

class poll: inversion for sampling from unit disc

want to generate (X, Y) uniform over the unit disc, i.e., over $\{(x, y) | x^2 + y^2 \leq 1\}$ given $U, V \sim U[0, 1]$ i.i.d rvs, which of the following gives the correct sample?

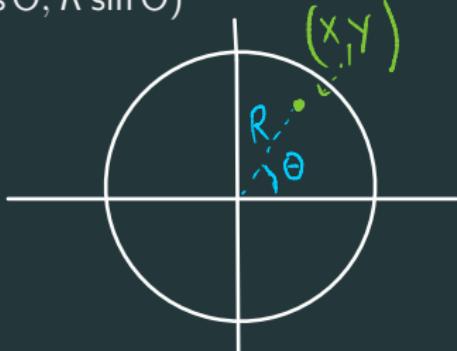
- (a) $R = U, \Theta = 2\pi V$ and $(X, Y) = (R \cos \Theta, R \sin \Theta)$

- (b) $R = \sqrt{U}, \Theta = 2\pi V$ and $(X, Y) = (R \cos \Theta, R \sin \Theta)$

- (c) $R = U^2, \Theta = 2\pi V$ and $(X, Y) = (R \cos \Theta, R \sin \Theta)$

- (d) $R = 2U - 1, \Theta = \pi V$ and $(X, Y) = (R \cos \Theta, R \sin \Theta)$
 $\Rightarrow \Theta \in [0, \pi]$ ✗

- (e) None of the above



solution: inversion for sampling from unit disc

want to generate (X, Y) uniform over the unit disc, i.e., over $\{(x, y) | x^2 + y^2 \leq 1\}$ given $U, V \sim U[0, 1]$ i.i.d rvs, which of the following gives the correct sample?

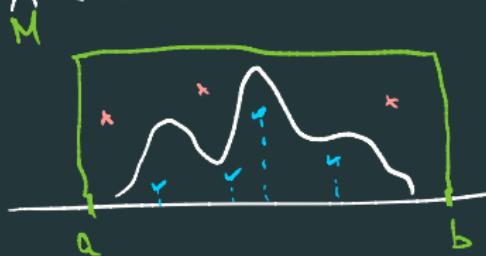


$$\begin{aligned} F_R(r) &= P[R \leq r] \\ &= \frac{\text{Area of } \odot \text{ of radius } r}{\text{Area of full } \odot} \\ &= \frac{\pi r^2}{\pi} = r^2 \text{ for } r \in [0, 1] \end{aligned}$$

Inversion - Want $R = F^{-1}(U) = \sqrt{U}$

Last lecture

- (inversion review - see recorded lecture)
- Acceptance-Rejection - main 'primitive' for simulation
 - Vanilla AR = bound $f(x)$ in a rectangle

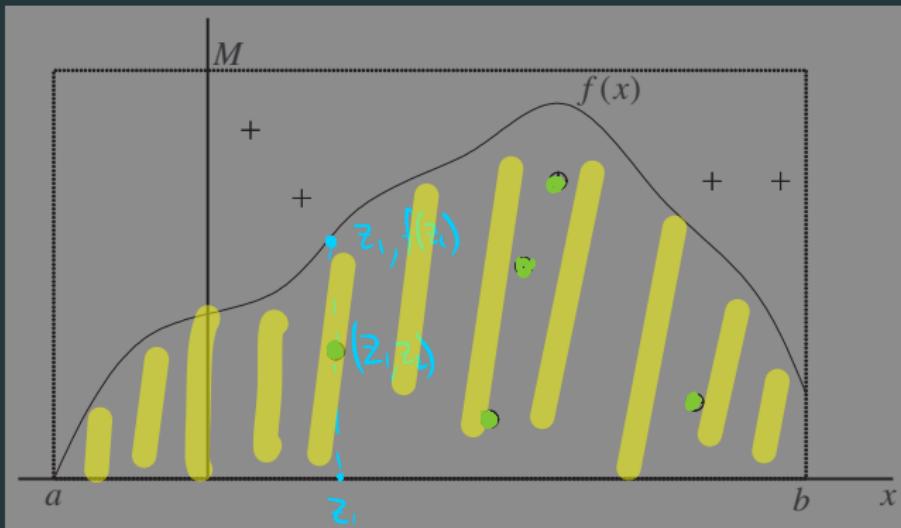


- Sample pts uniformly in rect /
- Accept if below $f(x)$
- Output x-coordinate

acceptance-rejection

Easy claim - Accepted pts (z_1, z_2) are uniformly distributed between $y=0$ and $y=f(x)$

want samples of a rv $X \in [a, b]$, with pdf $f(x) \leq M$



acceptance-rejection sampling

1. generate $U_1, U_2 \sim U[0, 1]$ $\underbrace{Z_1 \sim \text{Unif}[a, b]}_{\text{if } Z_2 \sim \text{Unif}[0, M]}$ $(Z_1, Z_2) \sim \text{Unif}$ in rectangle
2. set $Z_1 = a + (b - a)U_1$, $Z_2 = MU_2$
3. if $Z_2 \leq f(Z_1)$, return $X_o = Z_1$; else, reject and repeat

AR sampling: running time

how many $U[0, 1]$ samples do we need for one sample of X ?

$$\cdot \mathbb{P}\left[\underbrace{(z_1, z_2)}_{\text{'trial'}} \text{ accepted}\right] = \mathbb{P}\left[z_2 \leq f(z_1)\right] = \frac{1}{M(b-a)}$$

Area under $f(z)$
Area of rectL

Q: Define $N \equiv \# \text{ of } U[0, 1] \text{ samples required to get a single } X$
 $= 2R$, where $R = \# \text{ of 'trials' till we accept a sample}$

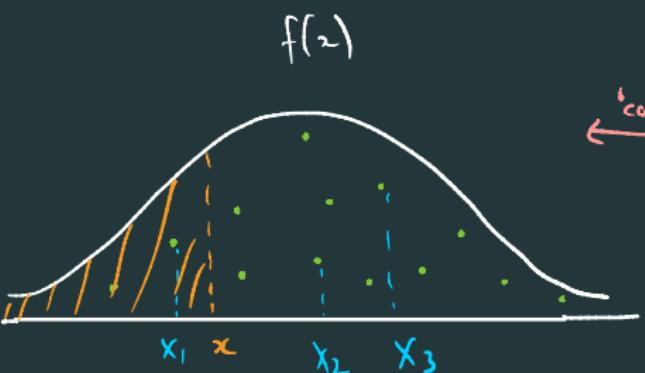
$$\Rightarrow \boxed{N \sim 2 \cdot \text{Geom}\left(\frac{1}{M(b-a)}\right)} = \# \text{ of trials till 'success'} \\ \sim \text{Geom}\left(\frac{1}{M(b-a)}\right)$$

$$\Rightarrow \mathbb{E}[N] = 2 \cdot \underline{M(b-a)}$$

Note - bigger rectangle
 \Rightarrow more samples needed

Claim - If you give me pt (X, Y) which is uniformly distributed between $y=0$ and $y=f(z)$ (for some pdf $f(\cdot)$), then $X \sim F(z) = \int_{-\infty}^z f(z)dz$

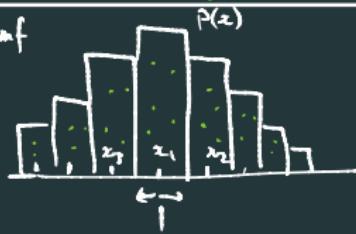
distributed according to CDF F



'Calculus'

'Bin' proof' - Pmf

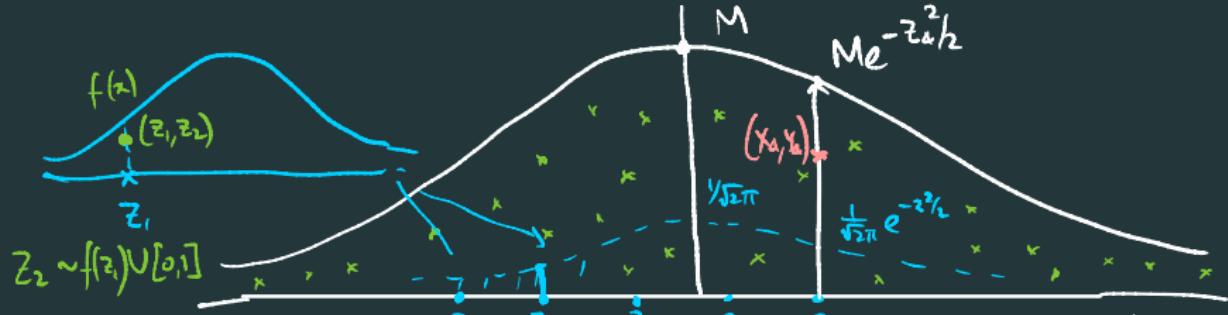
of pts in
rectangle $x < z$



Actual proof - $P[X \leq z] = \frac{\text{Area to the left of } z}{\text{Total area}}$

$$\Rightarrow P[X \leq z] = \frac{\int_{-\infty}^z f(z)dz}{\int_{-\infty}^{\infty} f(z)dz} = F(z)$$

Related question - Consider $g(x) = M e^{-x^2/2}$, $x \in \mathbb{R}$



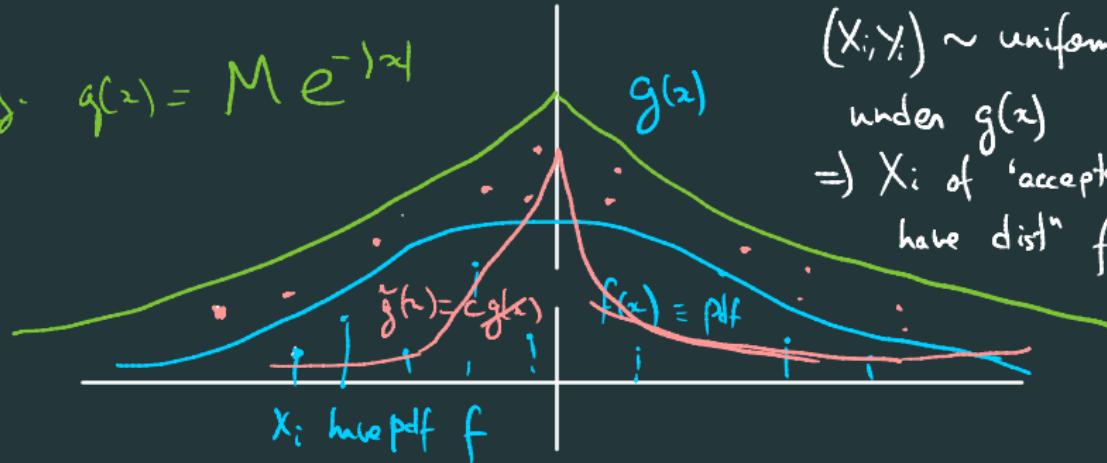
- Suppose I have samples $z_1, z_2, \dots \sim N(0,1)$
and $U_1, U_2, \dots \sim U[0,1]$

• Set $\boxed{X_i = z_i, Y_i = M e^{-z_i^2/2} U_i}$

Then (X_i, Y_i) are uniformly distributed between $y=0$
and $y=g(x)$

Generalized AR

E.g. $g(z) = M e^{-|z|}$



- If we have pts $(x_i, y_i) \sim \text{uniform}$ under $g(z)$
 $\Rightarrow X_i$ of 'accepted pts' have dist' $f(z)$

- How do we get uniform samples under $g(z)$?

- Sample $X_i \sim \frac{g(z)}{\int_{-\infty}^{\infty} g(z) dz}, Y_i \sim g(x_i) \cdot U_i$

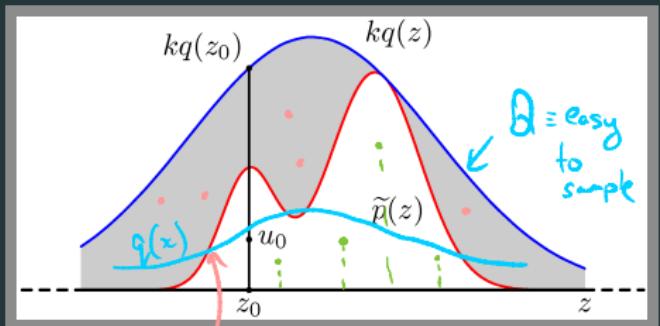
find a pdf

which has the same shape
as $g(z)$ (ie $\tilde{g}(z) = c g(z)$)

$$\int_{-\infty}^{\infty} g(z) dz$$

in example, $\tilde{g}(z) = \frac{1}{2} e^{-|z|}$
 'Laplace dist'
 i.e., 2-sided exponential

generalized AR sampling



difficult to sample pdf

- General process
- Sample $X_i \sim Q$ ^{easy to sample distn}

- Find some k such that

$$\frac{kq(x)}{\tilde{p}(x)} \geq 1 \text{ for all } x$$

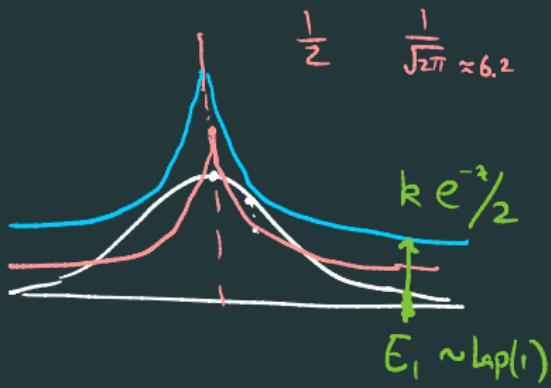
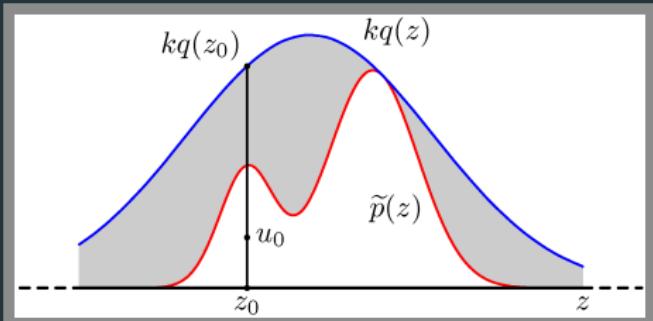
(ie, $kq(x)$ 'covers' $\tilde{p}(x)$)

(ie, set $k \geq \max_x \tilde{p}(x)/q_v(x)$)

- [Sample $Y_i \sim kq(X_i) \cdot U_i \Rightarrow (X_i, Y_i) \sim \text{Unif under } kq(x)$]
- [Accept X_i if $Y_i \leq \tilde{p}(X_i) \rightarrow kq(X_i) U_i \leq \tilde{p}(X_i)$]

ie, accept X_i with probability $\frac{\tilde{p}(X_i)}{kq(X_i)}$

generalized AR sampling

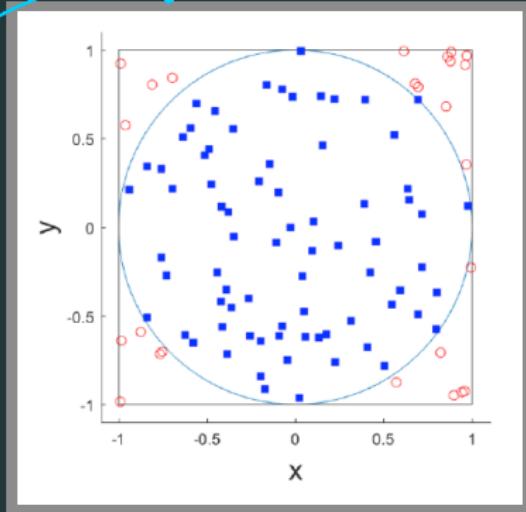


$$k \geq \max_x \frac{\hat{p}(x)}{q_b(x)}$$

$$\frac{1}{2} e^{-x}$$

(even more) generalized AR sampling

Given set A
which is easy to sample from, and set $B \subseteq A$ which is hard to sample from



Acceptance - Rejection Principle

dist' of $X_i \mid X_i \in B \equiv \text{Unif over } B$

↑
unif samples from A
have fn which tests membership in B

class poll: ordering conditional expectations

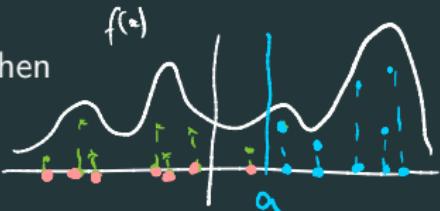
consider rv $X \in \mathbb{R}$ with cdf F , and any $a \in \mathbb{R}$; then

(a) $\mathbb{E}[X] \geq \mathbb{E}[X|X \geq a]$

(b) $\mathbb{E}[X] \leq \mathbb{E}[X|X \geq a]$

(c) depends on if a is positive or negative

(d) depends both on a and F



- AR way of thinking

$$X_1, X_2, \textcolor{red}{X_3}, \dots, \textcolor{red}{X_n} \sim F$$

$$(\mathbb{E}[x] \stackrel{\text{use}}{\approx} \frac{1}{n} \sum_i x_i)$$

Via AR
principle

$$\begin{matrix} X_1, X_2, \dots, X_n \\ | \quad | \quad | \quad | \quad | \quad \dots \\ Y_1, Y_2, \dots, Y_m \end{matrix} \quad \begin{matrix} \dots \\ \downarrow \\ Y_m \dots Y_n \end{matrix}$$

$$\mathbb{E}[x|x > a] = \mathbb{E}[y] \stackrel{\text{use}}{\approx} \frac{1}{n} \sum_i y_i$$