & t_3^3/2 \\ 1/2 & t_4/2 & t_4^2/2 & t_4^3/2 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1/12 & 1/8 \end{bmatrix}, \mathbf{b}_7 = \begin{bmatrix} y_1/2 \\ y_2/2 \\ y_3/2 \\ y_4/2 \\ 0 \\ 0 \\ 0 \end{bmatrix}.$$

5. (20 pts)

- (a) \mathbf{A} is invertible if and only if $\mathbf{a} \neq 0$.

We use the property that a square matrix is invertible if and only if its columns are linearly independent, i.e., $\mathbf{A}\mathbf{x} = 0$ implies $\mathbf{x} = 0$. We partition $\mathbf{x} = (\mathbf{y}, z)$ where \mathbf{y} is a n -vector and z is a scalar. Then

$$\mathbf{A}\mathbf{x} = \begin{bmatrix} \mathbf{I} & \mathbf{a} \\ \mathbf{a}^T & 0 \end{bmatrix} \begin{bmatrix} \mathbf{y} \\ z \end{bmatrix} = \begin{bmatrix} \mathbf{y} + \mathbf{a}z \\ \mathbf{a}^T \mathbf{y} \end{bmatrix}.$$

Therefore $\mathbf{A}\mathbf{x} = 0$ is equivalent to

$$\mathbf{y} + \mathbf{a}z = 0, \quad \mathbf{a}^T \mathbf{y} = 0.$$

If $\mathbf{a} = 0$, a nonzero \mathbf{x} that satisfied this equation is with $\mathbf{y} = 0$ and $z = 1$. Therefore \mathbf{A} is not invertible if $\mathbf{a} = 0$. Suppose $\mathbf{a} \neq 0$, then substituting $\mathbf{y} = -\mathbf{a}z$ in the second equation gives $-z\|\mathbf{a}\|^2 = 0$. Since $\mathbf{a} \neq 0$, this implies $z = 0$. Then $\mathbf{y} = -\mathbf{a}z = 0$. This shows that \mathbf{A} is invertible.

- (b) The inverse is given by

$$\mathbf{A}^{-1} = \frac{1}{\|\mathbf{a}\|^2} \begin{bmatrix} \|\mathbf{a}\|^2 \mathbf{I} - \mathbf{a}\mathbf{a}^T & \mathbf{a} \\ \mathbf{a}^T & -1 \end{bmatrix}.$$

You can directly verify that their product is the $(n+1) \times (n+1)$ identity matrix.

Let's try to derive this result. The i th column of \mathbf{A}^{-1} is the solution to the equation $\mathbf{A}\mathbf{x} = \mathbf{e}_i$.

For the first n column, it can be partitioned as

$$\begin{bmatrix} \mathbf{I} & \mathbf{a} \\ \mathbf{a}^T & 0 \end{bmatrix} \begin{bmatrix} \mathbf{y} \\ z \end{bmatrix} = \begin{bmatrix} \mathbf{y} + \mathbf{a}z \\ \mathbf{a}^T \mathbf{y} \end{bmatrix} = \begin{bmatrix} \mathbf{e}_i \\ 0 \end{bmatrix}.$$

Multiply the first block with \mathbf{a}^T we get

$$\mathbf{a}^T(\mathbf{y} + \mathbf{a}z) = \mathbf{a}^T \mathbf{e}_i \Rightarrow z\|\mathbf{a}\|^2 = a_i.$$

So $z = a_i/\|\mathbf{a}\|^2$ and $\mathbf{y} = \mathbf{e}_i - (a_i/\|\mathbf{a}\|^2)\mathbf{a}$. The last column gives us

$$\begin{bmatrix} \mathbf{I} & \mathbf{a} \\ \mathbf{a}^T & 0 \end{bmatrix} \begin{bmatrix} \mathbf{y} \\ z \end{bmatrix} = \begin{bmatrix} \mathbf{y} + \mathbf{a}z \\ \mathbf{a}^T \mathbf{y} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

Substituting $\mathbf{y} = -z\mathbf{a}$ into $\mathbf{a}^T \mathbf{y} = 1$ gives us $z = -1/\|\mathbf{a}\|^2$, and so $\mathbf{y} = \mathbf{a}/\|\mathbf{a}\|^2$. Stacking them together

$$\begin{aligned} \mathbf{A}^{-1} &= \begin{bmatrix} \mathbf{e}_1 - (a_1/\|\mathbf{a}\|^2)\mathbf{a} & \cdots & \mathbf{e}_n - (a_n/\|\mathbf{a}\|^2)\mathbf{a} & \mathbf{a}/\|\mathbf{a}\|^2 \\ a_1/\|\mathbf{a}\|^2 & \cdots & a_n/\|\mathbf{a}\|^2 & -1/\|\mathbf{a}\|^2 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{I} & 0 \\ 0 & 0 \end{bmatrix} - \frac{1}{\|\mathbf{a}\|^2} \begin{bmatrix} a_1\mathbf{a} & \cdots & a_n\mathbf{a} & -\mathbf{a} \\ -a_1 & \cdots & -a_n & 1 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{I} & 0 \\ 0 & 0 \end{bmatrix} - \frac{1}{\|\mathbf{a}\|^2} \begin{bmatrix} \mathbf{a}\mathbf{a}^T & -\mathbf{a} \\ -\mathbf{a}^T & 1 \end{bmatrix} \\ &= \frac{1}{\|\mathbf{a}\|^2} \begin{bmatrix} \|\mathbf{a}\|^2 \mathbf{I} - \mathbf{a}\mathbf{a}^T & \mathbf{a} \\ \mathbf{a}^T & -1 \end{bmatrix}. \end{aligned}$$