

pmf = función de masa (función de probabilidad)

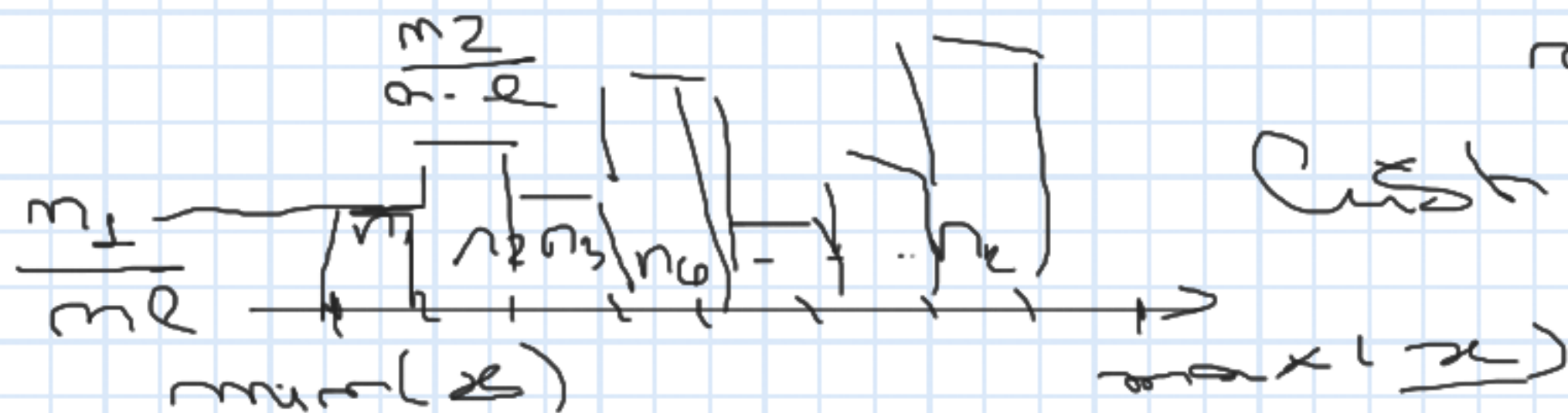
$P_X(x) \rightarrow$ proba \rightarrow a.

$P_X(x)$



$X \sim \text{Bin}(4, 1/2)$

$\underline{x} = [x_1, \dots, x_n]$ donde x_i es el valor de x en la i -ésima posición.
 X con dist. $f_X(x)$ (X es cont.)
 p_i es la probabilidad de que $X = x_i$.



Ex 1 $X = "$ # correct answers in 10 trials"

$$X \sim \text{Bin}(10, 1/2)$$

$$a) P(X=3) = \binom{10}{3} \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^{10-3}$$

$$= 0.1172$$

$$b) P(X \geq 3) = 1 - P(X < 3)$$

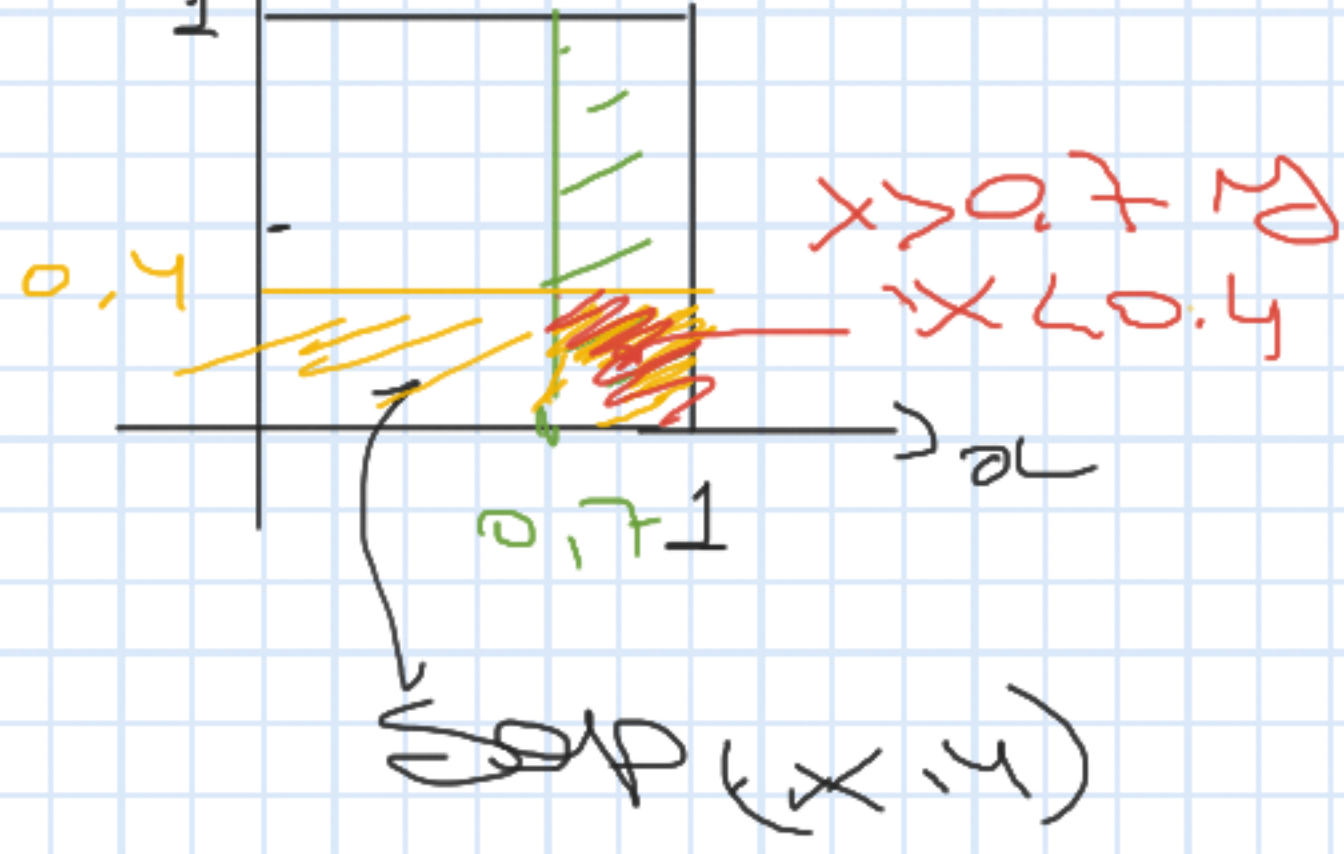
$$X \sim \text{Bin}(10, 0.4) = 1 - P(X \leq 2) \\ = 1 - \text{cdf}(2) = 0.8333$$

c) $Y = "$ Resultado de uma tirada"

$$\hookrightarrow P(X < 2)$$

$$P(X \geq 3 | Y=0) = \frac{P(X \geq 3, Y=0)}{P(Y=0)} = 0$$

$X, Y \sim \text{i.i.d. } U(0,1)$ independent & identically distributed.



$$P((X, Y) \in [0.7, 1] \times [0.4, 1]) = P(X > 0.7, Y < 0.4)$$

$$P(X > 0.7, Y < 0.4) = P(X > 0.7) P(Y < 0.4) = (1 - 0.7)(0.4) = 0.12$$

$$\underline{X} = (X_1, X_2)$$

$$\underline{\Sigma} = \begin{pmatrix} \sigma_{x_1}^2 & \text{cov}(x_1, x_2) \\ \text{cov}(x_1, x_2) & \sigma_{x_2}^2 \end{pmatrix} \quad \underline{\mu} = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}$$

$$f_{\underline{X}}(\underline{x}) = \frac{1}{2\pi |\underline{\Sigma}|^{1/2}} e^{-\frac{1}{2} (\underline{x} - \underline{\mu})^T \underline{\Sigma}^{-1} (\underline{x} - \underline{\mu})}$$

$$|\underline{\Sigma}| = \sigma_{x_1}^2 \sigma_{x_2}^2 - \text{cov}^2(x_1, x_2)$$

$$= \sigma_{x_1}^2 \sigma_{x_2}^2 (1 - \rho^2)$$

$$\rho = \frac{\text{cov}(x_1, x_2)}{\sigma_{x_1} \sigma_{x_2}}$$

$$f_{\underline{X}}(\underline{x}) = \frac{1}{2\pi \sigma_{x_1} \sigma_{x_2} \sqrt{1 - \rho^2}} e^{-\frac{1}{2(1-\rho^2)} \left\{ \frac{(x_1 - \mu_1)^2}{\sigma_{x_1}^2} + \frac{(x_2 - \mu_2)^2}{\sigma_{x_2}^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_{x_1} \sigma_{x_2}} \right\}}$$

(Coeff. of correlation)

$$\Sigma^{-1} = \begin{bmatrix} \frac{1}{\sigma_1^2} & 0 \\ 0 & \frac{1}{\sigma_2^2} \end{bmatrix} \Rightarrow \Sigma^{-1} = \begin{bmatrix} \frac{1}{\sigma_1^2} & 0 \\ 0 & \frac{1}{\sigma_2^2} \end{bmatrix}$$

$$\text{cov}(x_1, x_2) = 0 \Rightarrow \rho = 0$$

\Rightarrow des correlations

\Rightarrow ~~not~~ ~~indep~~ ~~or~~ ~~not~~ ~~normal~~ ~~indep~~

Σ is diag $\Rightarrow x_1, \dots, x_n$ are indep.

$$\frac{1}{2\pi \sqrt{\sigma_x^2 + \sigma_y^2}} \exp\left(-\frac{1}{2\sigma_x^2} \left(\frac{x_1 - \mu_1}{\sigma_x}\right)^2 - \frac{1}{2\sigma_y^2} \left(\frac{x_2 - \mu_2}{\sigma_y}\right)^2\right)$$

$$\frac{1}{2\pi \sqrt{\sigma_x^2 + \sigma_y^2}} \exp\left(-\frac{1}{2\sigma_x^2} \left(\frac{x_1 - \mu_1}{\sigma_x}\right)^2 - \frac{1}{2\sigma_y^2} \left(\frac{x_2 - \mu_2}{\sigma_y}\right)^2\right)$$

$$\exp\left(-\frac{1}{2\sigma_x^2} \left(\frac{x_1 - \mu_1}{\sigma_x}\right)^2 - \frac{1}{2\sigma_y^2} \left(\frac{x_2 - \mu_2}{\sigma_y}\right)^2\right)$$

$$\exp\left(-\frac{1}{2\sigma_y^2} \left(\frac{x_2 - \mu_2}{\sigma_y}\right)^2\right)$$

$$\frac{1}{\sqrt{2\pi} \sigma_x}$$

$$\frac{1}{\sqrt{2\pi} \sigma_y}$$

$$\sim \mathcal{N}(\mu_1, \sigma_x^2)$$

$$\sim \mathcal{N}(\mu_2, \sigma_y^2)$$

$$f_{X,Y}(x,y) = f_X(x) f_Y(y)$$

indep.

$$f(x) = \int_{-\infty}^{\infty} f_{x,y}(x,y) dy$$

→ gaussian any

$$\mathbb{E}[w^T x] = w^T \mathbb{E}[x] = w^T \mu$$

$$\begin{aligned} \text{cov}(w^T x) &= \text{cov}\left(\underbrace{(w^T x)}_{x^T w}\right) = \\ &= w^T \underbrace{\text{cov}(xx^T)}_N w = w^T \Sigma w \end{aligned}$$

$X \sim F_X(x)$ strictly monotone increasing

a) $T = G(X) \rightarrow$ tiene inversa

$$\begin{aligned} F_T(t) &= P(T \leq t) = P(G(X) \leq t) = \\ &= P(X \leq G^{-1}(t)) = \\ &= F_X(G^{-1}(t)) \end{aligned}$$

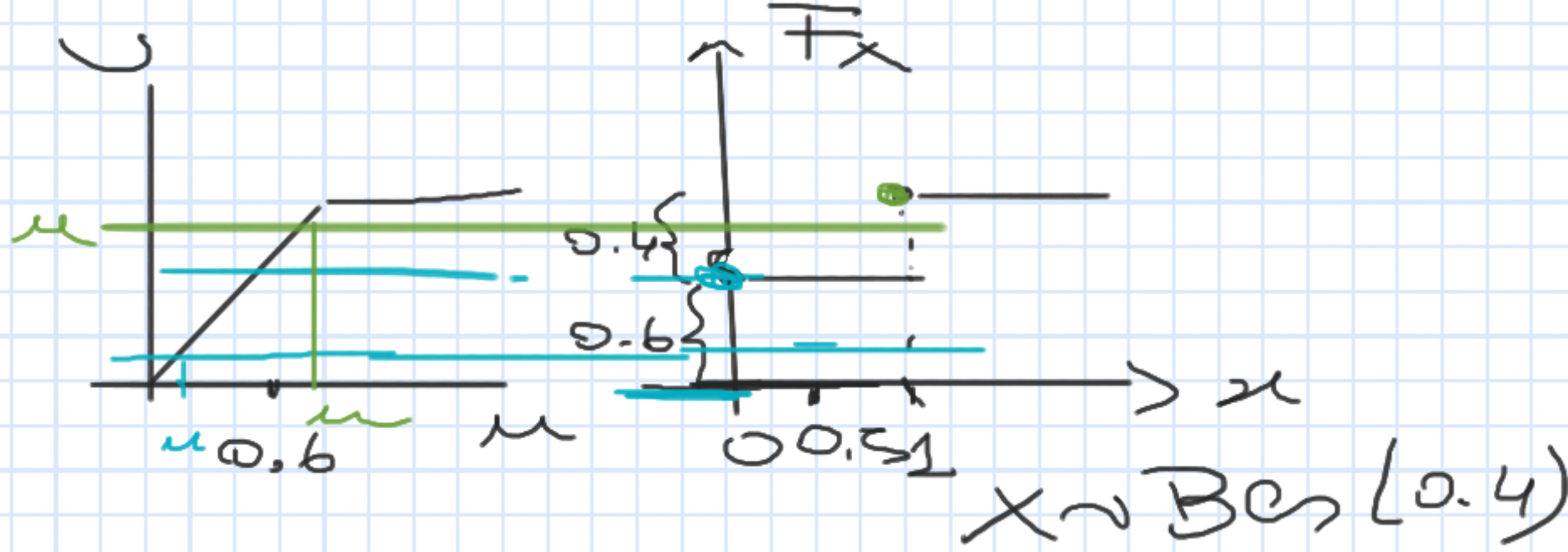
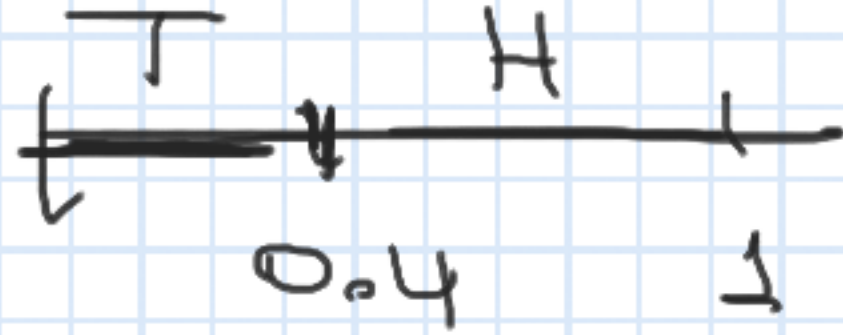
b) $U = F_X(X)$

$$F_U(u) = P(F_X(X) \leq u) = \{u \mid 0 < u < 1\}$$

$$f_U(u) = 1 \mid \{0 < u < 1\}$$

$$U \sim \mathcal{U}(0, 1)$$

$$\rightarrow F_X^{-1}(U) = X \text{ distribucion } F_X$$

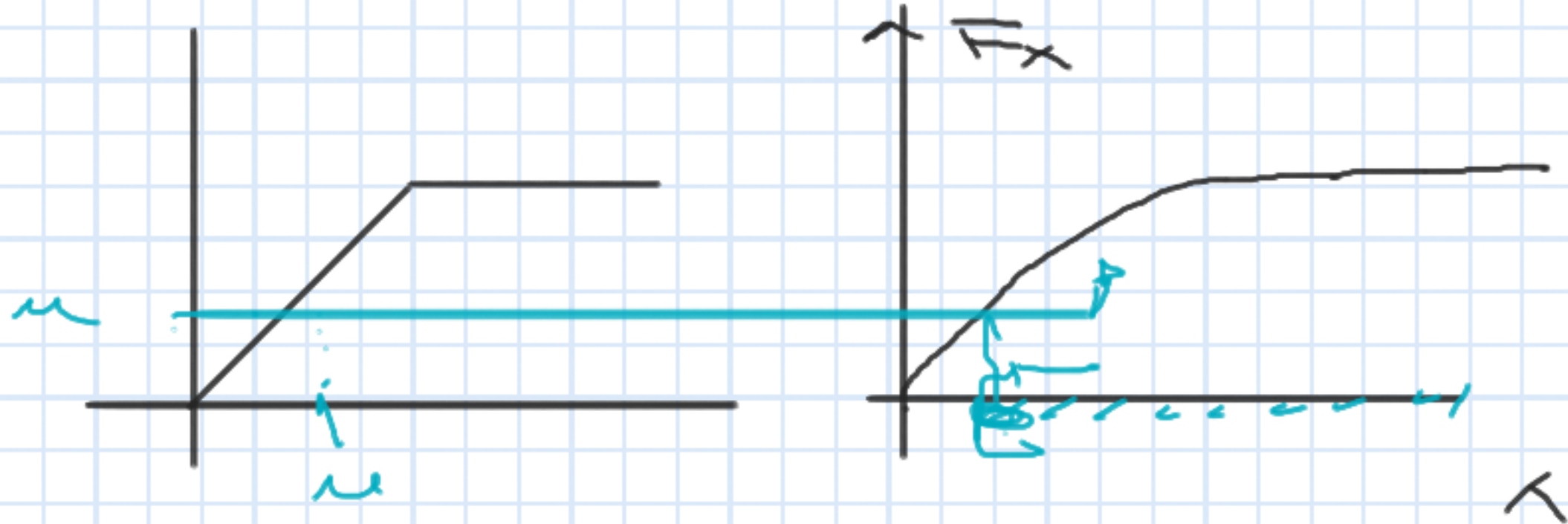


$$F_X(x) = P(X \leq x) \Rightarrow P(X \leq -10) = 0$$

$$P(X \leq 0.5) = P(X = 0)$$

$$P(X \leq 1.5) = 1$$

$$X = F^{-1}(u) = \begin{cases} 0 & 0 < u < 0.6 \\ 1 & 0.6 \leq u \leq 1 \end{cases}$$



$$F_x(x) = 1 - e^{-x}$$

$$F_x(x) = F_u(u)$$

$$F_x(x)$$

$$= u$$

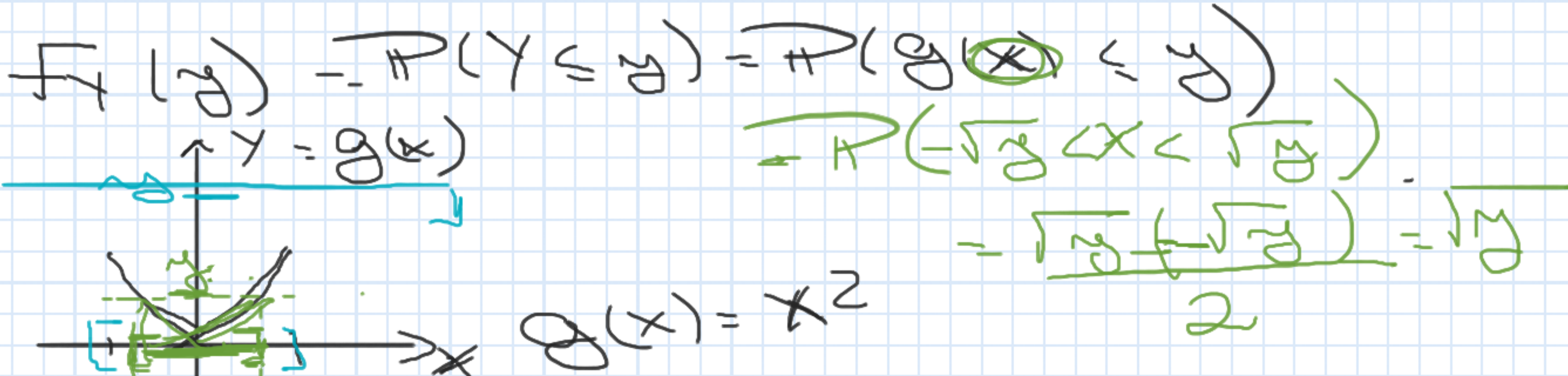
$$x = F_x^{-1}(u)$$

$$1 - e^{-x}$$

$$= u \rightarrow e^{-x} = 1 - u$$

$$x = -\ln(1 - u)$$

Beispiel $f = f_g(x) \Rightarrow f_g(y)?$
 Beispiel \times (oder $f_x(x)$) unendlich



$$\begin{aligned}
 &P(y \leq g(x)) = P(g(x) \leq y) \\
 &P(-\sqrt{y} < x < \sqrt{y}) \\
 &= \frac{\sqrt{y} - (-\sqrt{y})}{2} = \sqrt{y}
 \end{aligned}$$

$$\begin{aligned}
 &P(y < 1) \\
 &= \int_{-\sqrt{y}}^{\sqrt{y}} f(x) dx \\
 &= \int_{-\sqrt{y}}^{\sqrt{y}} x^2 dx \\
 &= \left[\frac{x^3}{3} \right]_{-\sqrt{y}}^{\sqrt{y}} \\
 &= \frac{(\sqrt{y})^3}{3} - \frac{(-\sqrt{y})^3}{3} \\
 &= \frac{y\sqrt{y}}{3} + \frac{y\sqrt{y}}{3} \\
 &= \frac{2y\sqrt{y}}{3}
 \end{aligned}$$

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 &= \frac{(\sqrt{y})^3}{3} - \frac{(-\sqrt{y})^3}{3} \\
 &= \frac{y\sqrt{y}}{3} + \frac{y\sqrt{y}}{3} \\
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 \end{aligned}$$