

1) (a) (i) True. We know $I = I^T$, since $AA^{-1} = I \Rightarrow AA^{-1} = (AA^{-1})^T$, since $(AB)^T = B^T A^T \Rightarrow AA^{-1} = (A^{-1})^T A^T$

$$\Rightarrow AA^{-1} = A^T A = (A^{-1})^T A^T, \text{ since } A = A^T \Rightarrow A^{-1} A = (A^{-1})^T A. \Rightarrow A^{-1} A A^{-1} = (A^{-1})^T A A^{-1} \Rightarrow A^{-1} = (A^{-1})^T$$

(ii) True. Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$. if $\det(A) = 1$, $\begin{bmatrix} a & c \\ b & d \end{bmatrix} = A^T = A^{-1} = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$.

$$\Rightarrow d = a, b = -c \Rightarrow A \text{ has the form } \begin{bmatrix} a & -c \\ c & a \end{bmatrix}, \text{ and } a^2 + c^2 = 1$$

$$\Rightarrow \text{We can write } A \text{ as } \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \text{ if } \det(A) = 1, \begin{bmatrix} a & c \\ b & d \end{bmatrix} = \begin{bmatrix} -d & b \\ c & -a \end{bmatrix}$$

$$\Rightarrow a = -d, b = c \Rightarrow A \text{ has the form } \begin{bmatrix} a & c \\ c & -a \end{bmatrix}, \text{ and } -(a^2 + c^2) = -1 \Rightarrow A = \begin{bmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{bmatrix}$$

(iii) eigenvalues of A are $-1, -4.899, 4.899 \Rightarrow A$ is not positive-definite

$\Rightarrow A$ cannot be decomposed as CC^T , the answer is false. #

2)

(a)

(i) $E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$, since $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy \Rightarrow E[X] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{X,Y}(x,y) dx dy$

since $f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} \Rightarrow E[X] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \cdot f_{X|Y}(x|y) \cdot f_Y(y) dx dy = E_Y[E_X[X|Y]]$ #

(ii) $E[I(X \in C)] = \int_X I(X \in C) dP = \int_X 1 \cdot dP = P(X \in C)$ #

(iii) $\text{Var}[X] = E[X^2] - (E[X])^2 = E_Y[E_X[X^2|Y]] - (E_Y[E_X[X|Y]])^2$

$$= E_Y[E_X[X^2|Y]] - E_Y[(E_X[X|Y])^2] + E_Y[(E_X[X|Y])^2] - (E_Y[E_X[X|Y]])^2$$

$$= E_Y[\text{Var}_X[X|Y]] + \text{Var}_Y[E_X[X|Y]]$$
 #

(iv) $E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_{X,Y}(x,y) dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xf_X(x) y f_Y(y) dx dy = \int_{-\infty}^{\infty} xf_X(x) dx \int_{-\infty}^{\infty} y f_Y(y) dy = E[X]E[Y]$ #

$$2) (a) (i) E[XY] = \sum xy P_r(X=x, Y=y) = P_r(X=1, Y=1).$$

$$E[X] = \sum x P_r(X=x, Y=y) = P_r(X=1, Y=0) + P_r(X=1, Y=1) = P_r(X=1)$$

$$E[Y] = \sum y P_r(X=x, Y=y) = P_r(X=1, Y=1) + P_r(X=0, Y=1) = P_r(Y=1)$$

$$\Rightarrow P_r(X=1, Y=1) = E[XY] = E[X]E[Y] = P_r(X=1) \cdot P_r(Y=1) \Rightarrow \text{independent}$$

$$\text{and for } X=x, Y=1-y, E[XY] = \sum x(1-y) P_r(X=x, Y=y) = P_r(X=1, Y=0)$$

$$E[X] = \sum x P_r(X=x, Y=y) = P_r(X=1), E[Y] = \sum (1-y) P_r(X=x, Y=y) = P_r(Y=0)$$

$$\Rightarrow P_r(X=1, Y=0) = E[XY] = E[X]E[Y] = P_r(X=1) P_r(Y=0) \rightarrow \text{independent.}$$

$$\text{for } X=1-X, Y=1-Y \text{ and } X=1-X, Y=y, \text{ the same as above.}$$

$$\Rightarrow E[XY] = E[X]E[Y] \rightarrow X, Y \text{ are independent.} \#$$

(b)

$$(i) P(H=h) = \sum_{i=1}^d P(H=h, D=d_i) \Rightarrow P(H=h, D=d) \leq P(H=h) \#$$

$$(ii) P(H=h|D=d) = \frac{P(H=h, D=d)}{P(D=d)} = a, 0 \leq a \leq 1 \Rightarrow \text{depends} \#$$

$$(iii) \frac{P(H=h, D=d)}{P(D=d)} \geq P(H=h, D=d) = P(D=d|H=h) \cdot P(H=h) \Rightarrow \geq \#$$

3)

$$(a) A = U \Lambda U^T, X^T A X = \sum_{i=1}^d X^T \cdot \lambda_i u_i u_i^T \cdot X = \sum_{i=1}^d \lambda_i (u_i^T X)^T \cdot u_i X.$$

$$\because (u_i^T X)^T \cdot u_i X \geq 0, \Rightarrow \lambda_i \geq 0 \Rightarrow A = \text{PSD.}$$

$$\text{If } A \text{ is PSD, then } X^T A X = \sum_{i=1}^d \lambda_i (u_i^T X)^T \cdot u_i X \geq 0 \text{ for } \lambda_i \geq 0 \#$$

$$(b) \text{ Same as (a) with } \lambda_i > 0. \#$$

4)

$$(a) f(x) = a^T x + b. \Rightarrow f(tx + (1-t)y) = a^T tx + a^T (1-t)y + b$$

$$= ta^T x + (1-t)a^T y + tb + (1-t)b \Rightarrow f(x) \text{ is convex and concave.}$$

$$f(x) \text{ is not strictly convex.}$$

4) (b) Let x_1, x_2 be global minimizer. $x' = \frac{x_1 + x_2}{2} \Rightarrow f(x') > f(x_1) = f(x_2)$
 but $f(\frac{x_1 + x_2}{2}) < \frac{1}{2} f(x_1) + \frac{1}{2} f(x_2) \Rightarrow$ Contradiction, f has at most one global minimizer.

(c) $\nabla f(x^*) = 0$. Let $x = x^* + th$, where $t \in \mathbb{R}, h \in \mathbb{R}^n$

$$f(x^* + th) = f(x^*) + \langle \nabla f(x^*), th \rangle + \frac{1}{2} \langle th, \nabla^2 f(x^*) th \rangle + o(\|th\|^2)$$

$$\Rightarrow 0 < \frac{f(x^* + th) - f(x^*)}{\|th\|^2} = \frac{(th)^T \nabla^2 f(x^*) (th)}{2\|th\|^2} + \frac{o(\|th\|^2)}{\|th\|^2}$$

$$\Rightarrow 0 < (th)^T \nabla^2 f(x^*) (th), \text{ where } th \in \mathbb{R}^n \Rightarrow \underline{t^2 h^T \nabla^2 f(x^*) h > 0} \#$$

(d) f is convex $\Rightarrow f(x) \geq f(y) + \langle \nabla f(y), x - y \rangle \Rightarrow \langle x - y, \nabla^2 f(y)(x - y) \rangle \geq 0$

$$\Rightarrow (x - y)^T \nabla^2 f(y) (x - y) \geq 0 \Rightarrow \underline{\nabla^2 f(y) \text{ is PSD}} \#$$

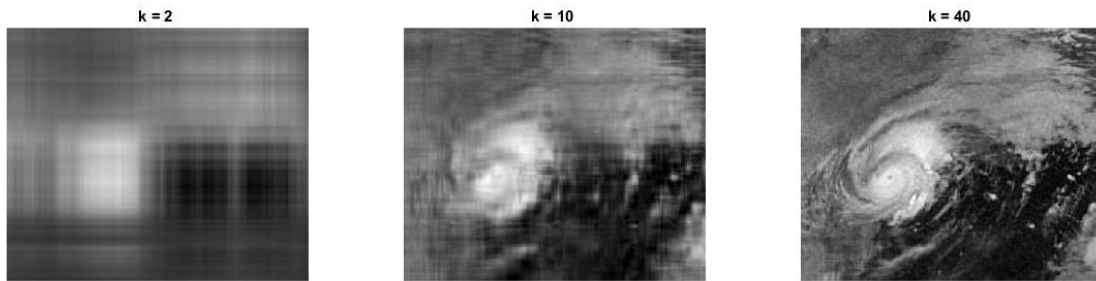
$$\nabla^2 f(y) \text{ is PSD, } \langle x - y, \nabla^2 f(y + t(x - y))(x - y) \rangle \geq 0.$$

$$\Rightarrow f(x) \geq f(y) + \langle \nabla f(y), x - y \rangle \Rightarrow \underline{f \text{ is convex}} \#.$$

(e) $f(x) = \frac{1}{2} x^T A x + b^T x + c$. $\nabla^2 f(x) = A$. f is $\begin{cases} \text{convex} \\ \text{strictly convex} \end{cases}$ if A is $\begin{matrix} \text{PSD} \\ \text{PD} \end{matrix}$. #

5.

(a)



For $k \in \{2, 10, 40\}$,
$$\frac{\|X - \bar{X}\|_F}{\|X\|_F} = [0.2815, 0.1587, 0.0837]$$

- Matlab Code

```
1 X = double(rgb2gray(imread('harvey-saturday-goes7am.jpg')));
2 [U,S,V] = svd(X);
3 k = [2 10 40];
4 X_approximate = zeros(size(X,1), size(X,2), size(k,2));
5 error = [];
6 for i = 1 : size(k,2)
7     X_approximate(:, :, i) = U(:, 1:k(i))*S(1:k(i), 1:k(i))*transpose(V(:, 1:k(i)));
8     error = [error norm(X - X_approximate(:, :, i), 'fro')/norm(X, 'fro')];
9     subplot(1, size(k,2), i);
10    imshow(uint8(X_approximate(:, :, i)));
11    txt = sprintf('k = %d ', k(i));
12    title(txt);
13 end
14 disp(error);
```

(b) The numbers (n) we need to describe the approximation

For $k = 2$, $n = 1296*2 + 2 + 2*1548$

For $k = 10$, $n = 1296*10 + 10 + 10*1548$

For $k = 40$, $n = 1296*40 + 40 + 40*1548$