

# Lecture 10 and 11

Longhai Li, October 12 and 14, 2021

plaats:

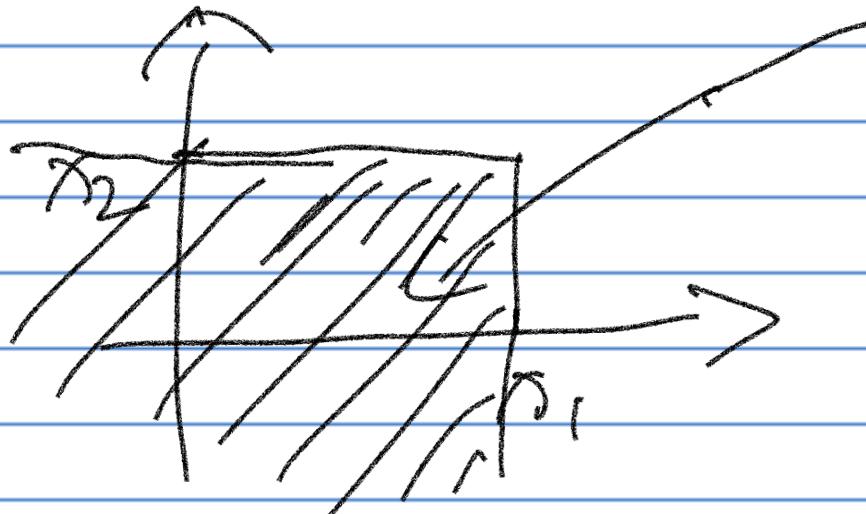
1. Sec 2.1.1. expectation.

2. Sec 2.1.2. transformation.

Joint C.D.F.

$$F(x_1, x_2) = P(X_1 \leq x_1, X_2 \leq x_2)$$

$$= \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} f(t_1, t_2) dt_1 dt_2$$



Expectation:

Def:  $(X_1, X_2)$  has a P.M.F.  $P(\pi_1, \pi_2)$

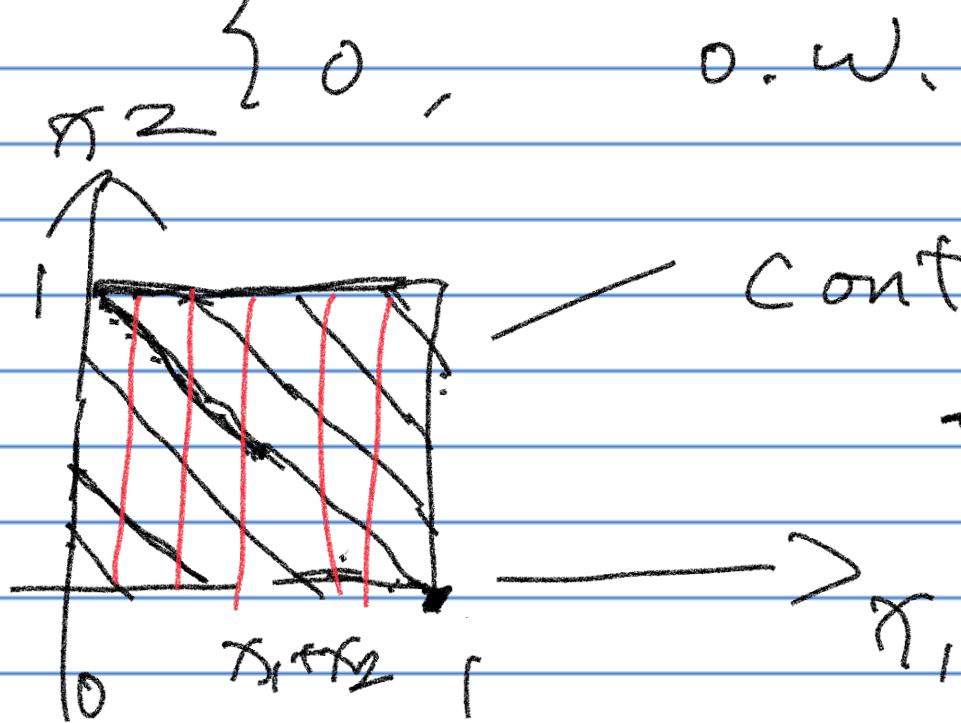
$$E(g(X_1, X_2)) = \sum_{\pi_1, \pi_2} g(\pi_1, \pi_2) P(\pi_1, \pi_2)$$

Def:  $(X_1, X_2)$  has a D.O.F.  $f(\pi_1, \pi_2)$

$$E(g(X_1, X_2)) = \iint g(\pi_1, \pi_2) f(\pi_1, \pi_2) d\pi_1 d\pi_2$$

Example:

$$f(x_1, x_2) = \begin{cases} x_1 + x_2, & \text{if } (x_1, x_2) \in [0, 1] \times [0, 1] \\ 0, & \text{o.w.} \end{cases}$$



Contour of  
 $f(x_1, x_2)$

$$g_1(x_1, x_2) = x_1 \quad E(g_1(x_1, x_2))$$

$$E(g_1(x_1, x_2)) = E(x_1)$$

$$= \int_0^1 \int_0^1 x_1 f(x_1, x_2) dx_1 dx_2$$

$$= \int_0^1 \int_0^1 x_1 \cdot (x_1 + x_2) dx_1 dx_2.$$

$$= \int_0^1 \int_0^1 x_1^2 dx_1 dx_2 + \underline{\int_1^1 \int_0^1 x_1 x_2 dx_1 dx_2}$$

$$= \frac{1}{3} + \frac{1}{4} = \frac{2}{12}$$

where

$$\begin{aligned} & \int_0^1 \left[ \int_0^1 (x_1^2) dx_2 \right] dx_1 \\ &= \int_0^1 x_1^2 \times 1 dx_1 = \left( \frac{x_1^3}{3} \right) \Big|_0^1 = \frac{1}{3} \end{aligned}$$

$$= \int_0^1 \left[ \int_0^1 x_1 x_2 dx_1 \right] dx_2$$

$$= \int_0^1 \left[ x_2 \int_0^1 x_1 dx_1 \right] dx_2$$

$$= \int_0^1 \left[ x_2 \cdot \frac{x_1^2}{2} \Big|_0^1 \right] dx_2$$

$$= \int_0^1 \frac{1}{2} x_2 dx_2 = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

$E(X_1) = ?$  To find  $f_{X_1}(x)$  first?

For  $x_1 \in [0, 1]$

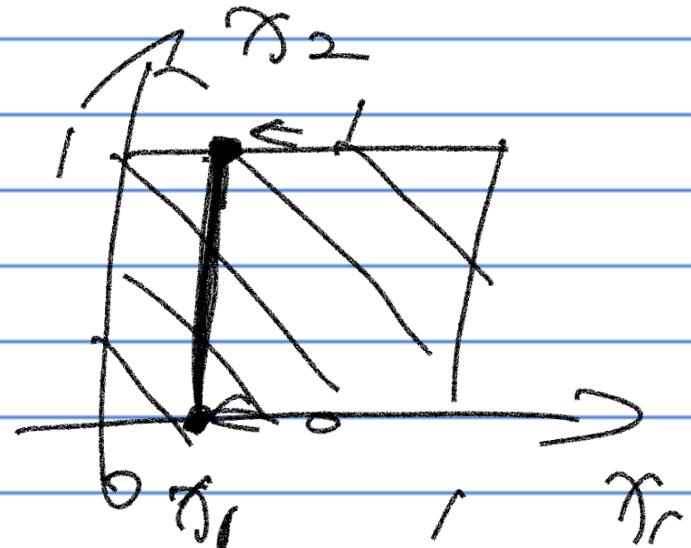
$$f_{X_1}(x_1)$$

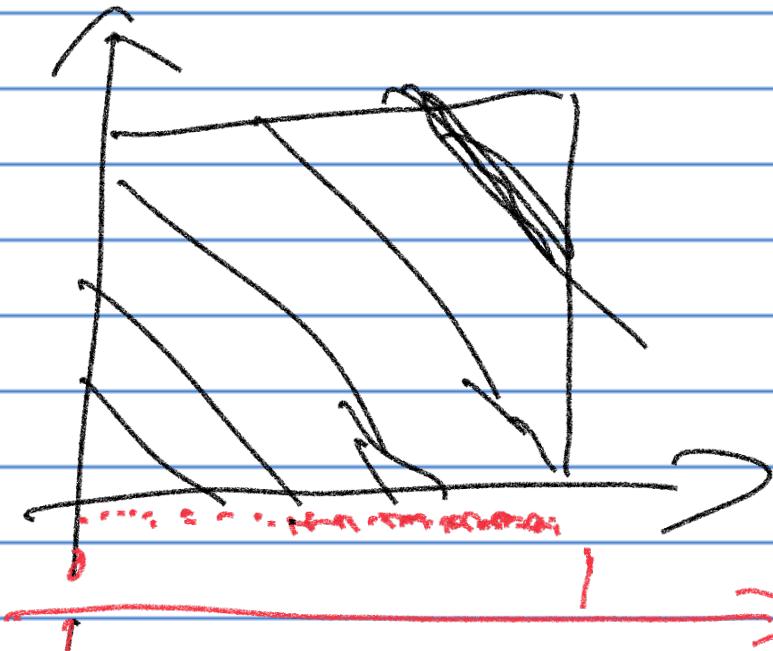
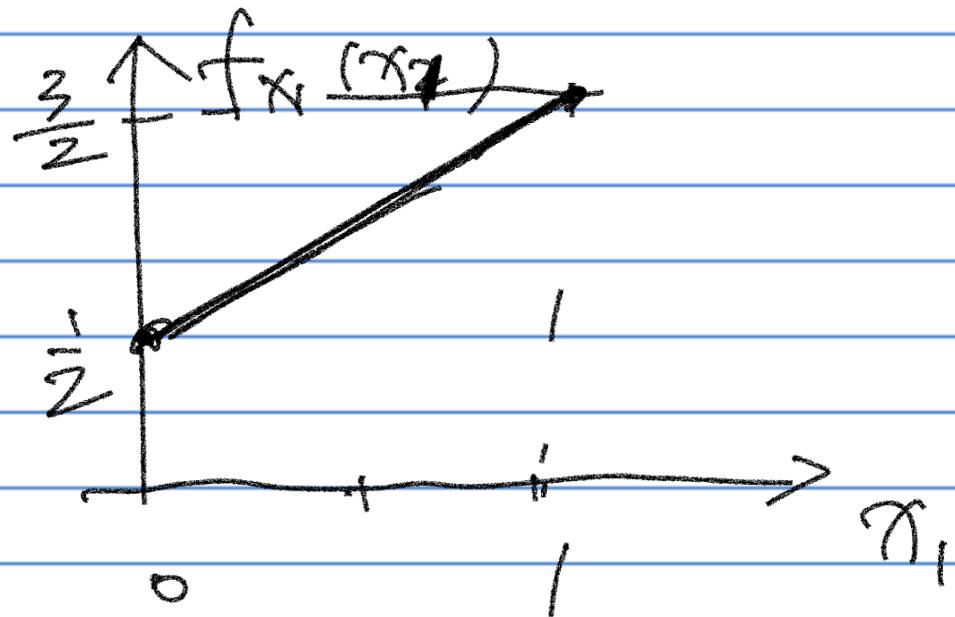
$$= \int_{-\infty}^{+\infty} f(x_1, x_2) dx_2$$

$$= \int_0^1 (x_1 + x_2) dx_2$$

$$= \int_0^1 x_1 dx_2 + \int_0^1 x_2 dx_2$$

$$= x_1 \cdot 1 + \frac{1}{2} = x_1 + \frac{1}{2}$$





$$f_{X_1}(x_1) = x_1 + \frac{1}{2}$$

$$\begin{aligned} E(X_1) &= \int_0^1 x_1 \cdot \left(x_1 + \frac{1}{2}\right) dx \\ &= \frac{1}{3} + \frac{1}{2} \cdot \frac{1}{2} = \frac{7}{12} \end{aligned}$$

$$\int \int x_1 f(x_1, x_2) dx_1 dx_2$$

$$= \int \left[ \int x_1 f(x_1, x_2) dx_2 \right] dx_1$$

$$= \int x_1 \cdot \left[ \int f(x_1, x_2) dx_2 \right] dx_1$$

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$$f_{x_1}(x_1)$$

Example:

$$f_{x_1, x_2}(x_1, x_2) =$$

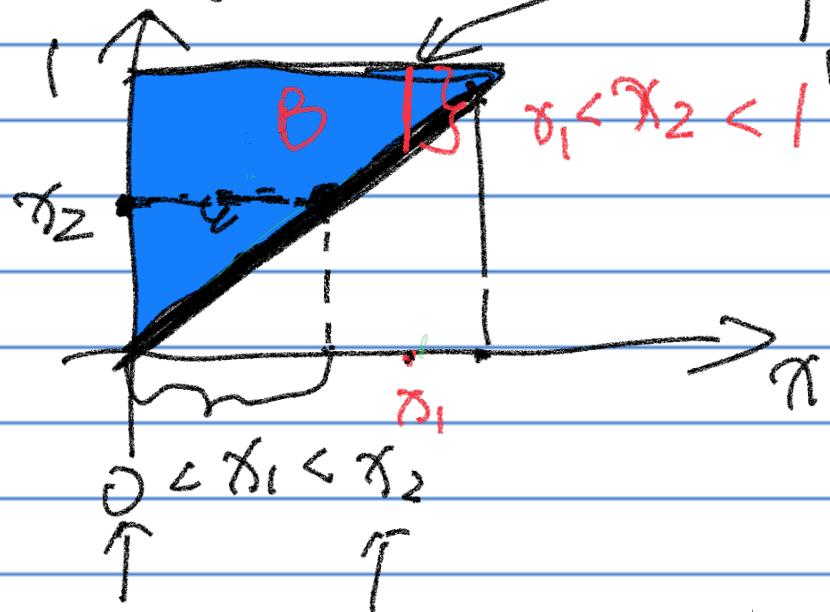
$$\int \int 8x_1 x_2^2 \cdot 8x_1 \cdot x_2 dx_1 dx_2$$

$$\phi < x_1 < x_2 < 1$$

$$E(x_1 x_2^2) = \int \int \int 8x_1 x_2^2 \cdot 8x_1 \cdot x_2 dx_1 dx_2$$

$$= \int_{0}^{1} \left[ \int_{0}^{x_1} [8x_1^2 x_2^3 dx_2] dx_1 \right]$$

$$= \int_{0}^{1} 8x_1^2 \cdot \left[ \frac{1}{4} x_2^4 \right]_0^{x_1} dx_1$$
$$= \int_{0}^{1} 8x_1^2 \cdot \frac{1}{4} (1 - x_1^4) dx_1$$
$$= \int_{0}^{1} 2x_1^2 dx_1 - \int_{0}^{1} 2x_1^6 dx_1$$
$$= \frac{2}{3} - 2 \times \frac{1}{7} = \frac{8}{21}$$

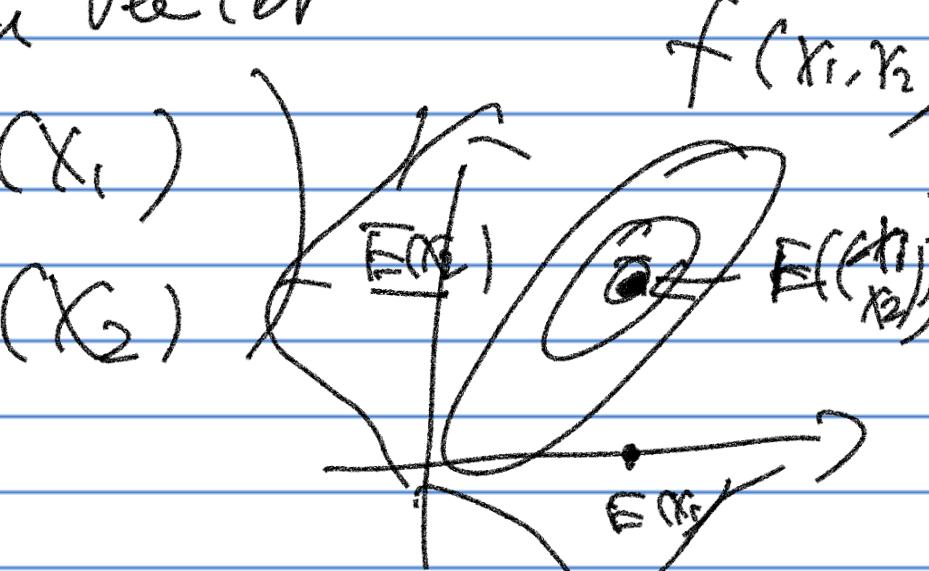


M.G.F. of Random Vector

Def.:

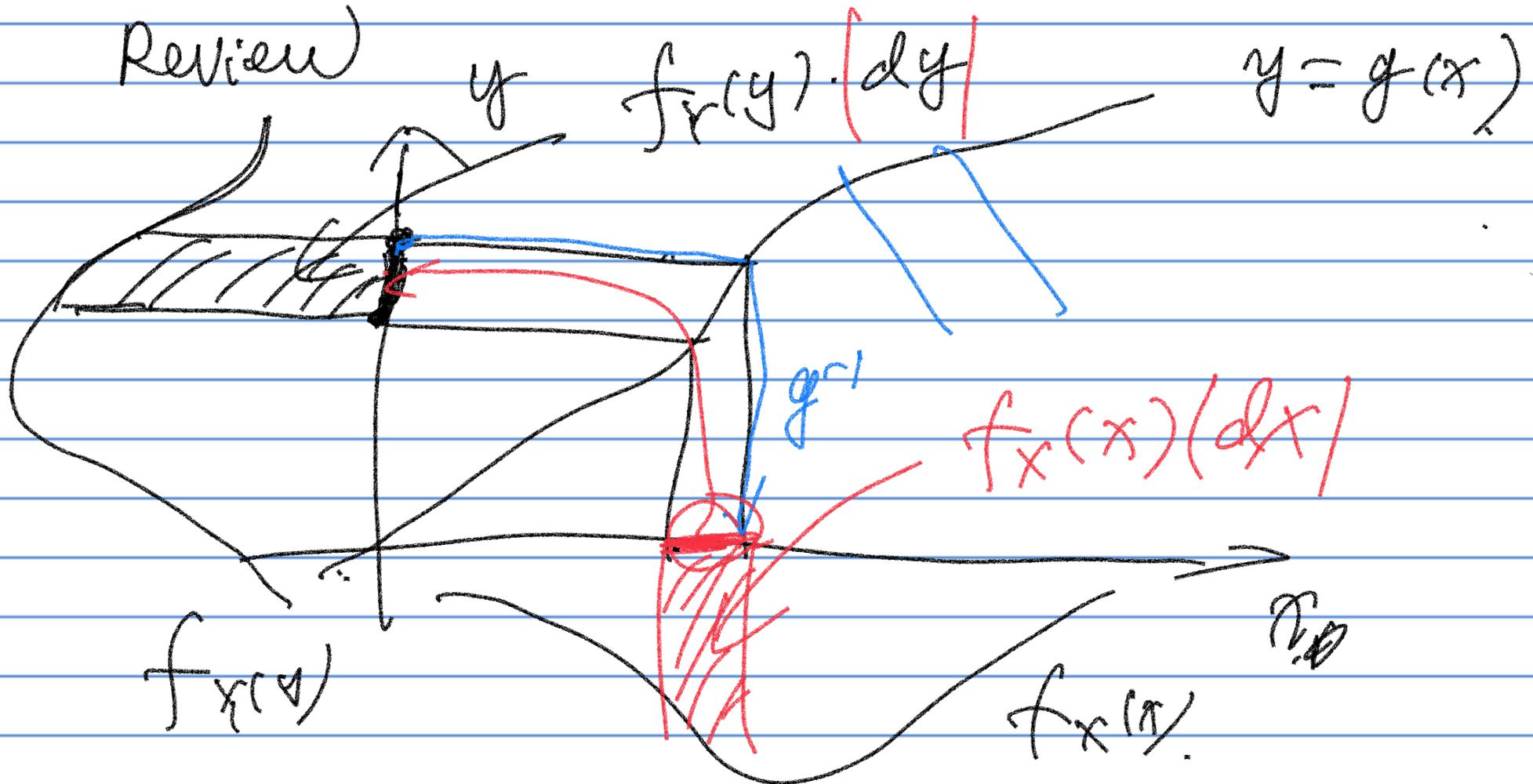
$$M_{(X_1, X_2)}(t_1, t_2) = E(e^{t_1 X_1 + t_2 X_2})$$

Expectation of Random Vector

$$E\begin{pmatrix} X_1 \\ X_2 \end{pmatrix} = \begin{pmatrix} E(X_1) \\ E(X_2) \end{pmatrix}$$


# Transformation of Random Vector

Review)

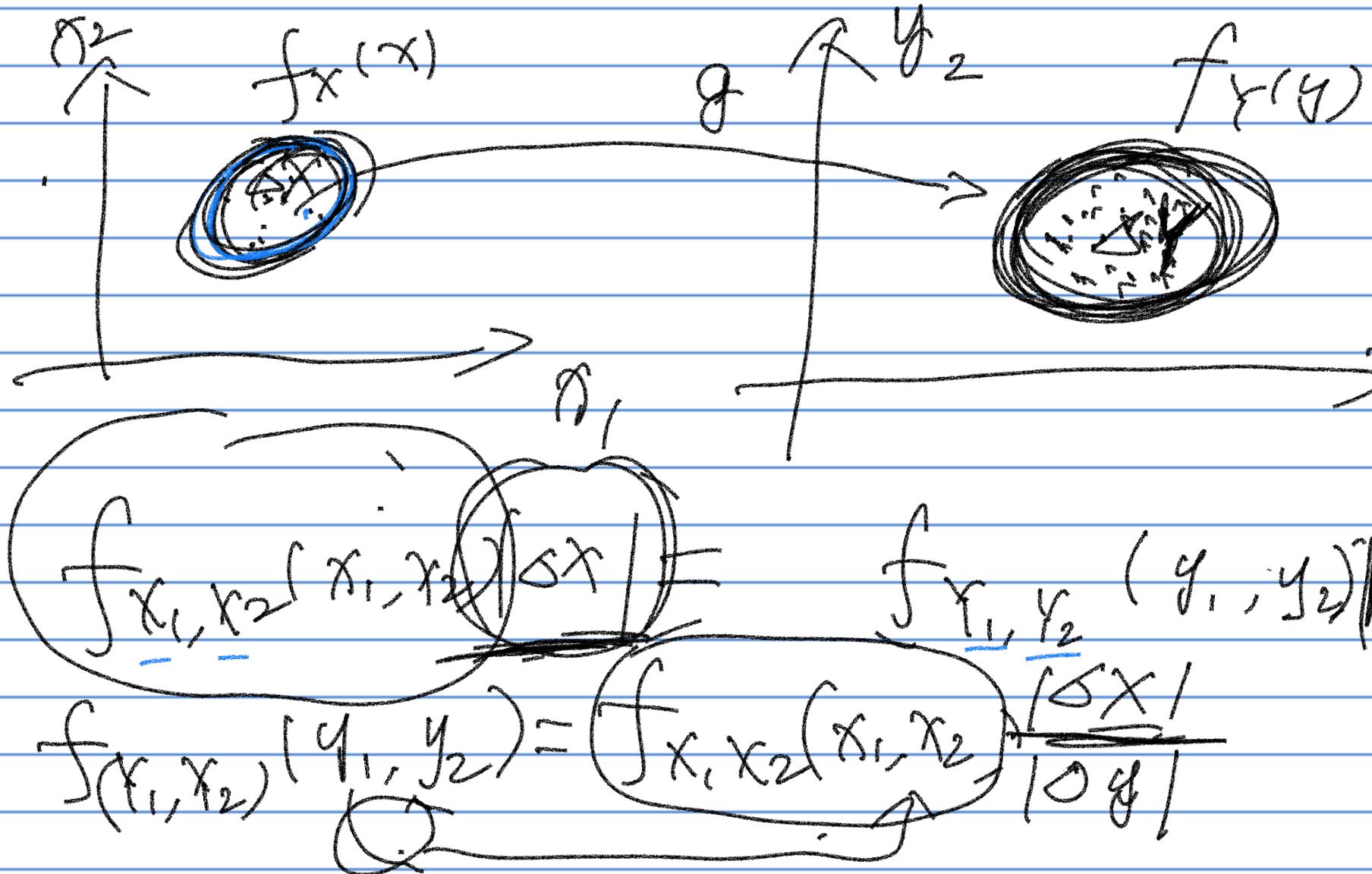


$$f_y(y) |dy| = f_x(x) |dx|$$

$$f_x(y) = f_x(g^{-1}(y)) \left| \frac{dx}{dy} \right|$$

$$= f_x(g^{-1}(y)) \left| \frac{dg^{-1}(y)}{dy} \right|$$

Jacobian.

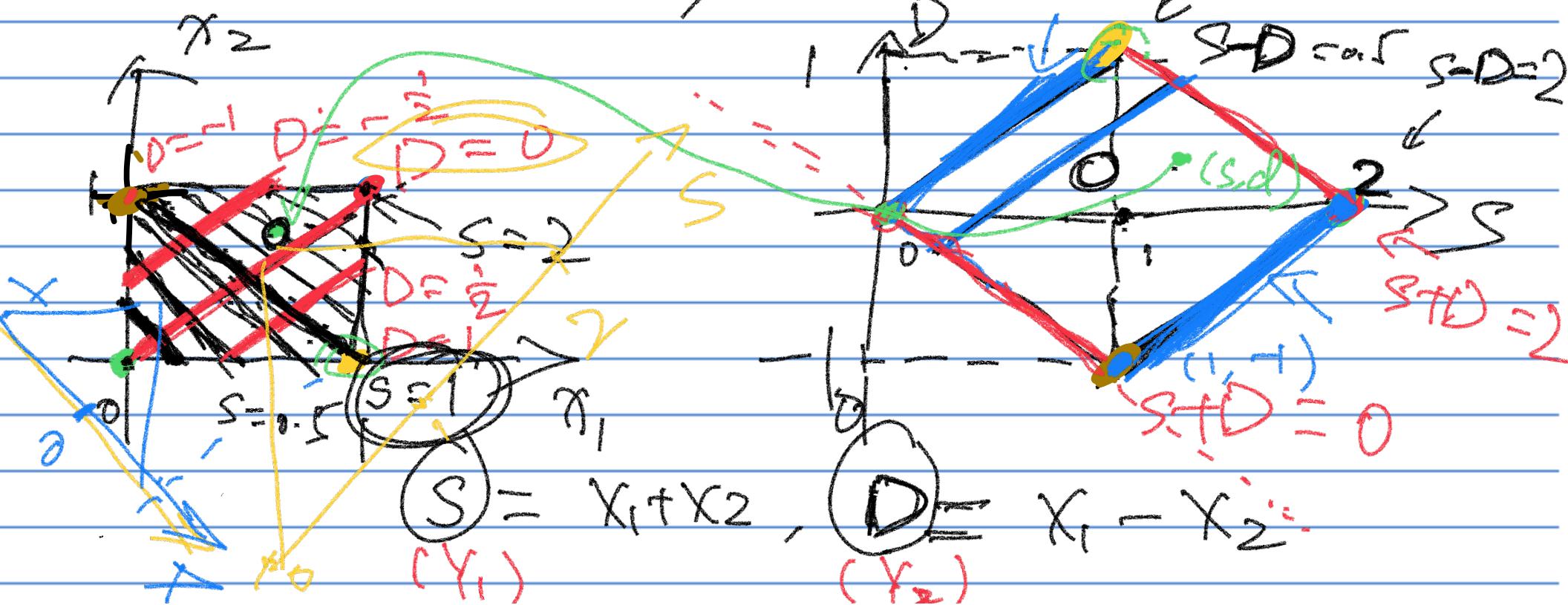


$$\frac{|\Delta X|}{|\Delta Y|} = \begin{vmatrix} \frac{\partial \theta_1}{\partial y_1} & \frac{\partial \theta_1}{\partial y_2} \\ \frac{\partial \theta_2}{\partial y_1} & \frac{\partial \theta_2}{\partial y_2} \end{vmatrix} + \checkmark$$

determinant

Example:

$$f(x_1, x_2) = \begin{cases} 1, & (x_1, x_2) \in [0,1] \times [0,1] \\ 0, & \text{o.w.} \end{cases}$$



$$\left\{ \begin{array}{l} D = X_1 - X_2 \\ S = X_1 + X_2 \end{array} \right. \quad \Leftrightarrow \quad \left\{ \begin{array}{l} X_1 = \frac{D+S}{2} \\ X_2 = \frac{S-D}{2} \end{array} \right.$$

$$\left\{ \begin{array}{l} 0 \leq \frac{D+S}{2} \leq 1 \\ 0 \leq \frac{S-D}{2} \leq 1 \end{array} \right. \quad \left\{ \begin{array}{l} 0 \leq S+D \leq 2 \\ 0 \leq S-D \leq 2 \end{array} \right. \quad ?$$

For  $0 \leq s+d \leq 2$ ,  $0 \leq s-d \leq 2$

$$f_{S,D}(s,d) = f_{X_1, X_2}(x_1, x_2) \cdot J$$

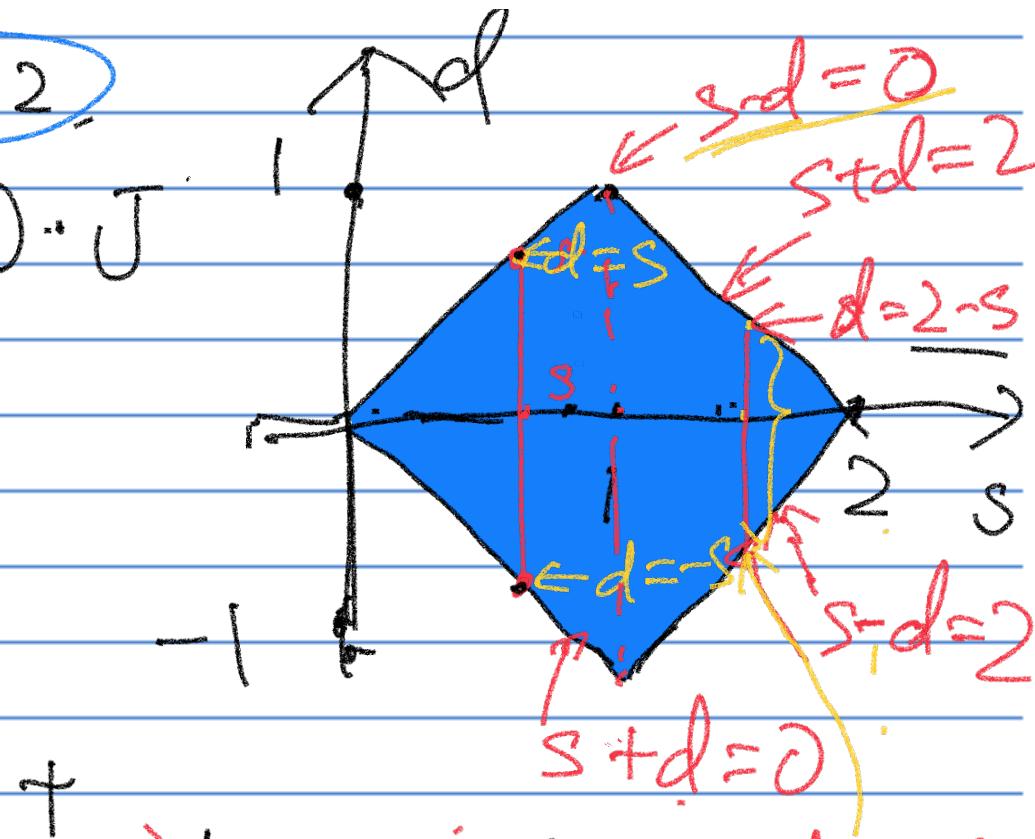
where

$$J = \begin{vmatrix} \frac{\partial \pi_1}{\partial s} & \frac{\partial \pi_1}{\partial d} \\ \frac{\partial \pi_2}{\partial s} & \frac{\partial \pi_2}{\partial d} \end{vmatrix} +$$

$$\textcircled{X}_1 = \frac{s+d}{2}$$

$$X_2 = \frac{\textcircled{X}_1}{2}$$

$$\bar{J} =$$



$$\bar{J} = \begin{vmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{vmatrix} = \left[ \frac{1}{4} - \frac{1}{4} \right] = \frac{1}{2}$$

$$f_{S,D}(s,d) = \underbrace{f_{X_1, X_2}(x_1, x_2)}_{+/-} \cdot J$$

$$= 1 \cdot \frac{1}{2}$$

$$= \frac{1}{2}.$$

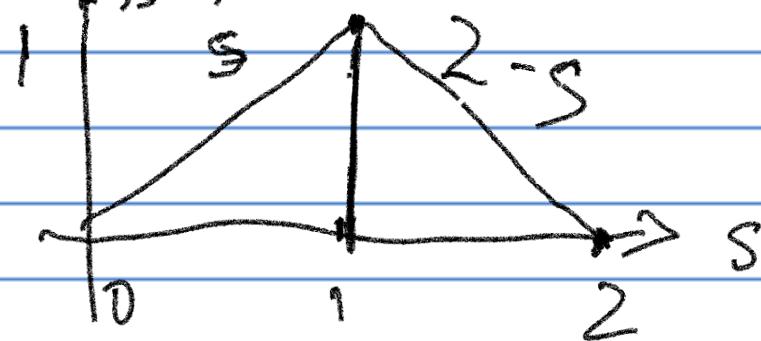
for  $0 \leq s+d \leq 2, 0 \leq s-d \leq 2$ .

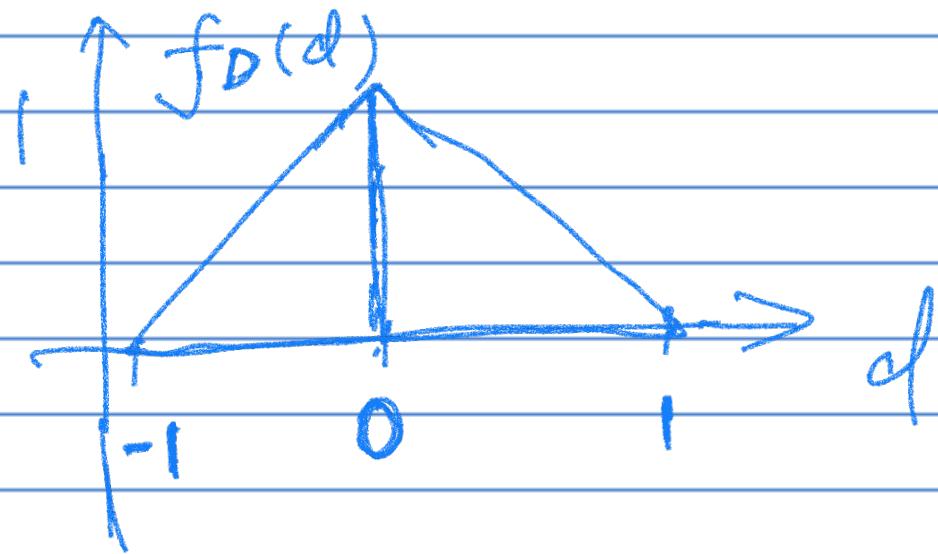
Find the P.D.F. of  $S$  &  $D$  resp.?

$$f_S(s) = \int_{-\infty}^{+\infty} f_{SD}(s, d) dd$$

$$= \int_{-s}^s \frac{1}{2} dd = s, \text{ if } 0 \leq s \leq 1$$

$$\left. \begin{array}{l} \int_{s-2}^{2-s} 1 dd = 2 - s \\ f_S(s) \end{array} \right\}, \text{ if } 1 \leq s \leq 2$$





exercice ?

Ghost method.

## Convolution Formula

$X_1, X_2$  has a joint P.D.F.  $f(x_1, x_2)$

What's the P.D.F. of  $S = X_1 + X_2$ .

$$\left\{ \begin{array}{l} S = X_1 + X_2 \\ Y = X_1 \end{array} \right. \quad \xrightarrow{\text{J}} \quad \left\{ \begin{array}{l} X_1 = Y \\ X_2 = S - Y \end{array} \right.$$
$$J = \begin{vmatrix} \frac{\partial \pi_1}{\partial s} & \frac{\partial \pi_1}{\partial y} \\ \frac{\partial \pi_2}{\partial s} & \frac{\partial \pi_2}{\partial y} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ 1 & -1 \end{vmatrix} = 1$$

$$f(s, y) = \underline{f_{x_1, x_2}(y, s-y)} \cdot 1$$

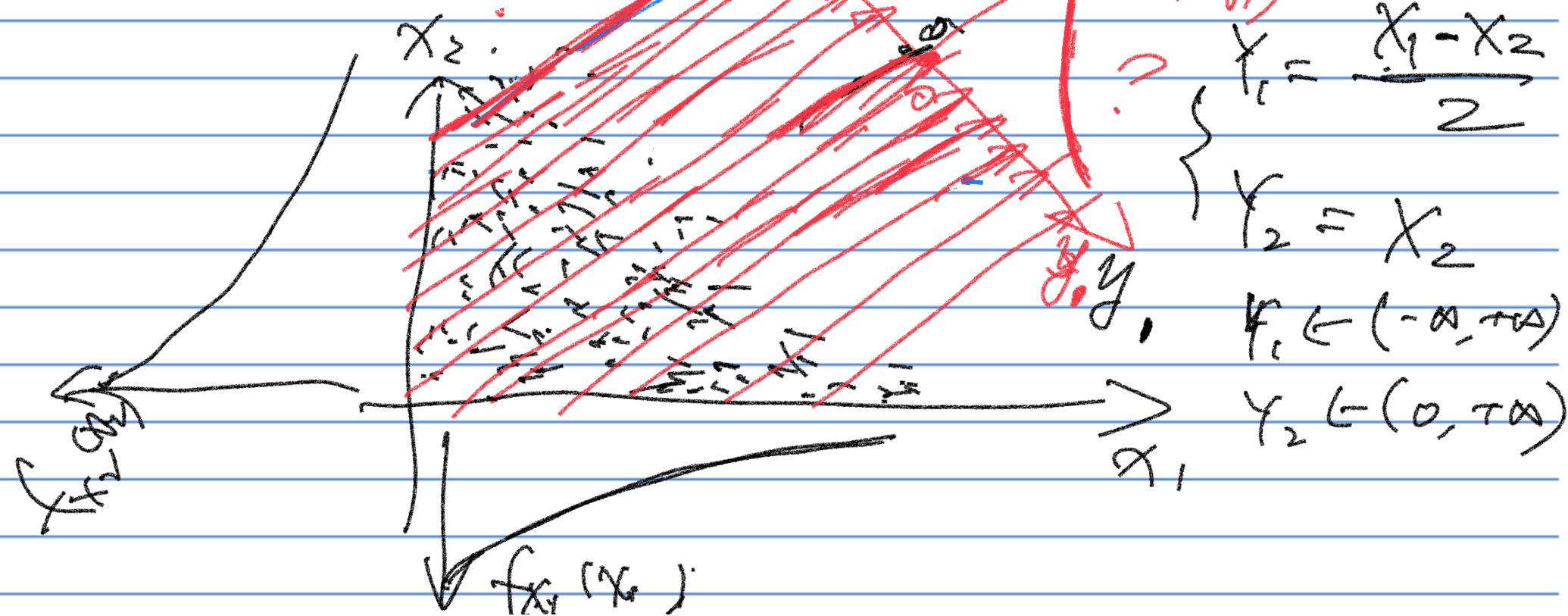
$$f(s) = \int_{-\infty}^{\tau(s)} f(s, y) dy$$

$$= \int_{-\alpha}^{\tau(\alpha)} f_{x_1, x_2}(y, s-y) dy$$

ghost method.

Example:

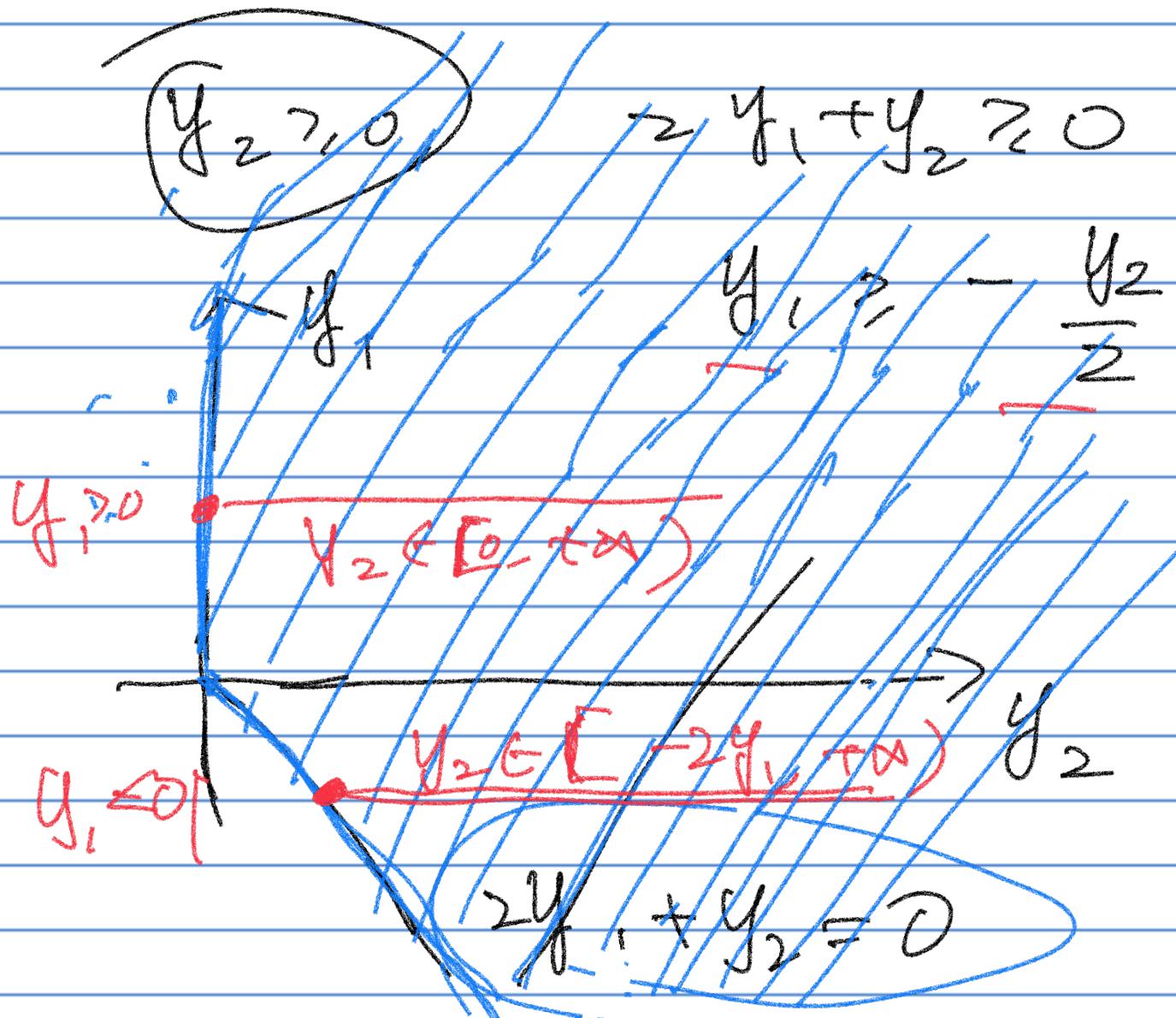
$$f_{X_1, X_2}(x_1, x_2) = \frac{1}{2} e^{-\frac{x_1+x_2}{2}} = \frac{1}{2} e^{-\frac{x_1}{2}} \cdot \frac{1}{2} e^{-\frac{x_2}{2}}$$



$$\left\{ \begin{array}{l} Y_1 = \frac{X_1 - X_2}{2} \\ Y_2 = X_2 \end{array} \right. \quad \Leftrightarrow \quad \left\{ \begin{array}{l} X_1 = 2Y_1 + Y_2 \\ X_2 = Y_2 \end{array} \right.$$

$$J = \begin{vmatrix} \frac{\partial X_1}{\partial Y_1} & \frac{\partial X_1}{\partial Y_2} \\ \frac{\partial X_2}{\partial Y_1} & \frac{\partial X_2}{\partial Y_2} \end{vmatrix} = \begin{vmatrix} 2 & 1 \\ 0 & 1 \end{vmatrix} = 2$$

$$2Y_1 + Y_2 \geq 0, \quad Y_2 \geq 0$$



$$f_Y(y_1, y_2) = f_X(x_1, x_2) \cdot J$$

$$= f_X(2y_1 + y_2, y_2) \cdot 2$$

$$= \frac{1}{4} e^{-\frac{1}{2}(2y_1 + 2y_2)} \cdot 2$$

$$= \frac{1}{2} e^{-(y_1 + y_2)}$$

$$\text{for } y_1 \geq -\frac{y_2}{2}, y_2 \geq 0 \quad 2y_1 \geq -y_2$$

$$f_{Y_1}(y_1) = \int f_Y(y_1, y_2) dy_2 \quad -2y_1 \leq y_2$$

$$\int_{-\infty}^{+\infty} \frac{1}{2} e^{-y_1} \cdot e^{-y_2} dy_2, \quad y_1 \geq 0$$

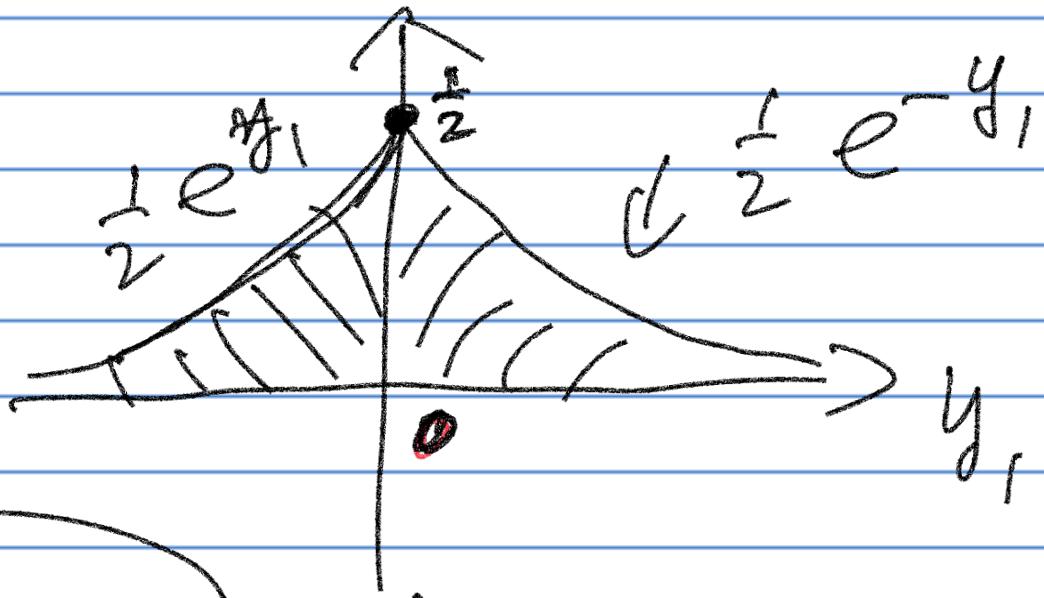
$$= \left\{ \int_{-2y_1}^{+\infty} \frac{1}{2} e^{-y_1} e^{-y_2} dy_2, \quad y_1 \leq 0 \right.$$

For  $y_1 > 0$

$$f_{Y_1}(y_1) = \left( \int_0^{+\infty} e^{-y_2} dy_2 \right) \cdot \frac{1}{2} e^{-y_1}$$
$$= \frac{1}{2} e^{-y_1}$$

For  $y_1 \leq 0$ ,

$$f_{Y_1}(y_1) = \int_{-2y_1}^{+\infty} e^{-y_2} dy_2 \cdot \frac{1}{2} e^{-y_1}$$
$$= -e^{-y_2} \Big|_{-2y_1}^{+\infty} \cdot \frac{1}{2} e^{-y_1}$$
$$= e^{+2y_1} \cdot \frac{1}{2} e^{-y_1} = \frac{1}{2} e^{y_1}$$



Laplace distribution.

LASSO

$$f_{\ell_1}(y_1) = \frac{1}{2} e^{-|y_1|}$$