

# Chapter 3 Joint distributions

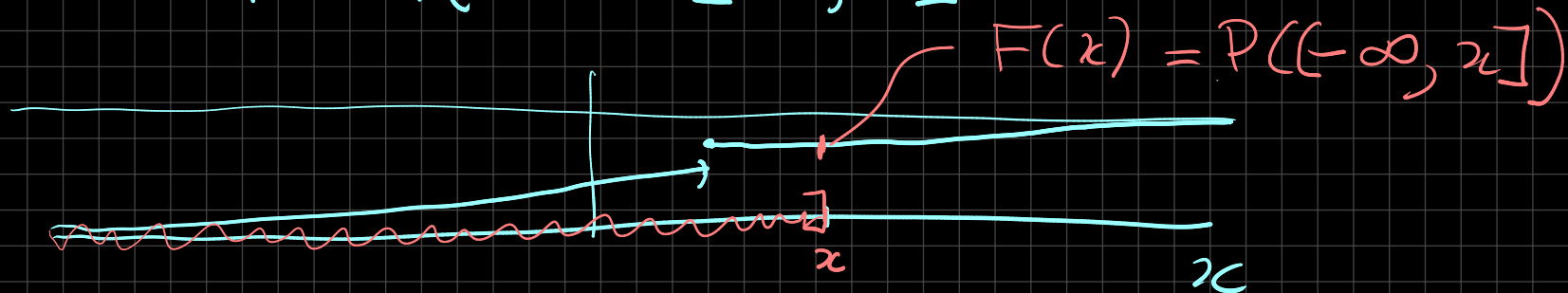
## 2 or more random variables

E.G. • Height • Weight

- Dose of a drug + time to recovery
- Gender, Age, Education, Income

CDF One RV:  $F(x) = P(X \leq x)$

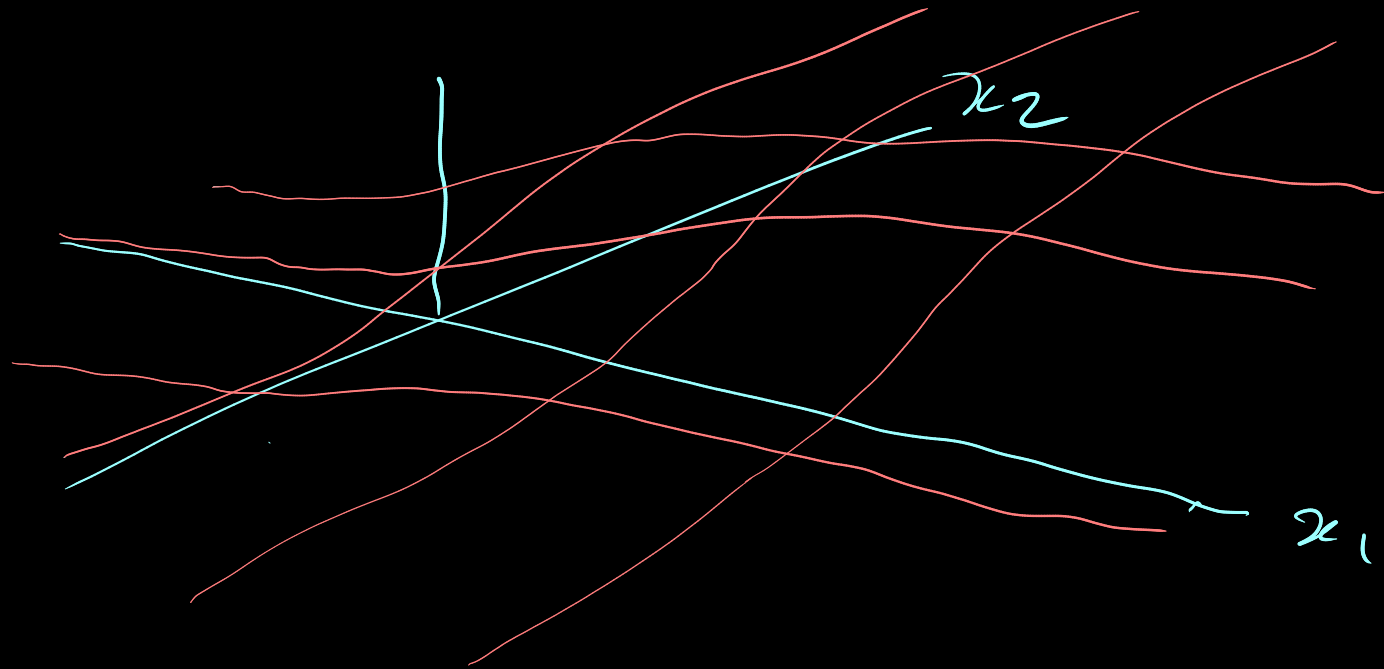
$$F: \mathbb{R} \rightarrow [0, 1]$$



CDF: Two R.V.s  $X_1, X_2$

$$F(x_1, x_2) = P(X_1 \leq x_1 \text{ and } X_2 \leq x_2)$$

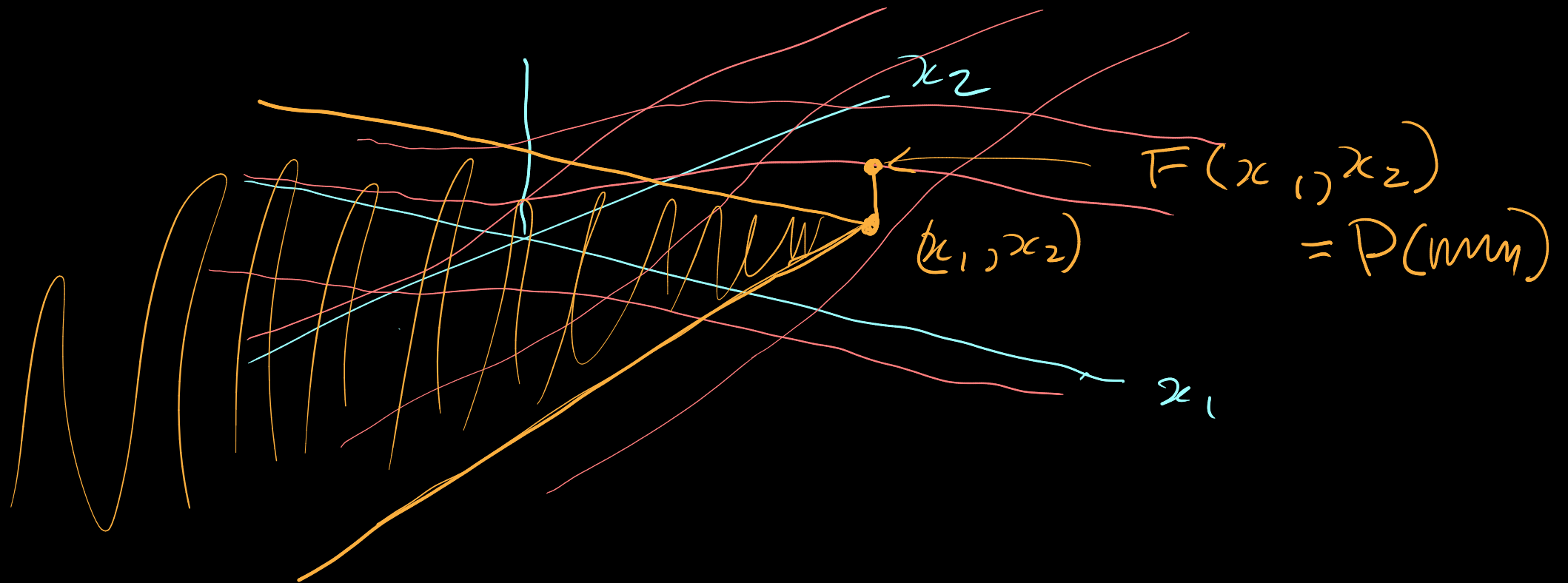
$$F: \mathbb{R}^2 \rightarrow [0, 1]$$



CDF: Two R.V.s  $X_1, X_2$

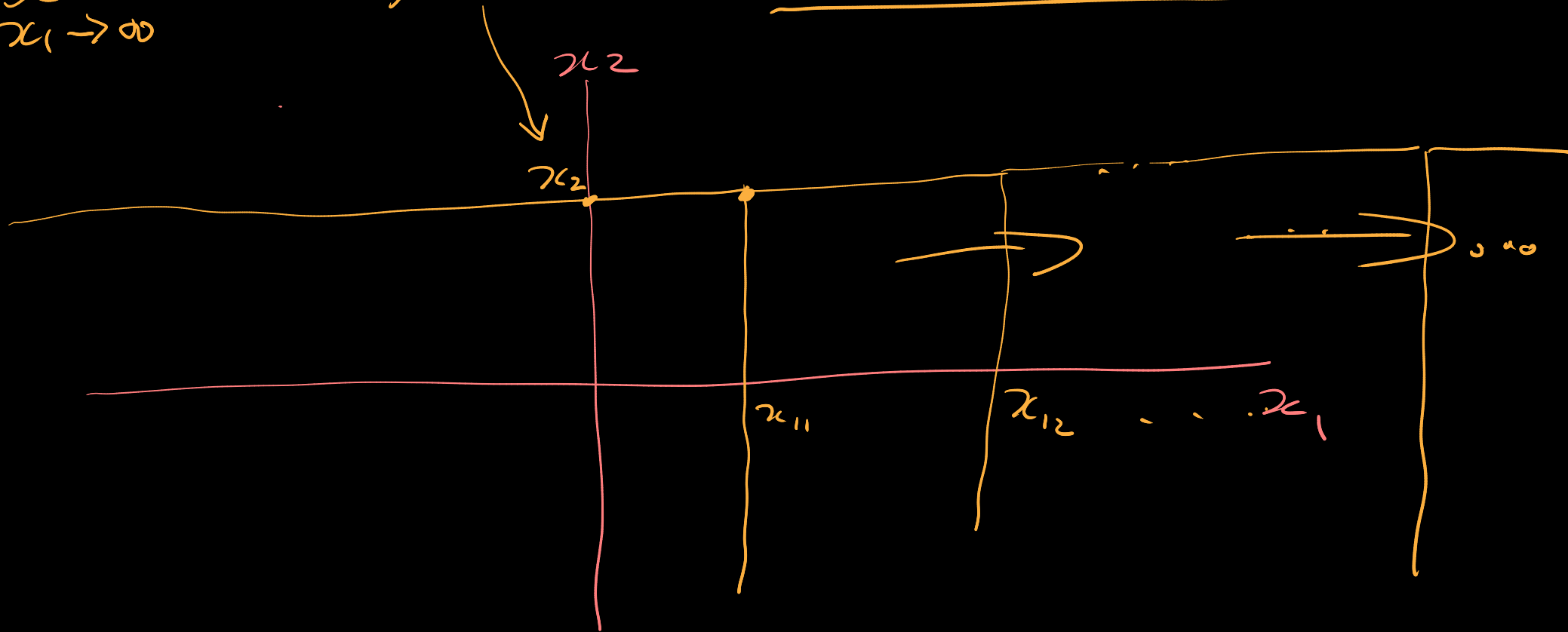
$$F(x_1, x_2) = P(X_1 \leq x_2 \text{ and } X_2 \leq x_2)$$

$$F: \mathbb{R}^2 \rightarrow [0, 1]$$



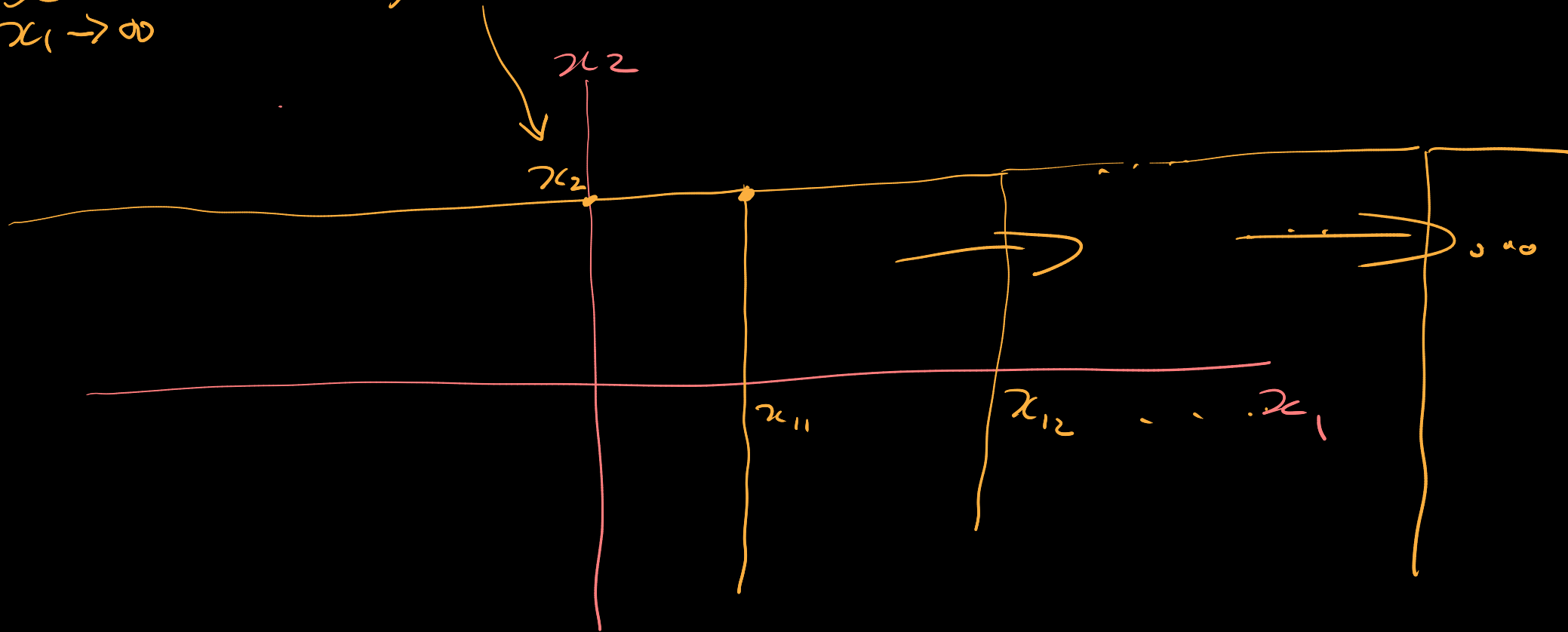
If you fix  $x_2$  and let  $x_1 \rightarrow \infty$

$$\lim_{x_1 \rightarrow \infty} F(x_1, x_2) =$$



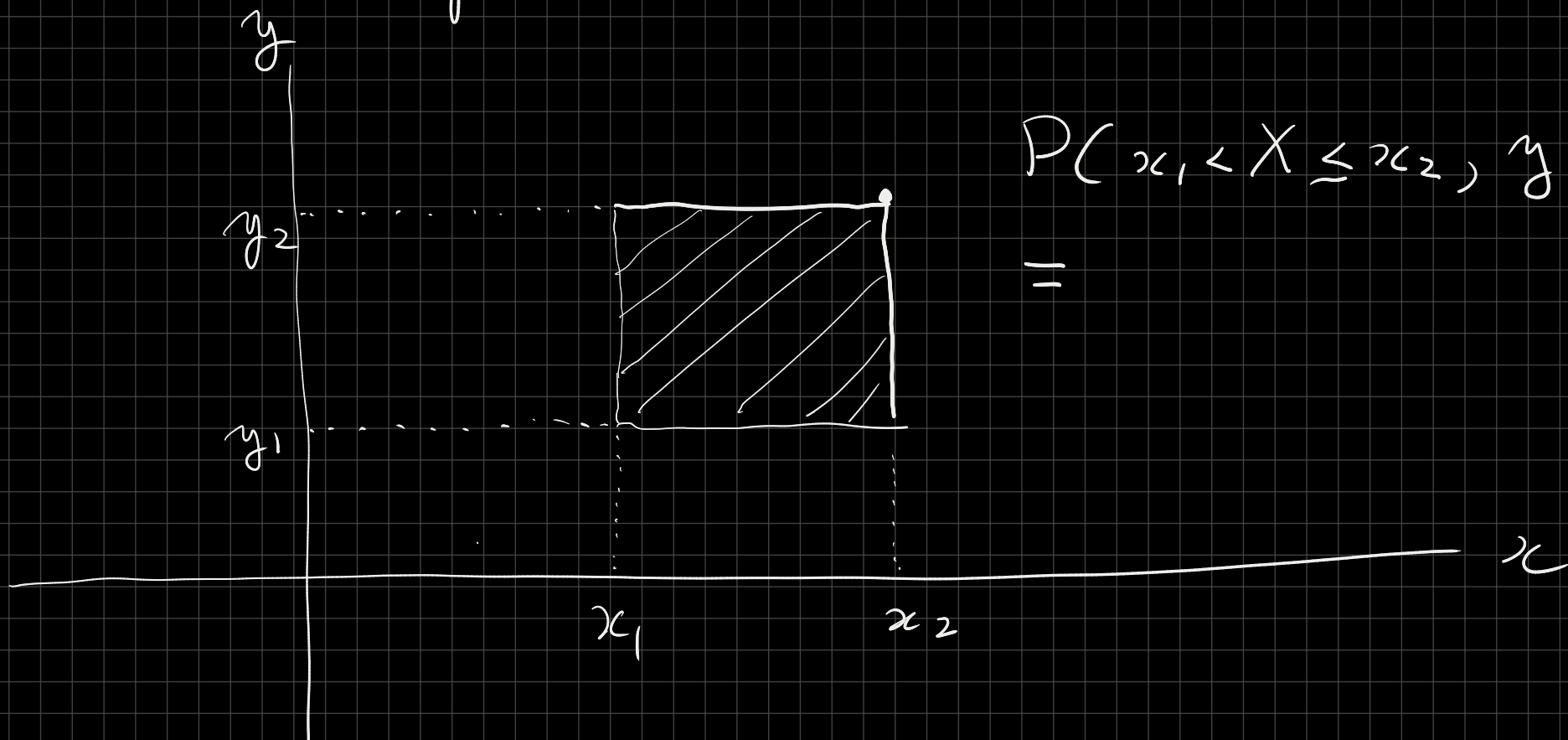
If you fix  $x_2$  and let  $x_1 \rightarrow \infty$

$$\lim_{x_1 \rightarrow \infty} F(x_1, x_2) = F_2(x_2)$$



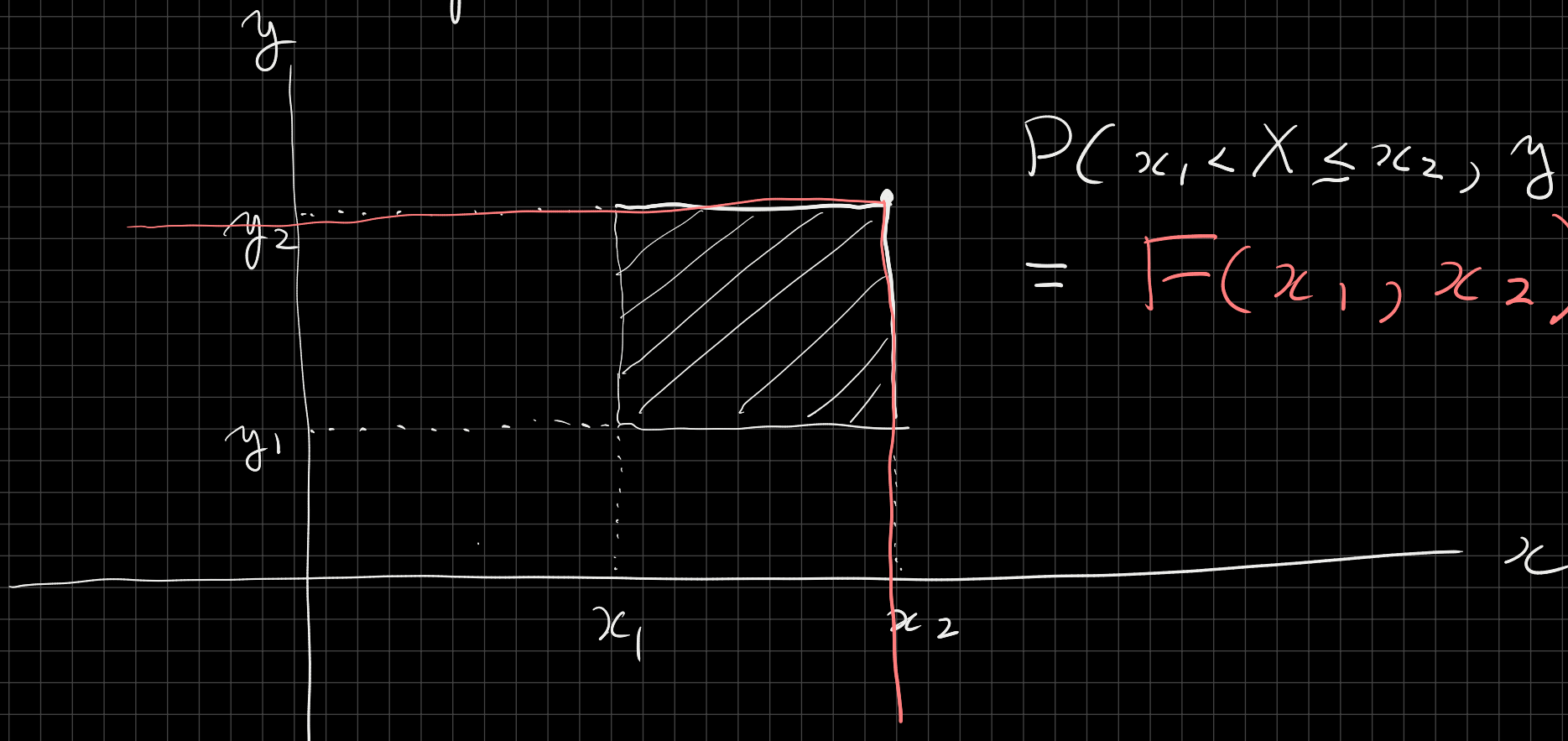
Similarly  $\lim_{x_2 \rightarrow \infty} F(x_1, x_2) = F(x_1)$

Probability of a rectangle:



Similarly  $\lim_{x_2 \rightarrow \infty} F(x_1, x_2) = F(x_1)$

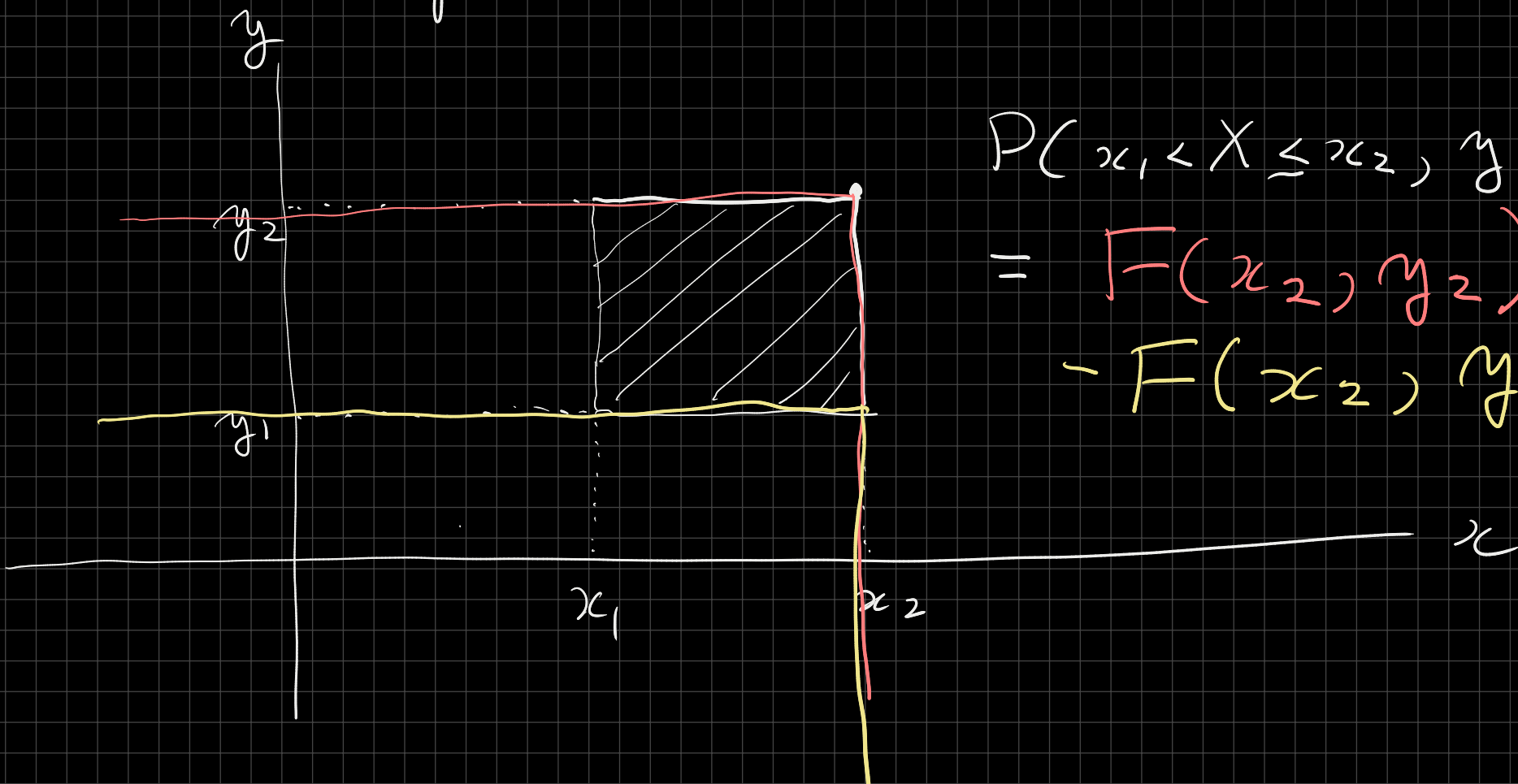
Probability of a rectangle:



$$P(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ = F(x_1, x_2)$$

Similarly  $\lim_{x_2 \rightarrow \infty} F(x_1, x_2) = F(x_1)$

Probability of a rectangle:

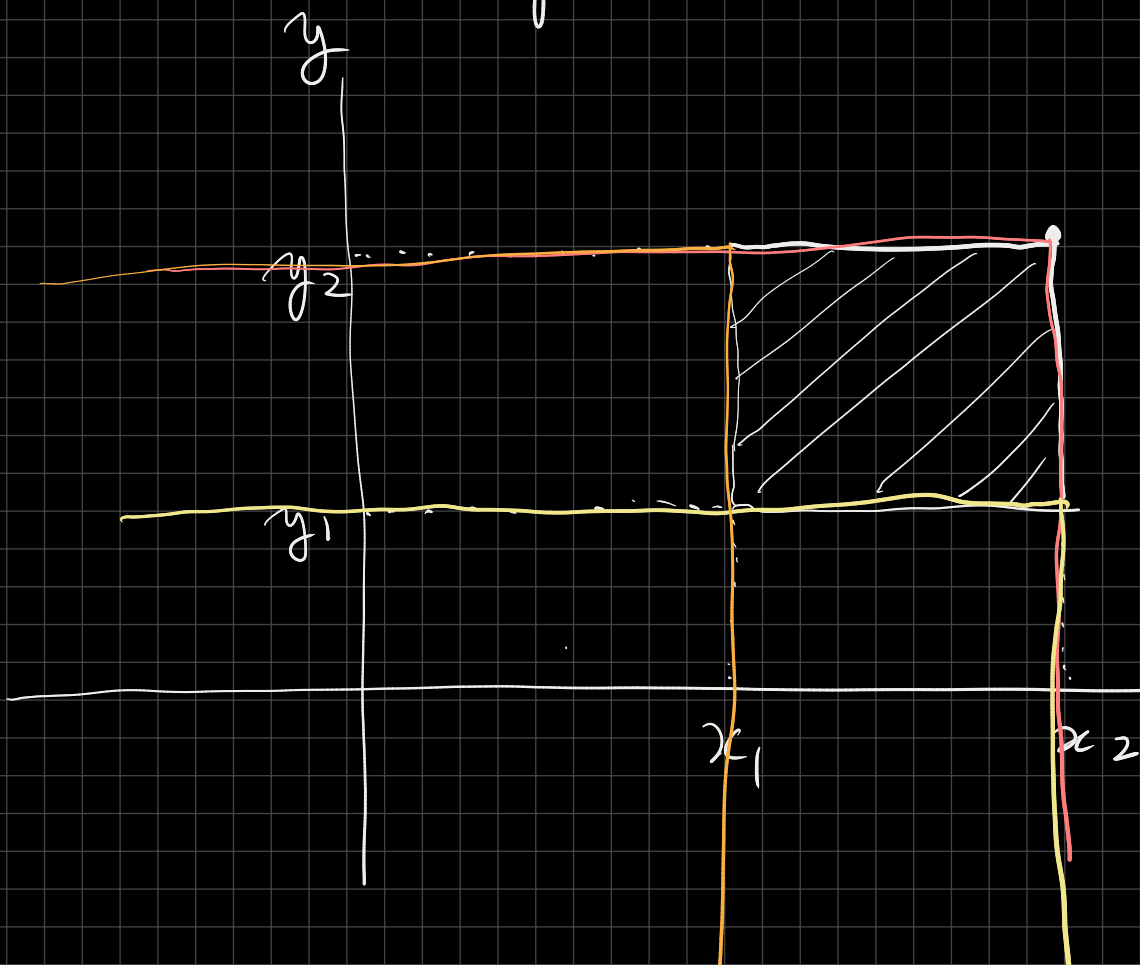


$$P(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ = F(x_2, y_2) \\ - F(x_2, y_1)$$



Similarly  $\lim_{x_2 \rightarrow \infty} F(x_1, x_2) = F(x_1)$

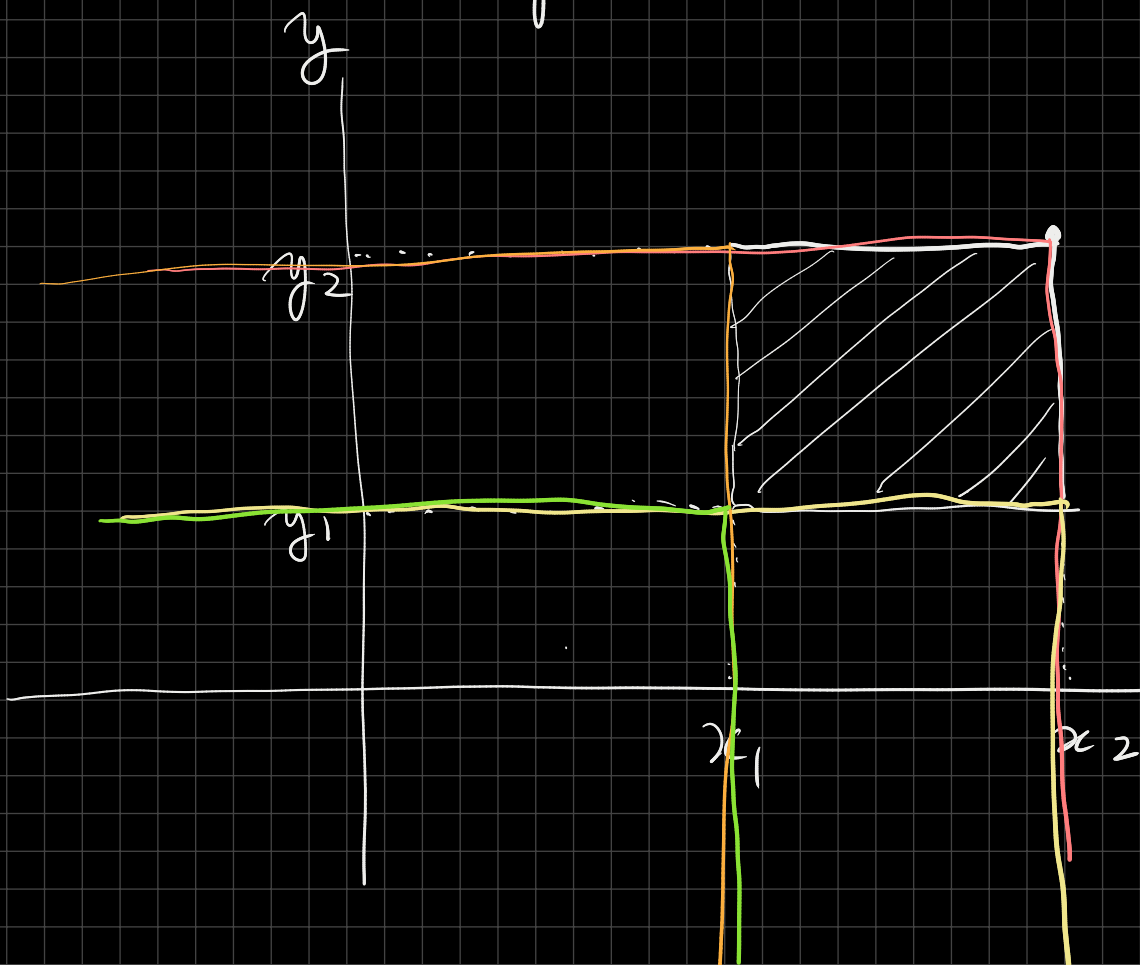
Probability of a rectangle:



$$\begin{aligned} P(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ &= F(x_2, y_2) \\ &\quad - F(x_2, y_1) \\ &\quad - F(x_1, y_2) \end{aligned}$$

Similarly  $\lim_{x_2 \rightarrow \infty} F(x_1, x_2) = F(x_1)$

Probability of a rectangle:



$$\begin{aligned} P(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ &= F(x_2, y_2) \\ &\quad - F(x_2, y_1) \\ &\quad - F(x_1, y_2) \\ &\quad + F(x_1, y_1) \end{aligned}$$

Recap:

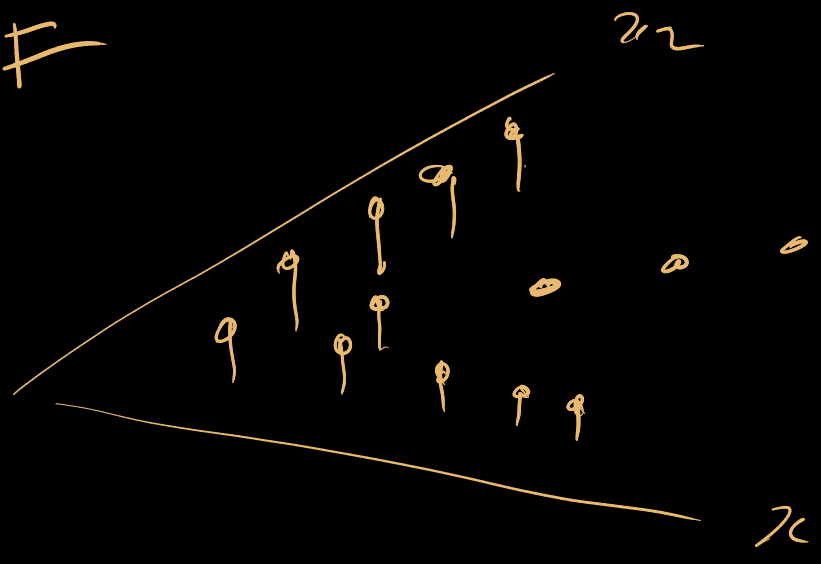
$$P(x_1 < X \leq x_2, y_1 < Y \leq y_2)$$

$$= F(x_2, y_2) - F(x_2, y_1) - F(x_1, y_2) + F(x_1, y_1)$$

Discrete random variables

$$\begin{aligned} \text{Joint PMF } P(X_1 = x_1, X_2 = x_2, X_3 = x_3) \\ = p(x_1, x_2, x_3) \end{aligned}$$

Joint PMF



Def  $\text{Supp}(X_1, X_2)$  is finite, we can use a table.  
 e.g. Toss a coin:  $X = \# \text{ of Heads}$  & a die,  $Y = \# \text{ rolled}$ .

		Y						
		1	2	3	4	5	6	
X	0	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
	1	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
		$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	1

Joint PMF

marginal PMF for X

marginal for Y

$P(\Omega)$

In general: Joint PMF

$$p(x_1, x_2, x_3, x_4)$$

To get marginal PMF just sum over the variables you don't want:

$$p(x_1, x_3) = \sum_{\substack{\text{all } x_2\text{'s} \\ \text{all } x_4\text{'s}}} p(x_1, x_2, x_3, x_4)$$

## Example: Multinomial distribution

Consider  $X \sim \text{Binomial}(n, p)$

$$P(H) = p \quad P(T) = 1 - p = q$$

Toss n Times      H H H T H H H T H H      ( $n=10$ )  
8 Hs and 2 Ts

$$\text{So } X = 8$$

But we could also record # of Ts, 2  
and think of this as a joint distribution

for  $X_1 = \# \text{ of Hs}$ , and  $X_2 = \# \text{ of Ts}$

Here  $(X_1, X_2) = (8, 2)$

Of course  $X_1 + X_2 = n$

Here's the joint distribution for  $n = 3$

		$X_2$ (Tails)		
$X_1$ (Heads)	0	0	0	$1/4$
	1	0	$1/2$	0
	2	$1/4$	0	0

$(X_1, X_2) \sim \text{Multinomial}(3, p = (1/2, 1/2))$

This generalized to any number of categories.

E.G. Take a sample of  $n$  students and record eye color: Black, Hazel, Blue, Gray

Counts:  $X_1$     $X_2$     $X_3$     $X_4$

$(X_1, X_2, X_3, X_4) \sim \text{Multinomial}(n, (p_1, p_2, p_3, p_4))$

proportions of B, H, B, G  
in population

Or.  $\underline{X} \sim \text{Multinomial}(n, \underline{P})$



PXF:

$$P(x_1, \dots, x_n) = \binom{n}{x_1, x_2, \dots, x_n} p_1^{x_1} p_2^{x_2} \dots p_n^{x_n}$$

$$\text{if } x_i \geq 0, \sum x_i = n,$$

$$p_i > 0, \sum p_i = 1$$

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Let  $X_1, X_2, X_3$  be Multinomial  $(n, (p_1, p_2, p_3))$

Then  $X_1 \sim \text{Binomial}(n, p_1)$

Proof:  $P(X_1 = x_1) = \sum_{\substack{x_2 \\ x_3 = n - x_1 - x_2}} \binom{n}{x_1, x_2, x_3} p_1^{x_1} p_2^{x_2} p_3^{x_3}$

pulling out factors that don't depend on summands  $= \frac{n!}{x_1!} p_1^{x_1} \sum_{x_2 + x_3 = n - x_1} \frac{1}{x_2! x_3!} p_2^{x_2} p_3^{x_3}$

Greatest tricks in math { multiply  $\times 1$   
add 0  
What can we do here?

$= \dots$

# Continuous random variables

$\iint$  instead of  $\sum_i$

$(X, Y)$  have a bivariate continuous distribution

if 
$$P((X, Y) \in A) = \iint_A f(x, y) dx dy$$

for some function  $f$  such that

1)  $f \geq 0$  on  $\mathbb{R}^2$

2)  $\iint f(x, y) dx dy = 1$

CDF: 
$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(u, v) dv du$$

Simple example :  $f(x, y) = c(x+y)$   $x, y \in (0, 1)$

1) Find  $c$  to make  $f$  a density

$$\iint f(x, y) dx dy = \int_0^1 \int_0^1 (x+y) dy dx$$

$$= \int_0^1 \left[ xy + \frac{y^2}{2} \right]_0^1 dx$$

$$= \int_0^1 \left( x + \frac{1}{2} \right) dx = \left[ \frac{x^2}{2} + \frac{1}{2}x \right]_0^1$$

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

So  $c = 1$

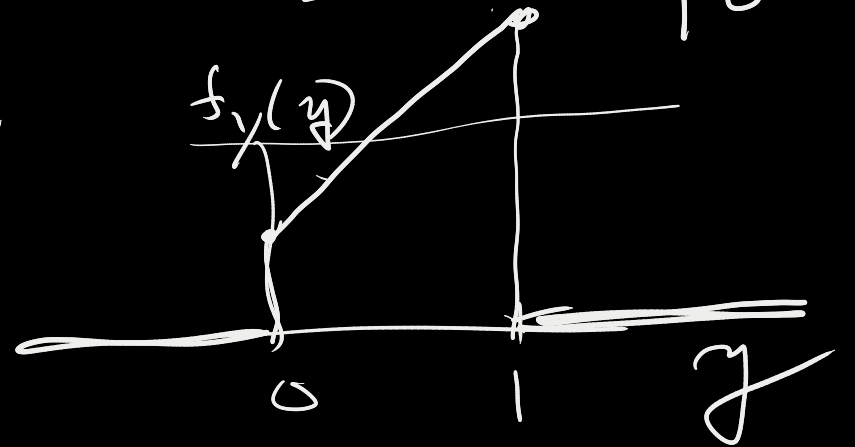
## Marginal distribution of $Y$

Integrate out  $X$ :

$$f_Y(y) = \int f(x, y) dx$$

$$= \int_0^1 x + y dx = \left[ \frac{x^2}{2} + xy \right]_0^1$$

$$= \frac{1}{2} + y$$



$$\begin{aligned} \text{So } P(Y \leq 1/2) &= \int_0^{1/2} f_Y(y) dy \\ &= \int_0^{1/2} \frac{1}{2} + y dy = \frac{3}{8} \end{aligned}$$

