# Drawing from distributions 

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## Discrete distributions

- Let $X$ be a discrete r.v. with pmf:

$$
P\left(X=x_{i}\right)=p_{i}, i=0, \ldots,
$$

where $\sum_{i} p_{i}=1$.

- The support can be finite or infinite.
- The following algorithm generates draws from this distribution

1. Let $r$ be a draw from $U(0,1)$.
2. Initialize $k=0, p=0$.
3. $p=p+p_{k}$.
4. If $r<p$, set $X=x_{k}$ and stop.
5. Otherwise, set $k=k+1$ and go to step 3.

Inverse transform method

## Inverse Transform Method: illustration



## Discrete distributions

Acceptance-rejection technique

- Attributed to von Neumann.
- Mostly useful with continuous distributions.
- We want to draw from $X$ with pmf $p_{i}$.
- We know how to draw from $Y$ with pmf $q_{i}$.

Define a constant $c \geq 1$ such that

$$
\frac{p_{i}}{q_{i}} \leq c \forall i \text { s.t. } p_{i}>0
$$

Algorithm:

1. Draw $y$ from $Y$
2. Draw $r$ from $U(0,1)$
3. If $r<\frac{p_{y}}{c q_{y}}$, return $x=y$ and stop. Otherwise,
C.start again.
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## Acceptance-rejection: analysis

Probability to be accepted during a given iteration:

$$
\begin{array}{rlrl}
P(Y=y, \text { accepted }) & & P(Y=y) & P(\text { accepted } \mid Y=y) \\
& =q_{y} & & p_{y} / c q_{y} \\
& =\frac{p_{y}}{c} &
\end{array}
$$

Probability to be accepted:

$$
\begin{aligned}
P(\text { accepted }) & =\sum_{y} P(\text { accepted } \mid Y=y) P(Y=y) \\
& =\sum_{y} \frac{p_{y}}{c q_{y}} q_{y} \\
& =1 / c .
\end{aligned}
$$

Probability to draw $x$ at iteration $n$

$$
P(X=x \mid n)=\left(1-\frac{1}{c}\right)^{n-1} \frac{1}{c}
$$

## Acceptance-rejection: analysis

Therefore,

$$
\begin{aligned}
P(X=x) & =\sum_{n=1}^{+\infty} P(X=x \mid n) \\
& =\sum_{n=1}^{+\infty}\left(1-\frac{1}{c}\right)^{n-1} \frac{p_{x}}{c} \\
& =c \frac{p_{x}}{c} \\
& =p_{x}
\end{aligned}
$$

Reminder: geometric series

$$
\sum_{n=0}^{+\infty} x^{n}=\frac{1}{1-x}
$$

## Acceptance-rejection: analysis

Remarks:

- Average number of iterations: $c$
- The closer $c$ is to 1 , the closer the pmf of $Y$ is to the pmf of $X$.


## Continuous distributions

## Inverse Transform Method

## Idea:

- Let $X$ be a continuous r.v. with CDF $F_{X}(\varepsilon)$
- Draw $r$ from a uniform $U(0,1)$
- Generate $F_{X}^{-1}(r)$.

Motivation:

- $F_{X}$ is monotonically increasing
- It implies that $\varepsilon_{1} \leq \varepsilon_{2}$ is equivalent to $F_{X}\left(\varepsilon_{1}\right) \leq F_{X}\left(\varepsilon_{2}\right)$.


## Inverse Transform Method



## Inverse Transform Method

More formally:

- Denote $F_{U}(\varepsilon)=\varepsilon$ the CDF of the r.v. $U(0,1)$
- Let $G$ be the distribution of the r.v. $F_{X}^{-1}(U)$

$$
\begin{aligned}
G(\varepsilon) & =\operatorname{Pr}\left(F_{X}^{-1}(U) \leq \varepsilon\right) \\
& =\operatorname{Pr}\left(F_{X}\left(F_{X}^{-1}(U)\right) \leq F_{X}(\varepsilon)\right) \\
& =\operatorname{Pr}\left(U \leq F_{X}(\varepsilon)\right) \\
& =F_{U}\left(F_{X}(\varepsilon)\right) \\
& =F_{X}(\varepsilon)
\end{aligned}
$$

## Inverse Transform Method

Examples: let $r$ be a draw from $U(0,1)$

| Name | $F_{X}(\varepsilon)$ | Draw |
| :--- | :--- | :--- |
| Exponential $(b)$ | $1-e^{-\varepsilon / b}$ | $-b \ln r$ |
| $\operatorname{Logistic}(\mu, \sigma)$ | $1 /(1+\exp (-(\varepsilon-\mu) / \sigma))$ | $\mu-\sigma \ln \left(\frac{1}{r}-1\right)$ |
| $\operatorname{Power}(n, \sigma)$ | $(\varepsilon / \sigma)^{n}$ | $\sigma r^{1 / n}$ |

Note: the CDF is not always available (e.g. normal distribution).

## Continuous distributions

Rejection Method

- We want to draw from $X$ with pdf $f_{X}$.
- We know how to draw from $Y$ with pdf $f_{Y}$.

Define a constant $c \geq 1$ such that

$$
\frac{f_{X}(\varepsilon)}{f_{Y}(\varepsilon)} \leq c \forall \varepsilon
$$

Algorithm:

1. Draw $y$ from $Y$
2. Draw $r$ from $U(0,1)$
3. If $r<\frac{f_{X}(y)}{c f_{Y}(y)}$, return $x=y$ and stop. Otherwise, start again.

## Rejection Method: example

Draw from a normal distribution

- Let $\bar{X} \sim N(0,1)$ and $X=|\bar{X}|$
- Probability density function: $f_{X}(\varepsilon)=\frac{2}{\sqrt{2 \pi}} e^{-\varepsilon^{2} / 2}, 0<\varepsilon<+\infty$
- Consider an exponential r.v. with pdf $f_{Y}(\varepsilon)=e^{-\varepsilon}, 0<\varepsilon<+\infty$
- Then

$$
\frac{f_{X}(\varepsilon)}{f_{Y}(\varepsilon)}=\frac{2}{\sqrt{2 \pi}} e^{\varepsilon-\varepsilon^{2} / 2}
$$

- The ratio takes its maximum at $\varepsilon=1$, therefore

$$
\frac{f_{X}(\varepsilon)}{f_{Y}(\varepsilon)} \leq \frac{f_{X}(1)}{f_{Y}(1)}=\sqrt{2 e / \pi} \approx 1.315 .
$$

- Rejection method, with $\frac{f_{X}(\varepsilon)}{c f_{Y}(\varepsilon)}=\frac{1}{\sqrt{e}} e^{\varepsilon-\varepsilon^{2} / 2}=e^{\varepsilon-\frac{\varepsilon^{2}}{2}-\frac{1}{2}}=e^{-\frac{(\varepsilon-1)^{2}}{2}}$


## Rejection Method: example

Draw from a normal

```
    1. Draw \(r\) from \(U(0,1)\)
    2. Let \(y=-\ln (1-r)\) (draw from the exponential)
    3. Draw \(s\) from \(U(0,1)\)
    4. If \(s<e^{-\frac{(y-1)^{2}}{2}}\) return \(x=y\) and go to step 5 .
        Otherwise, go to step 1 .
    5. Draw \(t\) from \(U(0,1)\).
    6. If \(t \leq 0.5\), return \(x\). Otherwise, return \(-x\).
```

Note: this procedure can be improved. See Ross, Chapter 5.

## Draws from the exponential



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## Rejected draws


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## Accepted draws



## Rejected and accepted draws



## The polar method

Draw from a normal distribution

- Let $X \sim N(0,1)$ and $Y \sim N(0,1)$ independent
- pdf:

$$
f(x, y)=\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2} \frac{1}{\sqrt{2 \pi}} e^{-y^{2} / 2}=\frac{1}{2 \pi} e^{-\left(x^{2}+y^{2}\right) / 2} .
$$

- Let $R$ and $\theta$ such that $R^{2}=X^{2}+Y^{2}$, and $\tan \theta=Y / X$.



## The polar method

Change of variables (reminder):

- Let $A$ be a multivariate r.v. distributed with pdf $f_{A}(a)$.
- Consider the change of variables $b=H(a)$ where $H$ is bijective and differentiable
- Then $B=H(A)$ is distributed with pdf

$$
f_{B}(b)=f_{A}\left(H^{-1}(b)\right)\left|\operatorname{det}\left(\frac{d H^{-1}(b)}{d b}\right)\right|
$$

Here: $A=(X, Y), B=\left(R^{2}, \theta\right)=(T, \theta)$

$$
H^{-1}(B)=\binom{T^{\frac{1}{2}} \cos \theta}{T^{\frac{1}{2}} \sin \theta} \quad \frac{d H^{-1}(B)}{d B}=\left(\begin{array}{cc}
\frac{1}{2} T^{-\frac{1}{2}} \cos \theta & -T^{\frac{1}{2}} \sin \theta \\
\frac{1}{2} T^{-\frac{1}{2}} \sin \theta & T^{\frac{1}{2}} \cos \theta
\end{array}\right)
$$

## The polar method

$$
H^{-1}(b)=\binom{T^{\frac{1}{2}} \cos \theta}{T^{\frac{1}{2}} \sin \theta} \quad \frac{d H^{-1}(b)}{d b}=\left(\begin{array}{cc}
\frac{1}{2} T^{-\frac{1}{2}} \cos \theta & -T^{\frac{1}{2}} \sin \theta \\
\frac{1}{2} T^{-\frac{1}{2}} \sin \theta & T^{\frac{1}{2}} \cos \theta
\end{array}\right)
$$

Therefore,

$$
\left|\operatorname{det}\left(\frac{d H^{-1}(b)}{d b}\right)\right|=\frac{1}{2} .
$$

and

$$
f_{B}(T, \theta)=\frac{1}{2} \frac{1}{2 \pi} e^{-T / 2}, \quad 0<T<+\infty, \quad 0<\theta<2 \pi .
$$

Product of

- an exponential with mean $2: \frac{1}{2} e^{-T / 2}$
- a uniform on $[0,2 \pi[: 1 / 2 \pi$


## The polar method

Therefore,

- $R^{2}$ and $\theta$ are independent
- $R^{2}$ is exponential with mean 2
- $\theta$ is uniform on $(0,2 \pi)$

Algorithm:

1. Let $r_{1}$ and $r_{2}$ be draws from $U(0,1)$.
2. Let $R^{2}=-2 \ln r_{1}$ (draw from exponential of mean 2)
3. Let $\theta=2 \pi r_{2}$ (draw from $U(0,2 \pi)$ )
4. Let

$$
\begin{aligned}
X & =R \cos \theta=\sqrt{-2 \ln r_{1}} \cos \left(2 \pi r_{2}\right) \\
Y & =R \sin \theta=\sqrt{-2 \ln r_{1}} \sin \left(2 \pi r_{2}\right)
\end{aligned}
$$

## The polar method

Issue: time consuming to compute sine and cosine Solution: generate directly the sine and the cosine

- Draw a random point $\left(s_{1}, s_{2}\right)$ in the circle of radius one centered at $(0,0)$.
- How? Draw a random point in the square $[-1,1] \times[-1,1]$ and reject points outside the circle
- Let $(R, \theta)$ be the polar coordinates of this point.
- $R^{2} \sim U(0,1)$ and $\theta \sim U(0,2 \pi)$ are independent

$$
\begin{aligned}
R^{2} & =s_{1}^{2}+s_{2}^{2} \\
\cos \theta & =s_{1} / R \\
\sin \theta & =s_{2} / R
\end{aligned}
$$

## The polar method

Original transformation:

$$
\begin{aligned}
& X=R \cos \theta=\sqrt{-2 \ln r_{1}} \cos \left(2 \pi r_{2}\right) \\
& Y=R \sin \theta=\sqrt{-2 \ln r_{1}} \sin \left(2 \pi r_{2}\right)
\end{aligned}
$$

Replace $r_{1}$ by $t=R^{2} \sim U(0,1)$, and the sine and cosine as described above

$$
\begin{aligned}
t & =s_{1}^{2}+s_{2}^{2} \\
X & =R \cos \theta=\sqrt{-2 \ln t} \frac{s_{1}}{\sqrt{t}}=s_{1} \sqrt{\frac{-2 \ln t}{t}} \\
Y & =R \sin \theta=\sqrt{-2 \ln t} \frac{s_{2}}{\sqrt{t}}=s_{2} \sqrt{\frac{-2 \ln t}{t}}
\end{aligned}
$$

## The polar method

## Algorithm:

1. Let $r_{1}$ and $r_{2}$ be draws from $U(0,1)$.
2. Define $s_{1}=2 r_{1}-1$ and $s_{2}=2 r_{2}-1$ (draws from $U(-1,1))$.
3. Define $t=s_{1}^{2}+s_{2}^{2}$.
4. If $t>1$, reject the draws and go to step 1.
5. Return

$$
x=s_{1} \sqrt{\frac{-2 \ln t}{t}} \text { and } y=s_{2} \sqrt{\frac{-2 \ln t}{t}} .
$$

## Transformations of standard normal

- If $r$ is a draw from $N(0,1)$, then

$$
s=b r+a
$$

is a draw from $N\left(a, b^{2}\right)$

- If $r$ is a draw from $N\left(a, b^{2}\right)$, then

$$
e^{r}
$$

is a draw from a log normal $L N\left(a, b^{2}\right)$ with mean

$$
e^{a+\left(b^{2} / 2\right)}
$$

and variance

$$
e^{2 a+b^{2}}\left(e^{b^{2}}-1\right)
$$

## Multivariate normal

- If $r_{1}, \ldots, r_{n}$ are independent draws from $N(0,1)$, and

$$
r=\left(\begin{array}{c}
r_{1} \\
\vdots \\
r_{n}
\end{array}\right)
$$

- then

$$
s=a+L r
$$

is a vector of draws from the $n$-variate normal $N\left(a, L L^{T}\right)$, where

- $L$ is lower triangular, and
- $L L^{T}$ is the Cholesky factorization of the variance-covariance matrix


## Multivariate normal

## Example:

$$
\begin{gathered}
L=\left(\begin{array}{rrr}
\ell_{11} & 0 & 0 \\
\ell_{21} & \ell_{22} & 0 \\
\ell_{31} & \ell_{32} & \ell_{33}
\end{array}\right) \\
s_{1}=\ell_{11} r_{1} \\
s_{2}=\ell_{21} r_{1}+\ell_{22} r_{2} \\
s_{3}=\ell_{31} r_{1}+\ell_{32} r_{2}+\ell_{33} r_{3}
\end{gathered}
$$

## Transforming draws

- Consider draws from the following distributions:
- normal: $N(0,1)$ (draws denoted by $\xi$ below)
- uniform: $U(0,1)$ (draws denoted by $r$ below)
- Draws $R$ from other distributions are obtained from nonlinear transforms.
- Lognormal(a,b)

$$
f(x)=\frac{1}{x b \sqrt{2 \pi}} \exp \left(\frac{-(\ln x-a)^{2}}{2 b^{2}}\right) \quad R=e^{a+b \xi}
$$

## Transforming draws

- Cauchy(a,b)

$$
f(x)=\left(\pi b\left(1+\left(\frac{x-a}{b}\right)^{2}\right)\right)^{-1} \quad R=a+b \tan \left(\pi\left(r-\frac{1}{2}\right)\right)
$$

- $\chi^{2}(a)$ ( $a$ integer)

$$
f(x)=\frac{x^{(a-2) / 2} e^{-x / 2}}{2^{a / 2} \Gamma(a / 2)} \quad R=\sum_{j=1}^{a} \xi_{j}^{2}
$$

- Erlang(a,b) (b integer)

$$
f(x)=\frac{(x / a)^{b-1} e^{-x / a}}{a(b-1)!} \quad R=-a \sum_{j=1}^{b} \ln r_{i}
$$

## Transforming draws

- Exponential(a)

$$
F(x)=1-e^{-x / a} \quad R=-a \ln r
$$

- Extreme Value(a,b)

$$
F(x)=1-\exp \left(-e^{-(x-a) / b}\right) \quad R=a-b \ln (-\ln r)
$$

- Logistic(a,b)

$$
F(x)=\left(1+e^{-(x-a) / b}\right)^{-1} \quad R=a+b \ln \left(\frac{r}{1-r}\right)
$$

## Transforming draws

- Pareto(a,b)

$$
F(x)=1-\left(\frac{a}{x}\right)^{b} \quad R=a(1-r)^{-1 / b}
$$

- Standard symmetrical triangular distribution

$$
f(x)=\left\{\begin{array}{ll}
4 x & \text { if } 0 \leq x \leq 1 / 2 \\
4(1-x) & \text { if } 1 / 2 \leq x \leq 1
\end{array} \quad R=\frac{r_{1}+r_{2}}{2}\right.
$$

- Weibull(a,b)

$$
F(x)=1-e^{-\left(\frac{x}{a}\right)^{b}} \quad R=a(-\ln r)^{1 / b}
$$

## Appendix

Uniform distribution: $X \sim U(a, b)$

- pdf

$$
f_{X}(x)= \begin{cases}1 /(b-a) & \text { if } a \leq x \leq b \\ 0 & \text { otherwise }\end{cases}
$$

- CDF

$$
F_{X}(x)= \begin{cases}0 & \text { if } x \leq a \\ (x-a) /(b-a) & \text { if } a \leq x \leq b, \\ 1 & \text { if } x \geq b\end{cases}
$$

- Mean, median: $(a+b) / 2$.
- Variance: $(b-a)^{2} / 12$

