

Lecture 9

Longhai Li, October 7, 2021

plans:

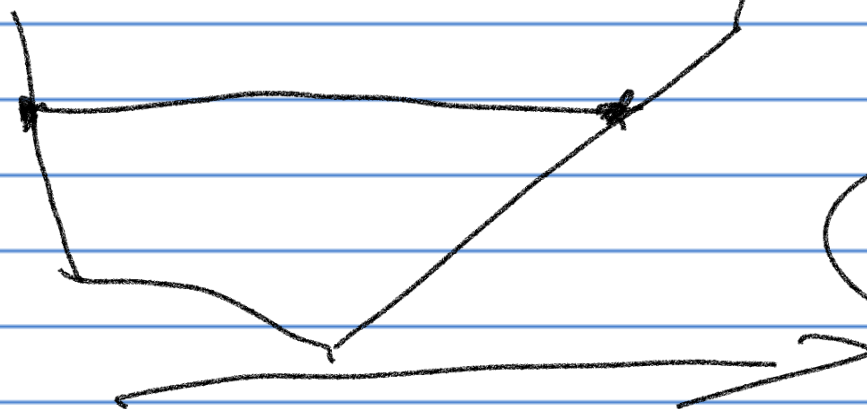
- 1) Jensen Inequality Sec 1.10
- 2) Joint distribution Sec 2.1

Jensen Inequality.

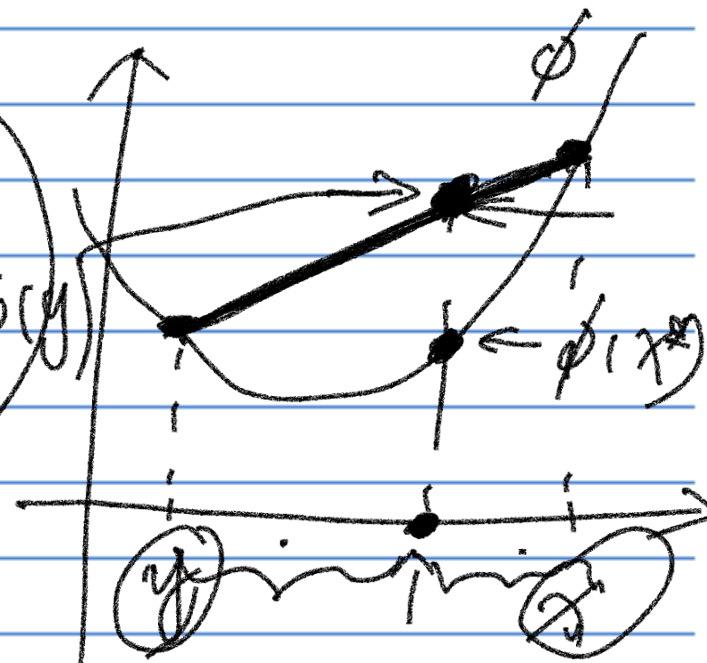
Convex functions

$$\phi(r\lambda + (1-r)\mu) \leq$$

$$r\phi(\lambda) + (1-r)\phi(\mu)$$



$$r\phi(\lambda) + (1-r)\phi(\mu)$$



$$r \in (0, 1)$$

$$x^* = r\lambda + (1-r)\mu$$

$$r \in (0, 1)$$

$\phi'(x)$ exists

ϕ is convex if

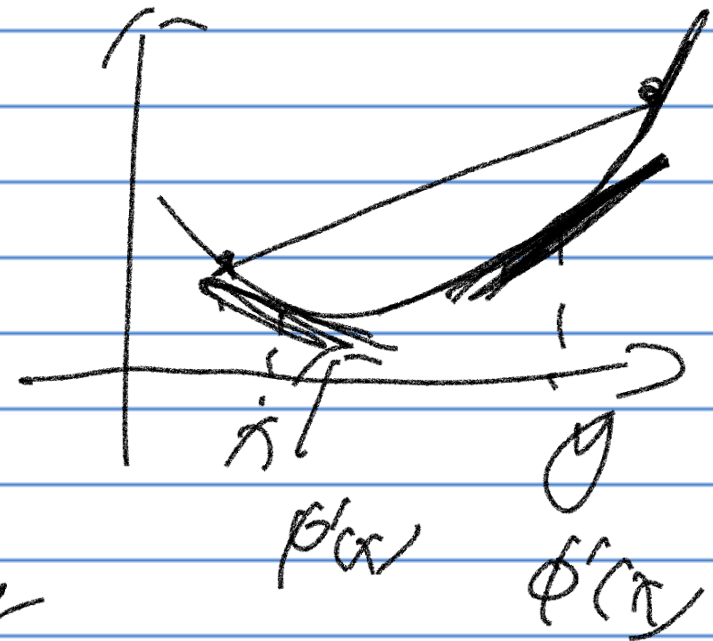
$\phi'(x)$ is non-decreasing

$\phi'(x) \leq \phi'(y)$ if $x \leq y$

$\phi''(x)$ exists.

ϕ is convex if

$\phi''(x) \geq 0$



Example:

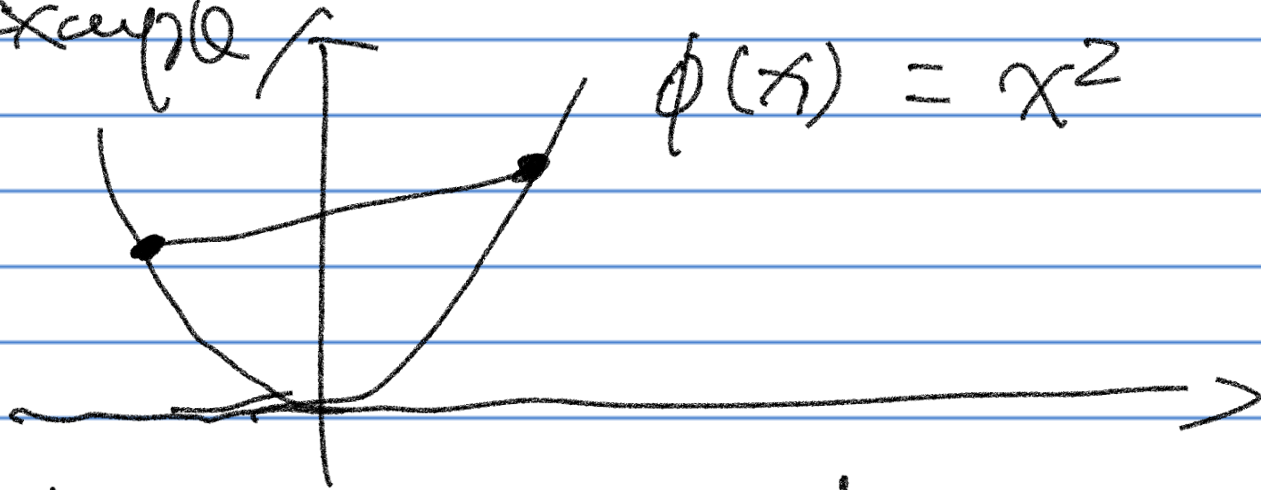


$$\phi(x) = -\log(x)$$

$$\phi'(x) = -\frac{1}{x} \quad \text{increasing}$$

$$\phi''(x) = \frac{1}{x^2} \geq 0 \quad \text{positive.}$$

Example



$$\phi(x) = x^2$$

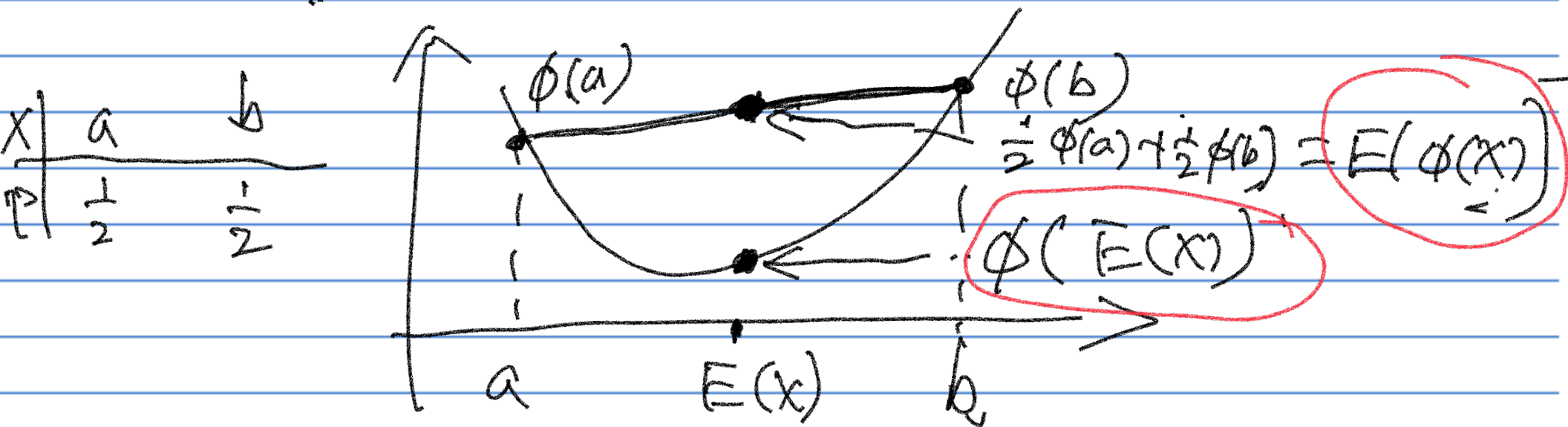
$$\phi'(x) = 2x, \text{ increasing}$$

$$\phi''(x) = 2, \neq 0$$

Jensen Inequality

ϕ is a convex function.

$$E(\phi(X)) \geq \phi(E(X))$$



PF: Assuming $\phi'(u)$ exists, $u = E(X)$

$$\phi(x) \geq \phi(u) + \phi'(u)(x-u) \text{ for all } x$$

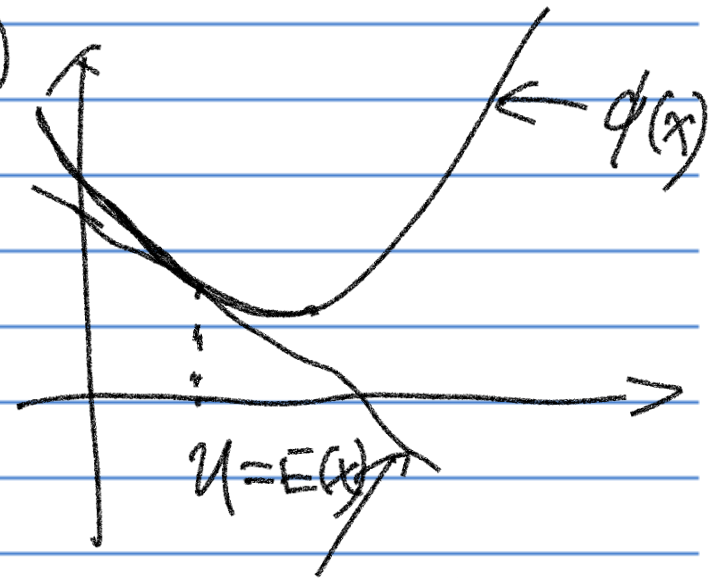
$$\phi(x) \geq \phi(u) + \phi'(u)(x-u) \text{ for all } x$$

$$E(\phi(X)) \geq \phi(u) + \phi'(u) E(X-u)$$

$$= \phi(u) + \phi'(u) \underbrace{(E(X)-u)}_0$$

$$= \phi(u)$$

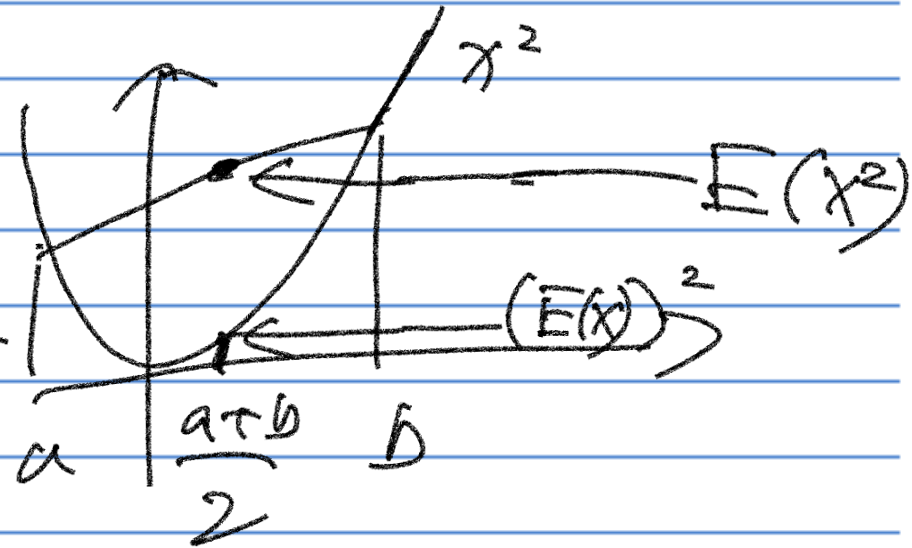
$$= \phi(E(X)).$$



$$\phi(x) = \phi(u) + \phi'(u)(x-u)$$

Examples:

1) $\phi(x) = x^2$



$$E(X^2) \geq [E(X)]^2$$

$$E(X^2) \geq (E(X))^2$$

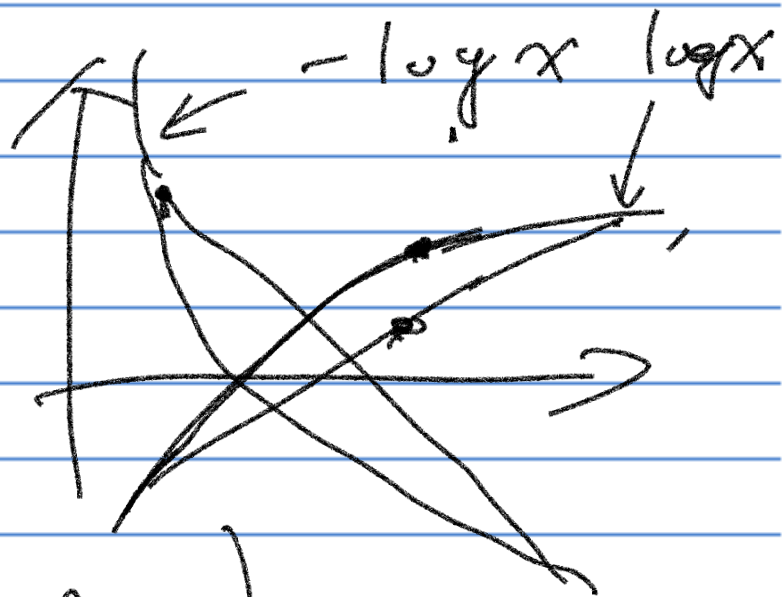
$$V(X) = E(X^2) - (E(X))^2 \geq 0$$

2)

$$E(-\log f(x))$$

$$\geq -\log(E(x))$$

$$E(\log(x)) \leq \log(E(x))$$



suppose $X \sim$

x	a_1	a_2	\dots	a_n
p	$\frac{1}{n}$	$\frac{1}{n}$		$\frac{1}{n}$

$$a_i > 0$$

$$E(\log(x)) \neq \log(E(x))$$

$$\log(x) \sim \left| \begin{array}{cccc} \log a_1 & \dots & \log a_n \\ \frac{1}{n} & & \frac{1}{n} \end{array} \right.$$

$$\begin{aligned} E[\log(x)] &= \frac{\sum_{i=1}^n \log a_i}{n} && \log x + \log y \\ &= \frac{\log\left(\prod_{i=1}^n a_i\right)}{n} && = \log(x \cdot y) \\ &= \log\left(\left(\prod_{i=1}^n a_i\right)^{\frac{1}{n}}\right) && = \log\left(\prod_{i=1}^n a_i\right)^{\frac{1}{n}} \end{aligned}$$

$$\log(E(X))$$

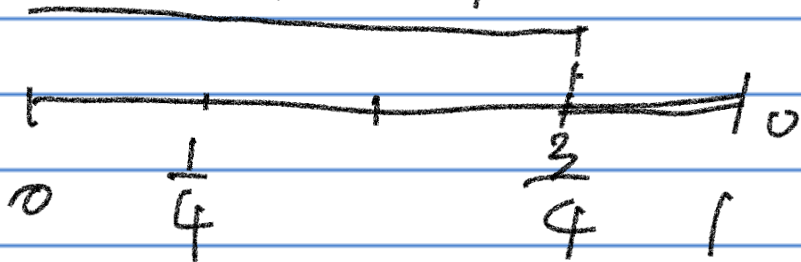
$$= \log\left(\frac{a_1 + \dots + a_n}{n}\right)$$

$$\log\left(\left[\prod_{i=1}^n a_i\right]^{\frac{1}{n}}\right) \leq \log\left(\frac{\sum a_i}{n}\right)$$

$$\left[\prod_{i=1}^n a_i\right]^{\frac{1}{n}} \leq \frac{\sum_{i=1}^n a_i}{n}$$

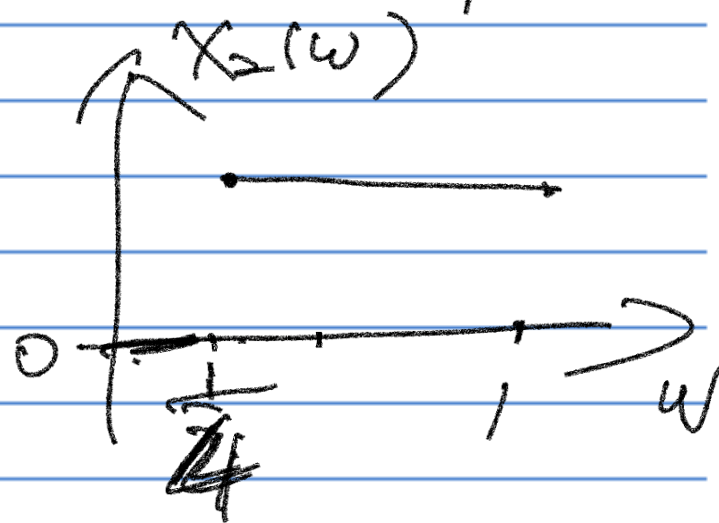
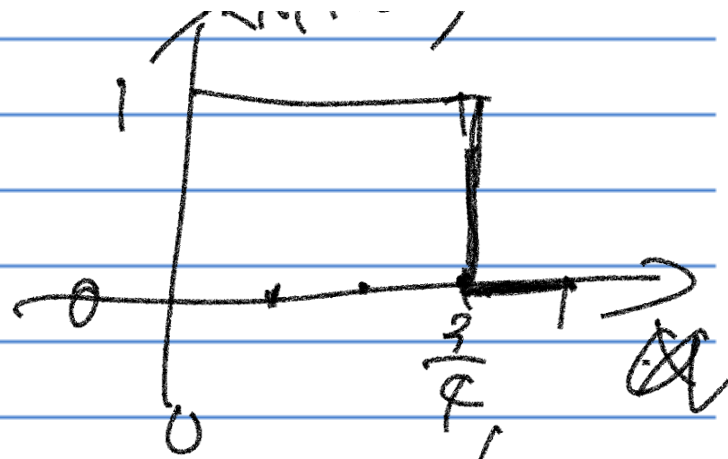
$$G.M. \leq A.M.$$

$$W \sim \text{Unif}([0, 1])$$



$$X_1(W) = \mathbb{1}(W \leq \frac{3}{4})$$

$$X_2(W) = \mathbb{1}(W \geq \frac{1}{4})$$

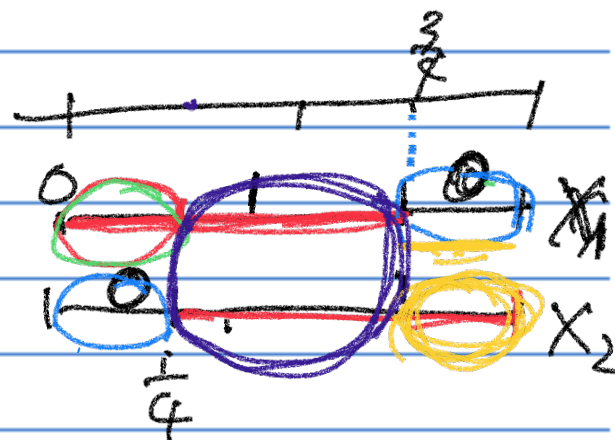


Joint P.M.F.

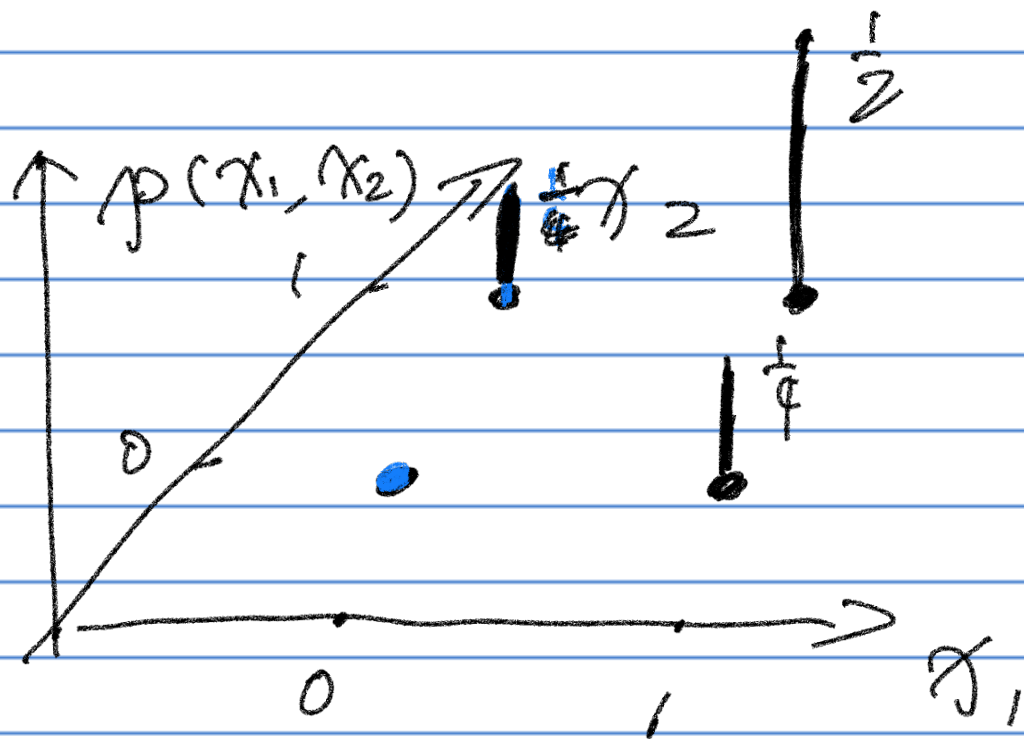
$$P(x_1, x_2) = P(X_1 = x_1, X_2 = x_2)$$

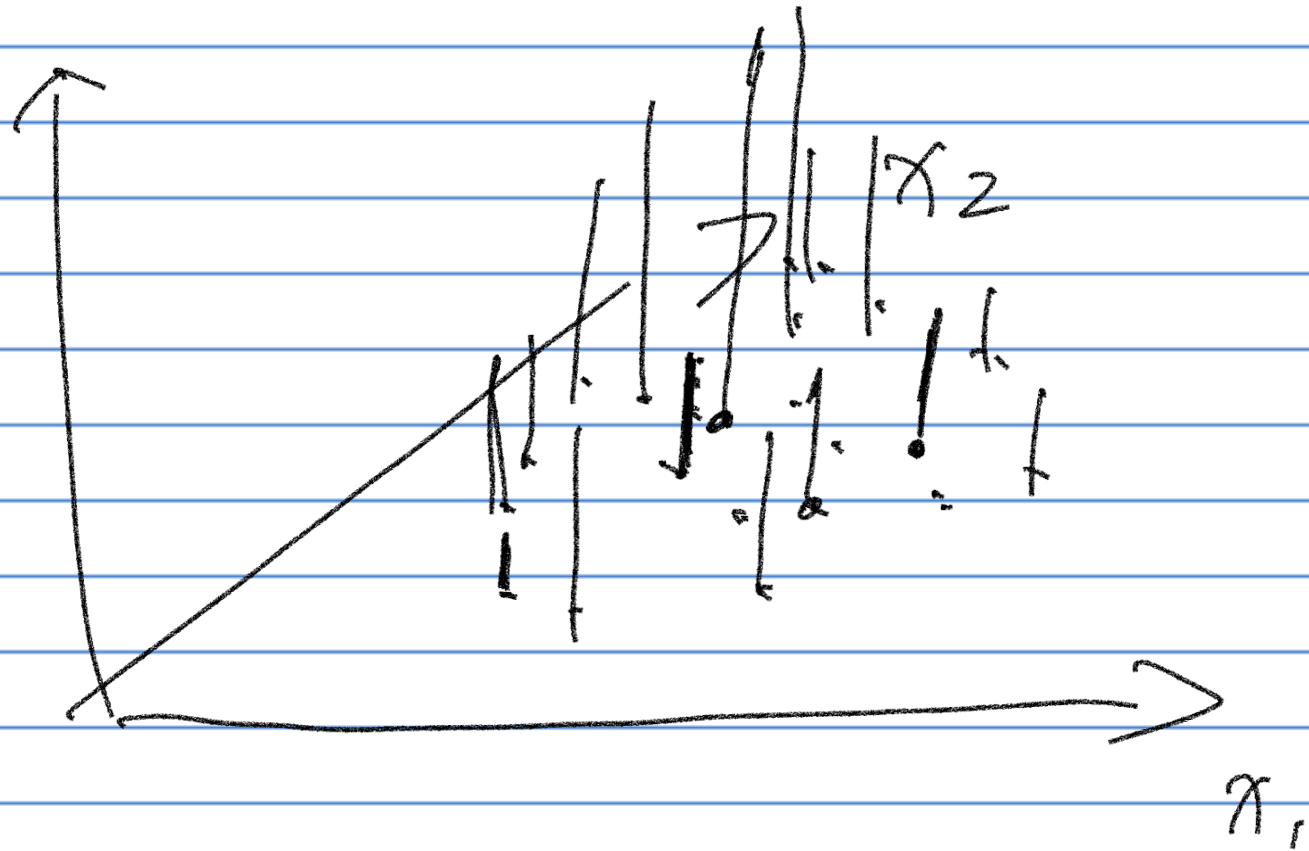
for all possible values that X_1 & X_2 can take.

$x_2 \backslash x_1$	0	1
0	0	$\frac{1}{4}$
1	$\frac{1}{4}$	$\frac{1}{2}$



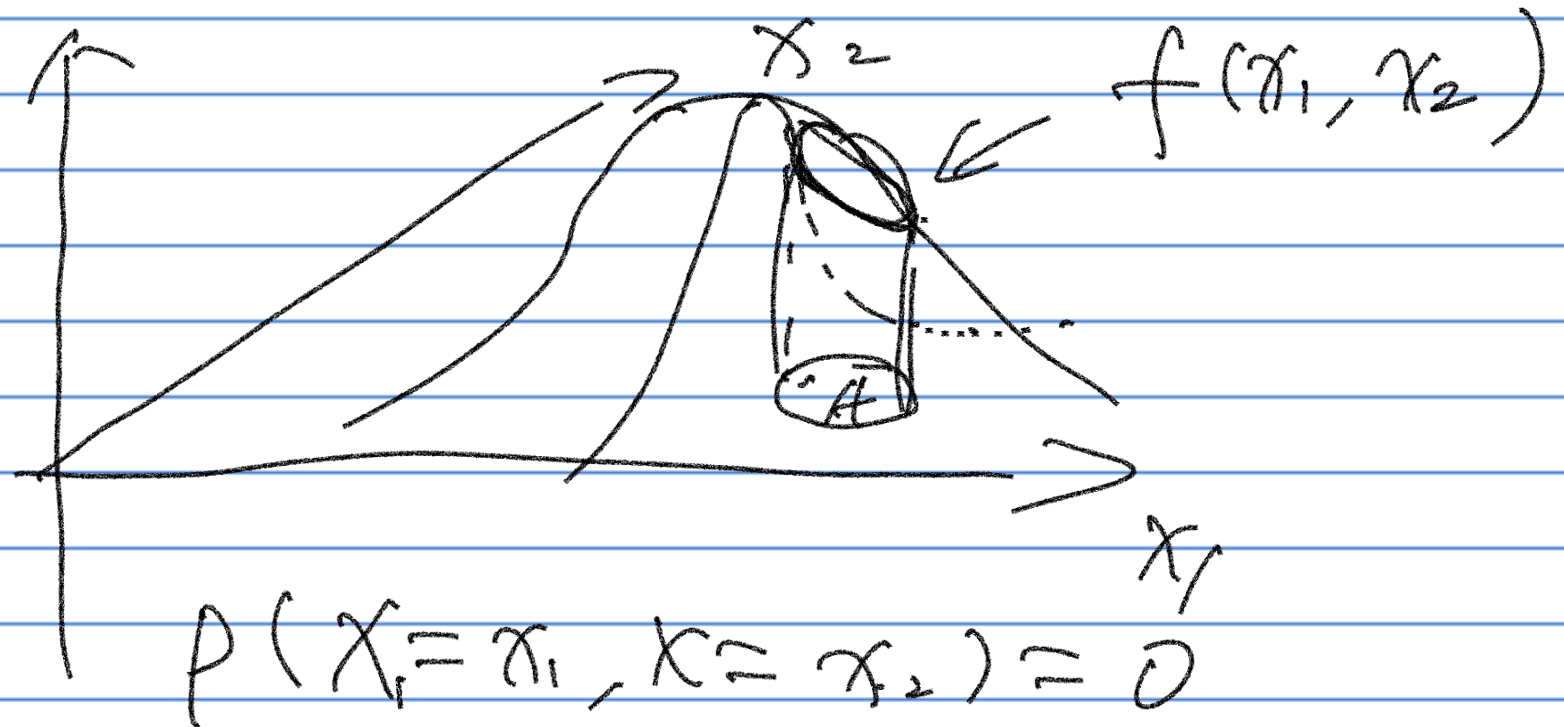
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Joint P.D.F.

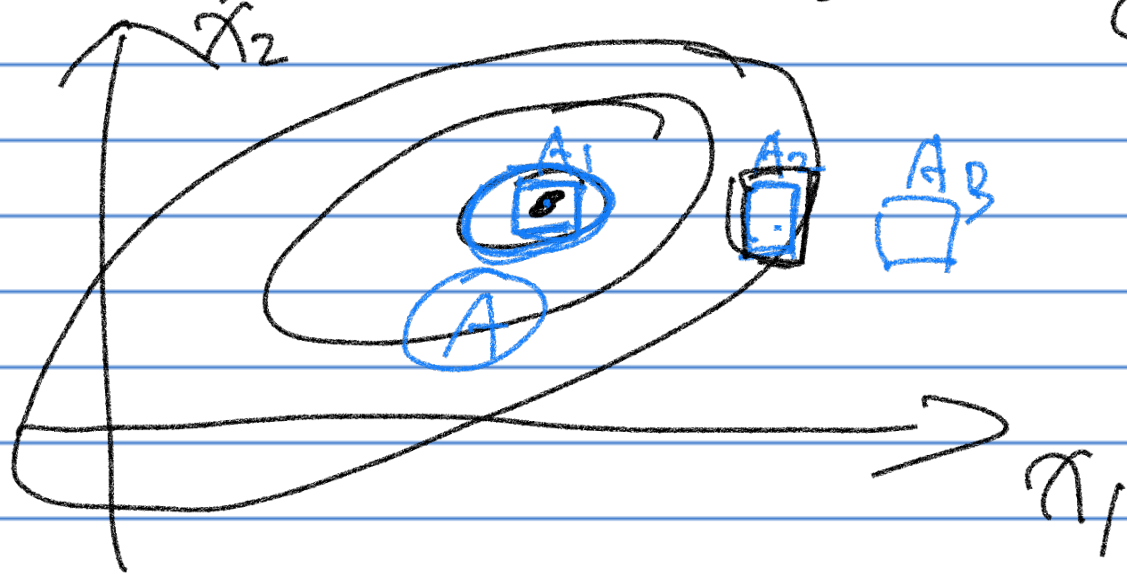
X_1, X_2 are both continuous



We say $f(x_1, x_2)$ is a joint P.D.F. of (X_1, X_2) if

$$P((X_1, X_2) \in A) = \iint_A f(x_1, x_2) dx_1 dx_2$$

Contour of $f(x_1, x_2)$



Marginal P.M.F. & P.D.F.

Example

$x_2 \backslash x_1$	0	1	$P_{X_2}(x_2)$	$P_{X_2}(x_2)$
0	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	$= \sum_{x_1} p(x_1, x_2)$
1	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{3}{4}$	
$P_{X_1}(x_1)$	$\frac{1}{4}$	$\frac{3}{4}$		

$$P_{X_1}(0) = p(0, 0) + p(0, 1)$$
$$= \frac{1}{4}$$

$$P_{X_1}(x_1) = \sum_{x_2} p(x_1, x_2)$$

Marginal P.D.F.

Suppose $f(x_1, x_2)$ is a joint P.D.F.

of (x_1, x_2)

$$f_{x_1}(x_1) = \int_{-\infty}^{+\infty} f(x_1, x_2) dx_2$$

$$f_{x_2}(x_2) = \int_{-\infty}^{+\infty} f(x_1, x_2) dx_1.$$

