
NASDE 22.10.2024



Today: more examples on inversion ①

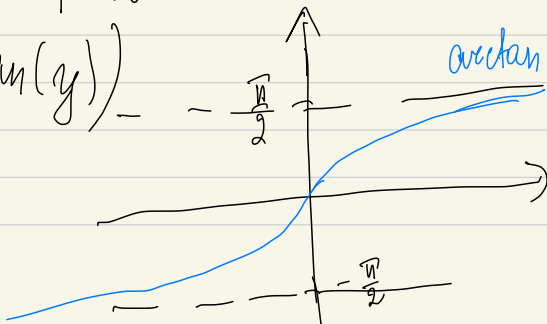
- Acceptance-rejection method
- Methods for normal distribution

⑥ Cauchy distribution:

density : $f(x) = \frac{1}{\pi} \cdot \frac{1}{1+x^2}$, $x \in \mathbb{R}$

$$F(x) = \int_{-\infty}^x f(y) dy = \frac{1}{\pi} \int_{-\infty}^x \frac{1}{1+y^2} dy$$
$$= \frac{1}{\pi} \arctan(y) \Big|_{-\infty}^x = \frac{1}{\pi} (\arctan(x))$$

- $\lim_{y \rightarrow \infty} \arctan(y) = \frac{\pi}{2}$



$$= \frac{1}{\pi} (\arctan(x) + \frac{\pi}{2}) \quad \text{②}$$

$$= \frac{1}{\pi} \arctan(x) + \frac{1}{2} \quad \text{— injective}$$

$$I_F(y) = F^{-1}(y) = \tan\left(\pi\left(y - \frac{1}{2}\right)\right)$$

$$y = \frac{1}{\pi} \arctan(x) + \frac{1}{2}, \quad \pi\left(y - \frac{1}{2}\right) = \arctan(x)$$

$$\text{Cauchy d-n} \sim \tan\left(\pi\left(u - \frac{1}{2}\right)\right)$$

⑦ Discrete d-n:

$$P(X=n) = p_n, \quad \forall n=0,1,\dots$$

$$\sum_{n=0}^{\infty} p_n = 1$$

$$p_n \in [0,1]$$

$$F(x) = P(X \leq x) \quad k \leq x < k+1 \quad (3)$$

$$= P(X < k+1)$$

$$= P(\{X=0\} \cup \{X=1\} \cup \dots \cup \{X=k\})$$

$$= P(X=0) + P(X=1) + \dots + P(X=k)$$

$$= \sum_{n=0}^k P_n$$

$$F(x) = \sum_{n=0}^{\lfloor x \rfloor} P_n$$

$\lfloor x \rfloor$ - largest integer smaller than x

$$\lfloor 1,2 \rfloor = 1, \quad \lfloor 1,9 \rfloor = 1, \quad \lfloor 2 \rfloor = 2$$

$$\lceil x \rceil, \quad \lceil 1,2 \rceil = 2 \quad \leftarrow \text{round up}$$

$$I_F(y) = \inf\{x \in \mathbb{R} : F(x) \geq y\}$$

$$F \text{ is piecewise constant} \quad (4)$$

$$= \min\{k \in \mathbb{N}_0 : F(k) \geq y\}$$

$$= \min\{k \in \mathbb{N}_0 : \sum_{n=0}^k P_n \geq y\}$$

Bernoulli d_n :

$$P_0 = 1-p, \quad P_1 = p, \quad P_2 = P_3 = \dots = 0$$

$$I_F(y) = \begin{cases} 0, & \text{if } 0 < y \leq p_0 \\ 1, & \text{if } 1-p < y < 1 \end{cases}$$

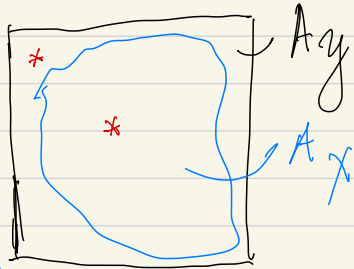
Algorithm:

Generate $U \sim U(0,1)$

$$X = \begin{cases} 0, & \text{if } U \leq 1-p \\ 1, & \text{if } 1-p < U < 1 \\ & = \text{otherwise} \end{cases}$$

- Acceptance-rejection method: (5)

Idea - "Generate complex d-n from a simpler d-n"



Let $Z_x \sim \mathcal{U}(A_x)$
 $Z_y \sim \mathcal{U}(A_y)$

where $A_x \subseteq A_y$
 $0 < \lambda(A_x) \leq \lambda(A_y) < \infty$

Algorithm: (6)

- Generate a realization

$$Z_y \sim \mathcal{U}(A_y)$$

If $Z_y \in A_x$

$$Z_x = Z_y \leftarrow \text{accept}$$

if not restart

Lemma (A-R on $A_x \subseteq A_y$)

$$Z_x \sim \mathcal{U}(A_x)$$

Proof: Take $\forall B \in \mathcal{B}(\mathbb{R}^2)$. Then

$$P(Z_x \in B) = P(Z_y \in B \mid Z_y \in A_x)$$

$$= \frac{P(Z_y \in B \cap A_x)}{P(Z_y \in A_x)}$$

$$P(z_y \in A_x) = \frac{1}{\kappa(A_y)} \int_{A_x} 1 dz$$

$$= \frac{\kappa(A_x)}{\kappa(A_y)}$$

$$P(z_y \in B \cap A_x) = \frac{1}{\kappa(A_y)} \int_{B \cap A_x} 1 dz$$

$$= \frac{1}{\kappa(A_y)} \int_B 1_{A_x}(z) dz$$

here

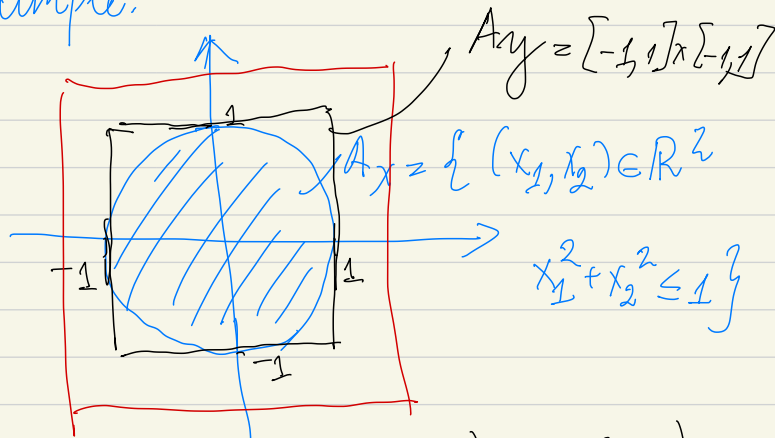
$$1_A(y) = \begin{cases} 1, & \text{if } y \in A \\ 0, & \text{otherwise} \end{cases}$$

$$\frac{P(z_y \in B \cap A_x)}{P(z_y \in A_x)} = \frac{1}{\kappa(A_x)} \int_B 1_{A_x}(z) dz$$

$$\textcircled{7} = P(u(A_x) \in B)$$

$$\textcircled{8}$$

Example:



Generate $u = (u_1, u_2) \sim U(A_y)$

Check: $u_1^2 + u_2^2 \leq 1$

if yes accept, otherwise restart

→ Extensions to non-uniform d. us (under certain conditions) also exist.

Normally d - d random variables (9)

→ Inversion method

$\Phi^{-1}(u)$ ← no analytic formula

In Matlab, `erfinv` - costly to compute

- Central limit theorem

If $(Y_n, n \in \mathbb{N})$ are iid (independent identically d - d) with

$E[|Y_1|^{2d}] < \infty$ and $\text{Var}(Y_1) > 0$

then

$$\frac{\sum_{n=1}^N Y_n - N \cdot E[Y_1]}{\sqrt{N \cdot \text{Var}(Y_1)}} \xrightarrow[N \rightarrow \infty]{d-n} \mathcal{N}(0, 1)$$

convergence in d - n .

$$V_1 \sim \mathcal{U}(0, 1) \quad (10)$$

Choose $N \sim$ large

$$U_1, \dots, U_N \sim \mathcal{U}(0, 1)$$

$$\sum_{n=1}^N U_n - N \cdot \frac{1}{2}$$

$$\sqrt{N \cdot \frac{1}{12}}$$

is close
in d - n
as
 $\mathcal{N}(0, 1)$

→ However, not exact.

→ (Box-Muller method)

Let us assume $X_1, X_2 \sim \mathcal{N}(0, 1)$ iid
 $X = (X_1, X_2) \overset{\mathbb{P}}{\sim} \mathcal{N}(0, \mathbb{I}_{\mathbb{R}^2})$

$g: \mathbb{R}^2 \rightarrow \mathbb{R}$ measurable: (11)

$$E[g(X_1, X_2)] = \frac{1}{2\pi} \iint_{\mathbb{R} \times \mathbb{R}} g(x, y) \cdot e^{-\frac{x^2+y^2}{2}} dx dy$$

Change coordinates to Polar coordinates

$$(x, y) = (r \cos \varphi, r \sin \varphi)$$

$$dx dy = r dr d\varphi$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \int_0^{\infty} g(r \cos \varphi, r \sin \varphi) \cdot e^{-\frac{r^2}{2}} \cdot r dr d\varphi$$

Again: $s = \frac{r^2}{2}$, $ds = r dr$

$$= \frac{1}{2\pi} \int_0^{2\pi} \int_0^{\infty} g(\sqrt{2s} \cdot \cos \varphi, \sqrt{2s} \sin \varphi) \cdot e^{-s} ds d\varphi$$

Again: $u_1 = e^{-s}$, $du_1 = -e^{-s} ds$ (12)

$$s = -\ln(u_1) \quad [e^{-s} \xrightarrow{s \rightarrow 0} 1]$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \int_0^1 g(\sqrt{-2 \ln(u_1)} \cdot \cos \varphi, \sqrt{-2 \ln(u_1)} \sin \varphi) du_1 d\varphi$$

Again (last one): $u_2 = \frac{1}{2\pi} \varphi$

$$du_2 = \frac{1}{2\pi} d\varphi$$

$$= \frac{1}{2\pi} \int_0^1 \int_0^1 g(\sqrt{-2 \ln(u_1)} \cdot \cos(2\pi u_2), \sqrt{-2 \ln(u_1)} \sin(2\pi u_2)) du_1 du_2$$

$$\text{Def } f: (0,1) \times (0,1) \rightarrow \mathbb{R}^2 \quad (13)$$

$$f(u_1, u_2) = (\sqrt{-2 \ln(u_1)} \cos(2\pi u_2), \sqrt{-2 \ln(u_1)} \sin(2\pi u_2))$$

$$= \mathbb{E} [g(f(\tilde{U}_1, \tilde{U}_2))] \quad \text{where}$$
$$\tilde{U}_1, \tilde{U}_2 \sim \mathcal{U}_{(0,1)} \quad \text{i.i.d.}$$

We have showed

$$\mathbb{E} [g(X_1, X_2)] = \mathbb{E} [g(f(\tilde{U}_1, \tilde{U}_2))]$$

$\forall g$ measurable.

$\forall B \in \mathcal{B}(\mathbb{R}^2)$, $g = \mathbb{1}_B$ -measurable

$$\Rightarrow \mathbb{P}((X_1, X_2) \in B) \quad (14)$$

$$= \mathbb{P}(f(\tilde{U}_1, \tilde{U}_2) \in B) \Rightarrow$$

$$(X_1, X_2) \stackrel{\mathcal{L}(\mathbb{R}^2)}{\sim} f(\tilde{U}_1, \tilde{U}_2)$$

Theorem (Box-Muller method)

Let $U_1, U_2: \Omega \rightarrow \mathbb{R}$ be i.i.d. $\mathcal{U}_{(0,1)}$

Define $X = (X_1, X_2): \Omega \rightarrow \mathbb{R}^2$ by

$$X = f(U_1, U_2)$$

Then

$$X \sim \mathcal{N}(0, I_{\mathbb{R}^2})$$

Next method:

Marsaglia Polar: uses the

following theorem: Let

$f: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ given by

$$f(x_1, x_2) = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \cdot \frac{1}{\sqrt{x_1^2 + x_2^2}} \sqrt{-2 \ln(x_1^2 + x_2^2)}$$

let $U: \Omega \rightarrow \mathbb{R}^2$ be $U(\tilde{B}(0,1))$
d-d where

$$\tilde{B}(0,1) = \{(x_1, x_2) \in \mathbb{R}^2 : x_1^2 + x_2^2 \in (0,1)\}$$

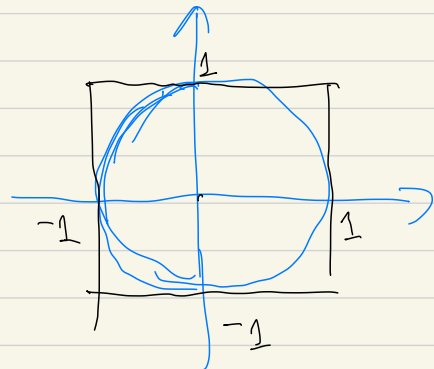
(open ball without the centre)

then $X = f(U) \sim W(0, I_{\mathbb{R}^2})$

(15)

A-R.

(16)



Generate $U_1, U_2 \sim U(0,1)$

$$U_1 = 2U_1 - 1, \quad U_2 = 2U_2 - 1$$

$$(U_1, U_2 \sim U(-1,1))$$

$$Q = U_1^2 + U_2^2$$

until $Q \in (0,1)$

$$W = \sqrt{-2 \ln Q / Q}$$

$$X_1 = U_1 \cdot W, \quad X_2 = U_2 \cdot W$$