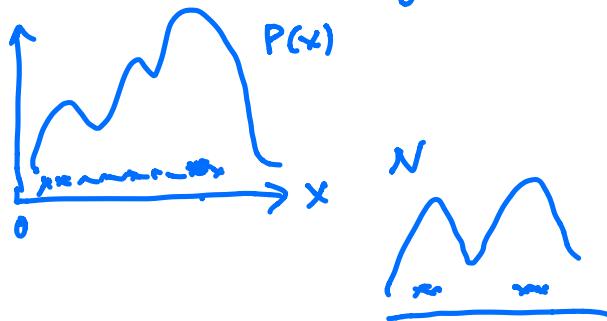


Woo Sampling from a probability distribution

X - variable

$P(x)$ - probability density funcn.

$\{x_i\}_{i=1}^N, x_1, \dots, x_N \sim$ according to the $P(x)$



N large

$$\langle f(x) \rangle_p = \int f(x) \cdot p(x) dx$$

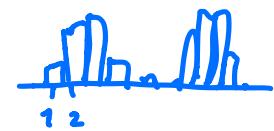
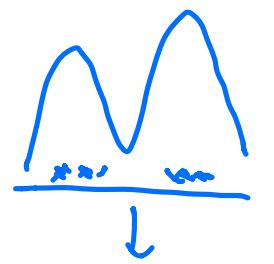
$$\{x_i\}_{i=1}^N \quad \simeq \frac{1}{N} \sum_{i=1}^N f(x_i)$$

$$\simeq \frac{1}{N} \sum_b N_b f(x_b)$$

$$= \sum_b \left(\frac{N_b}{N} \right) f(x_b) \sim p(x_b)$$

$f(x) = x$

$\mu = \langle x \rangle = \frac{1}{N} \sum_i x_i$



$$f(x) = (x - \mu)^2$$

$$\langle f(x) \rangle = \frac{1}{N} \sum_i (x_i - \mu)^2$$

$$= \sigma^2$$

Why sampling?

i) to generate alternative hypothesis.

$\{x_i\}$ - model $P(x_i | \text{model})$

data fit H_0 samples truly?



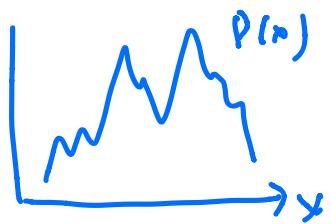
ii) $P(x)$ is complicated

$\{x_i\}$ allows do averages of any function.

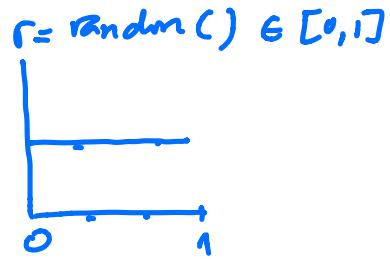
$$\tilde{\langle f(x) \rangle} = \frac{1}{N} \sum_i f(x_i)$$

iii) $P(y)$ $P(m)$ \rightarrow single parameter value M .

Main point: an arbitrary $P(x)$



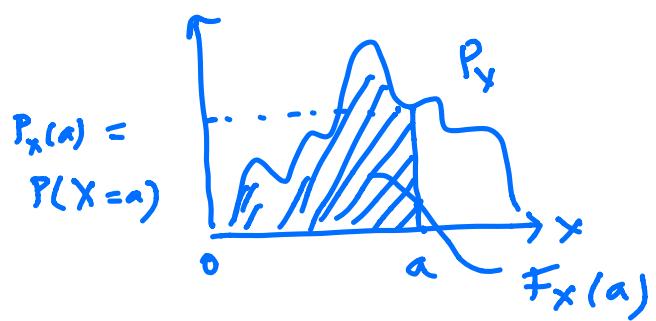
$\{X_i\}$ from $P \rightarrow$ Uniform distibn $[0,1]$



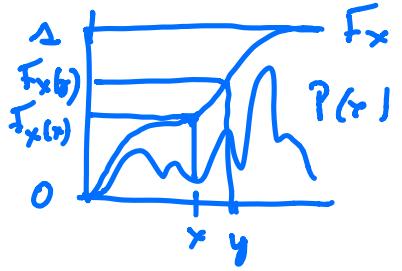
PDF, CDF

PDF = probability density function $P_X(x) = P(\tilde{x}=x)$

CDF = cumulative " " " $F_X(x) = P(\underline{x} \leq x)$



$$F_X(a) = \int_0^a P(x) dx$$
$$= \sum_{x \leq 0}^a P(x)$$



$$\lim_{x \rightarrow -\infty} F_X(x) = 0$$

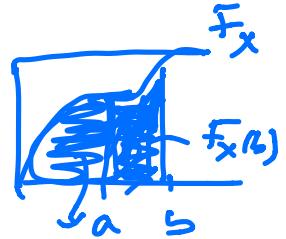
$$\lim_{x \rightarrow +\infty} F_X(x) = 1$$

• always increasing.

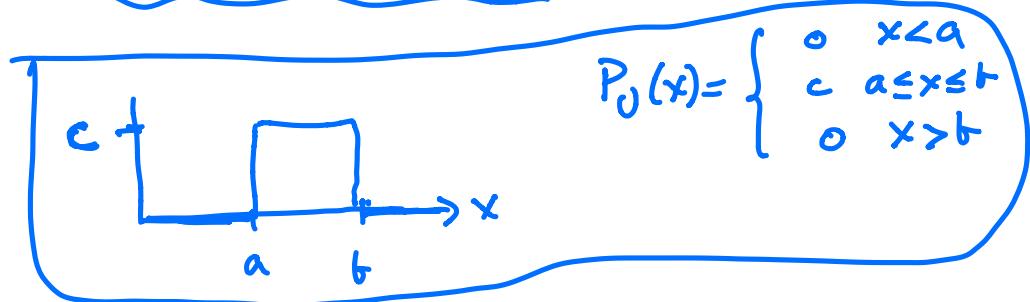
$$x \leq y \quad F_X(x) \leq F_X(y)$$

$$F_X(a) = P(X \leq a)$$

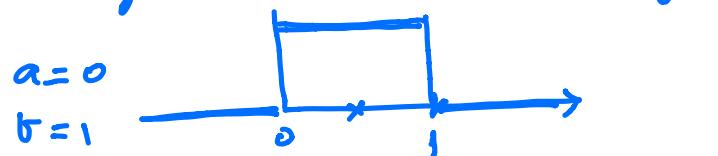
$$P(a \leq X \leq b) = F_X(b) - F_X(a)$$



The Uniform Distribution U in $x \in [a, b]$



$$a \leq x \leq b \quad P_U(x) = P_U(y) = c$$



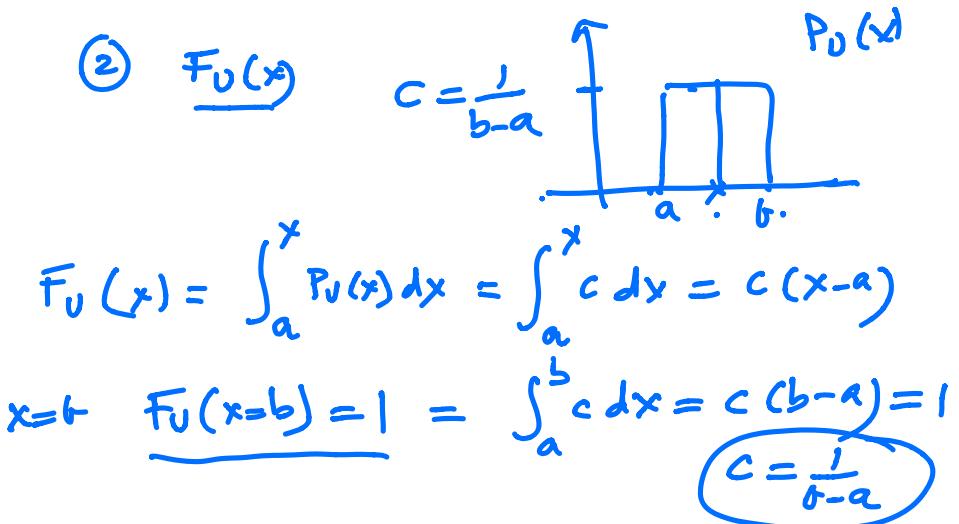
① c (a, b) ?

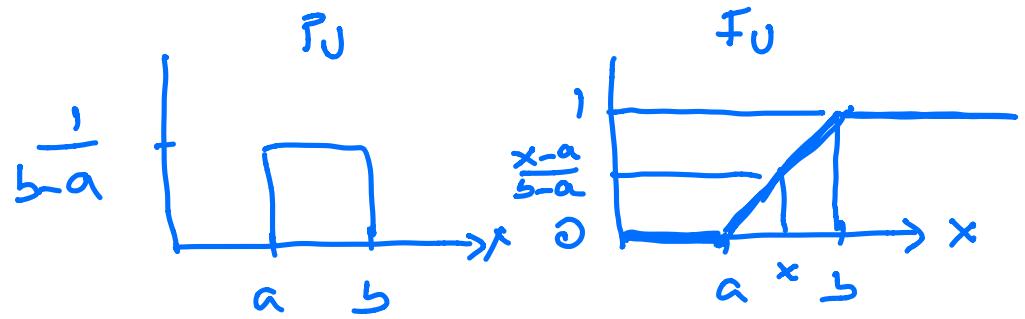
② $F_U(x)$

$$U \quad P_U(x) = \begin{cases} 0 & x < a \\ c & a \leq x \leq b \\ 0 & x > b \end{cases}$$

③? $F_U(x)$

$$\textcircled{1} \quad \stackrel{c=1}{=} P_V(x) = \begin{cases} 0 & x < a \\ c & a \leq x \leq b \\ 0 & x > b \end{cases}$$



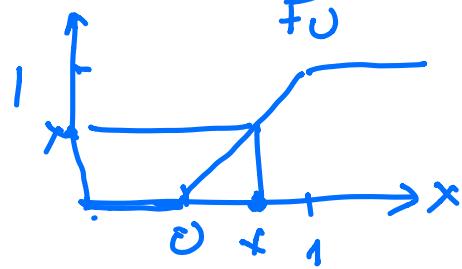
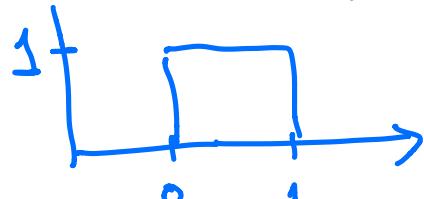


$$F_U(x) = c(x-a) = \frac{x-a}{b-a} \quad \begin{cases} x=a & F_U=0 \\ x=b & F_U=1 \end{cases}$$

$a=0$] random

$a=0$ Standard uniform dist $[0,1]$

$b=1$ $P_U(x) = 1 \quad (0 \leq x \leq 1)$

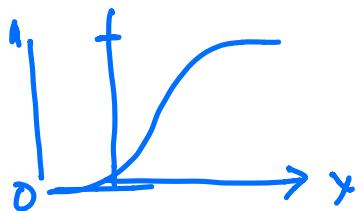


$$F_U(x) = \frac{x-a}{b-a} = x$$

$$\left\{ \begin{array}{l} F_U(x) = x \\ 0 \leq x \leq 1 \end{array} \right.$$

$$X \quad F_X \quad 0 \leq \underline{F_X(x)} \leq 1$$

CDF - random #



$$P(0 \leq F_X(x)) = \bar{F}_X(x)$$

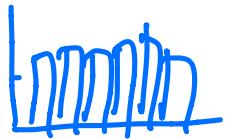
$$P_U(x) = x$$

$\{\bar{F}_X(x)\}$ is $U[0,1]$

$F_X(x)$ random variable.

$\{x_1, \dots, x_N\} P(x) \cdot$ 

$$\underline{u}_1 = F_X(x_1), \dots, \underline{u}_i = F_X(x_i), \dots, \underline{u}_N = F_X(x_N)$$

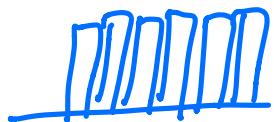




$F_E(x)$

$\underline{y}_S = \text{scipy.stats.expon.cdf}(x_i)$

$\underline{hist}(y_S)$



The Inverse Transformation Method

