# **Drawing from distributions**

Michel Bierlaire

michel.bierlaire@epfl.ch

Transport and Mobility Laboratory





### Discrete distributions

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

$$P(X=x_i)=p_i, i=0,\ldots,$$

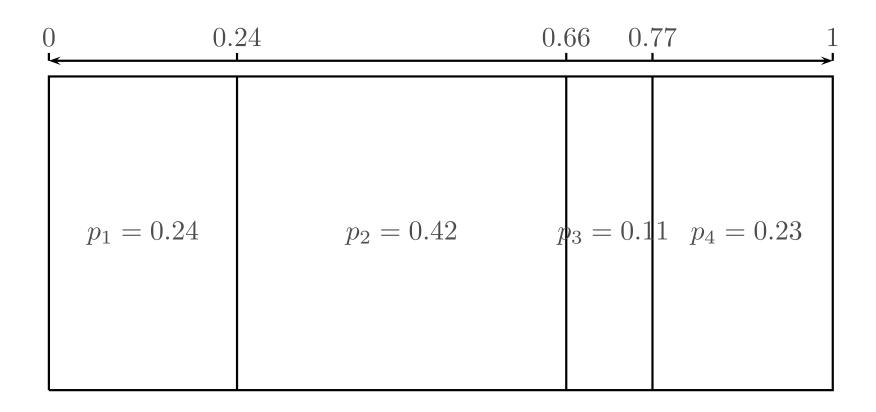
where  $\sum_{i} p_i = 1$ .

- 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

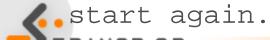
- Attributed to von Neumann.
- Mostly useful with continuous distributions.
- We want to draw from X with pmf  $p_i$ .
- We know how to draw from Y with pmf  $q_i$ .

Define a constant  $c \ge 1$  such that

$$\frac{p_i}{q_i} \le c \ \forall i \ \text{s.t.} \ p_i > 0.$$

#### Algorithm:

- 1. Draw y from Y
- 2. Draw r from U(0,1)
- 3. If  $r < \frac{p_y}{cq_y}$ , return x = y and stop. Otherwise,



# Acceptance-rejection: analysis

Probability to be accepted during a given iteration:

$$\begin{array}{rcl} P(Y=y, \mathsf{accepted}) & = & P(Y=y) & P(\mathsf{accepted}|Y=y) \\ & = & q_y & & p_y/cq_y \\ & = & \frac{p_y}{c} \end{array}$$

Probability to be accepted:

$$\begin{array}{ll} P(\mathsf{accepted}) &=& \sum_y P(\mathsf{accepted}|Y=y) P(Y=y) \\ &=& \sum_y \frac{p_y}{cq_y} q_y \\ &=& 1/c. \end{array}$$

Probability to draw x at iteration n







# Acceptance-rejection: analysis

Therefore,

$$P(X = x) = \sum_{n=1}^{+\infty} P(X = x | n)$$

$$= \sum_{n=1}^{+\infty} \left(1 - \frac{1}{c}\right)^{n-1} \frac{p_x}{c}$$

$$= c \frac{p_x}{c}$$

$$= p_x.$$

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:

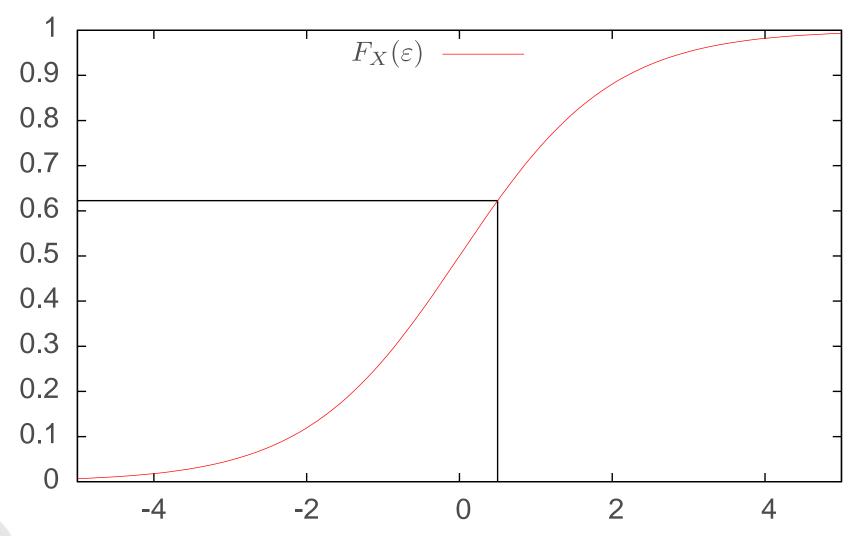
- Let X be a continuous r.v. with CDF  $F_X(\varepsilon)$
- Draw r from a uniform U(0,1)
- Generate  $F_X^{-1}(r)$ .

#### Motivation:

- $F_X$  is monotonically increasing
- It implies that  $\varepsilon_1 \leq \varepsilon_2$  is equivalent to  $F_X(\varepsilon_1) \leq F_X(\varepsilon_2)$ .



### **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.  $F_X^{-1}(U)$

$$G(\varepsilon) = \Pr(F_X^{-1}(U) \le \varepsilon)$$

$$= \Pr(F_X(F_X^{-1}(U)) \le F_X(\varepsilon))$$

$$= \Pr(U \le F_X(\varepsilon))$$

$$= F_U(F_X(\varepsilon))$$

$$= F_X(\varepsilon)$$



### **Inverse Transform Method**

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

Name	$F_X(arepsilon)$	Draw
Exponential(b)	$1 - e^{-\varepsilon/b}$	$-b \ln r$
-		
Logistic( $\mu$ , $\sigma$ )	$1/(1 + \exp(-(\varepsilon - \mu)/\sigma))$	$\mu - \sigma \ln(\frac{1}{\pi} - 1)$
		, , , , , , , , , , , , , , , , , , , ,
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  $c \ge 1$  such that

$$\frac{f_X(\varepsilon)}{f_Y(\varepsilon)} \le c \ \forall \varepsilon$$

#### Algorithm:

- 1. Draw y from Y
- 2. Draw r from U(0,1)
- 3. If  $r < \frac{f_X(y)}{cf_Y(y)}$ , 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

$$\frac{f_X(\varepsilon)}{f_Y(\varepsilon)} \le \frac{f_X(1)}{f_Y(1)} = \sqrt{2e/\pi} \approx 1.315.$$

• Rejection method, with  $\frac{f_X(\varepsilon)}{cf_Y(\varepsilon)} = \frac{1}{\sqrt{e}}e^{\varepsilon-\varepsilon^2/2} = e^{\varepsilon-\frac{\varepsilon^2}{2}-\frac{1}{2}} = e^{-\frac{(\varepsilon-1)^2}{2}}$ 



# Rejection Method: example

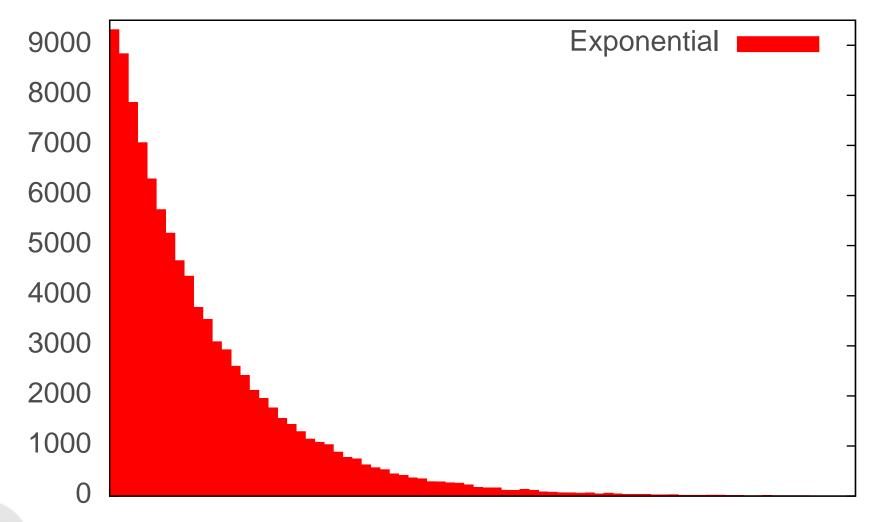
#### Draw from a normal

- 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  $t \le 0.5$ , return x. Otherwise, return -x.

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

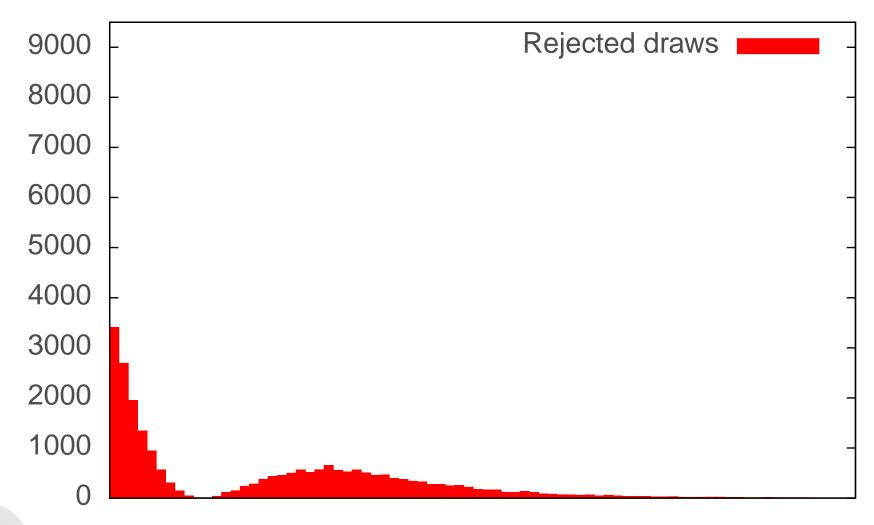


# Draws from the exponential

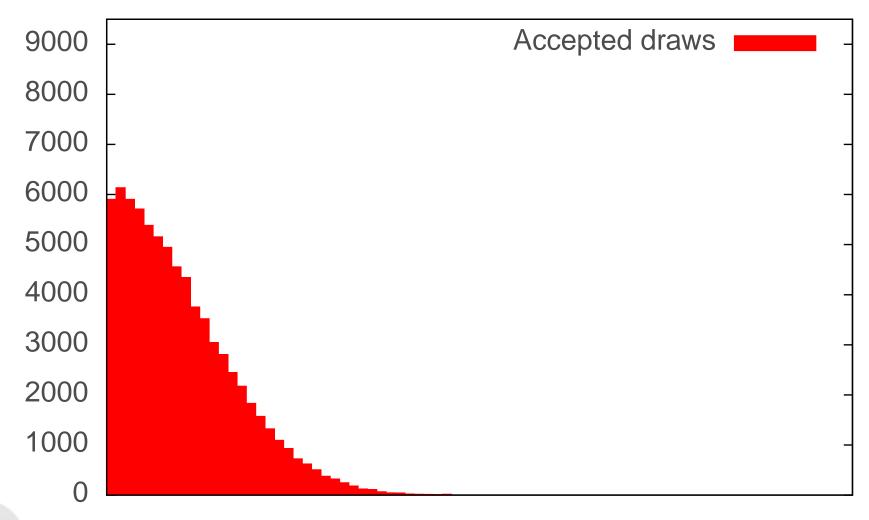




# Rejected draws

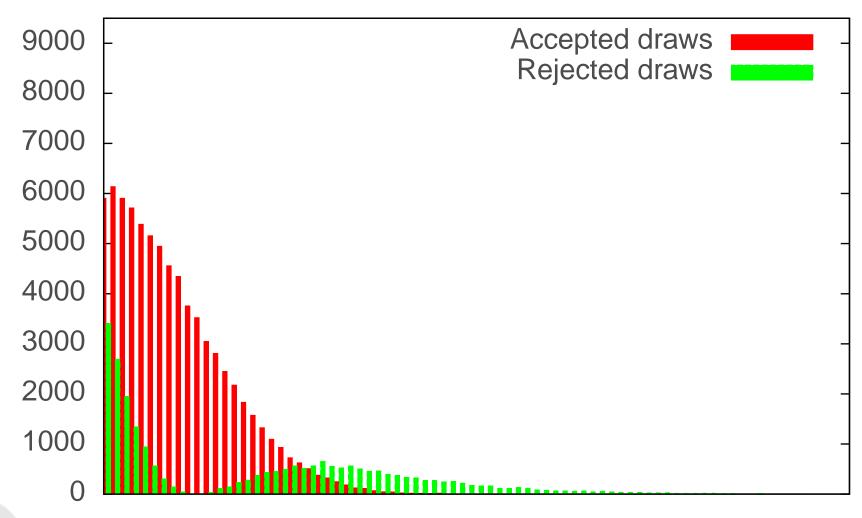


# **Accepted draws**





# Rejected and accepted draws



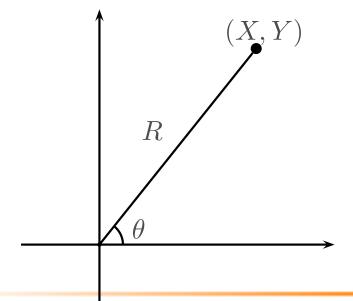


#### Draw from a normal distribution

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

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

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







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

$$f_B(b) = f_A(H^{-1}(b)) \left| \det \left( \frac{dH^{-1}(b)}{db} \right) \right|.$$

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

$$H^{-1}(B) = \begin{pmatrix} T^{\frac{1}{2}}\cos\theta \\ T^{\frac{1}{2}}\sin\theta \end{pmatrix} \frac{dH^{-1}(B)}{dB} = \begin{pmatrix} \frac{1}{2}T^{-\frac{1}{2}}\cos\theta & -T^{\frac{1}{2}}\sin\theta \\ \frac{1}{2}T^{-\frac{1}{2}}\sin\theta & T^{\frac{1}{2}}\cos\theta \end{pmatrix}$$



$$H^{-1}(b) = \begin{pmatrix} T^{\frac{1}{2}}\cos\theta \\ T^{\frac{1}{2}}\sin\theta \end{pmatrix} \frac{dH^{-1}(b)}{db} = \begin{pmatrix} \frac{1}{2}T^{-\frac{1}{2}}\cos\theta & -T^{\frac{1}{2}}\sin\theta \\ \frac{1}{2}T^{-\frac{1}{2}}\sin\theta & T^{\frac{1}{2}}\cos\theta \end{pmatrix}$$

Therefore,

$$\left| \det \left( \frac{dH^{-1}(b)}{db} \right) \right| = \frac{1}{2}.$$

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  $[0, 2\pi]$ :  $1/2\pi$





#### 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

$$X = R\cos\theta = \sqrt{-2\ln r_1}\cos(2\pi r_2)$$

$$Y = R\sin\theta = \sqrt{-2\ln r_1}\sin(2\pi r_2)$$





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

- Draw a random point  $(s_1, s_2)$  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

$$R^{2} = s_{1}^{2} + s_{2}^{2}$$

$$\cos \theta = s_{1}/R$$

$$\sin \theta = s_{2}/R$$



Original transformation:

$$X = R\cos\theta = \sqrt{-2\ln r_1}\cos(2\pi r_2)$$

$$Y = R\sin\theta = \sqrt{-2\ln r_1}\sin(2\pi r_2)$$

Replace  $r_1$  by  $t = R^2 \sim U(0,1)$ , and the sine and cosine as described above

$$t = s_1^2 + s_2^2$$

$$X = R\cos\theta = \sqrt{-2\ln t} \frac{s_1}{\sqrt{t}} = s_1 \sqrt{\frac{-2\ln t}{t}}$$

$$Y = R\sin\theta = \sqrt{-2\ln t} \frac{s_2}{\sqrt{t}} = s_2 \sqrt{\frac{-2\ln t}{t}}$$



#### Algorithm:

- 1. Let  $r_1$  and  $r_2$  be draws from U(0,1).
- 2. Define  $s_1=2r_1-1$  and  $s_2=2r_2-1$  (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

$$x = s_1 \sqrt{\frac{-2 \ln t}{t}}$$
 and  $y = s_2 \sqrt{\frac{-2 \ln t}{t}}$ .



### Transformations of standard normal

• If r is a draw from N(0,1), then

$$s = br + a$$

is a draw from  $N(a, b^2)$ 

• If r is a draw from  $N(a, b^2)$ , then

$$e^r$$

is a draw from a log normal  $LN(a,b^2)$  with mean

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

and variance

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





### 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

$$s = a + Lr$$

is a vector of draws from the n-variate normal  $N(a, LL^T)$ , where

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





### Multivariate normal

#### Example:

$$L = \begin{pmatrix} \ell_{11} & 0 & 0 \\ \ell_{21} & \ell_{22} & 0 \\ \ell_{31} & \ell_{32} & \ell_{33} \end{pmatrix}$$

$$s_1 = \ell_{11}r_1$$
  
 $s_2 = \ell_{21}r_1 + \ell_{22}r_2$   
 $s_3 = \ell_{31}r_1 + \ell_{32}r_2 + \ell_{33}r_3$ 



- 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)

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





Cauchy(a,b)

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

•  $\chi^2(a)$  (a integer)

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

• Erlang(a,b) (b integer)

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





Exponential(a)

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

Extreme Value(a,b)

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

• Logistic(a,b)

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



Pareto(a,b)

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

Standard symmetrical triangular distribution

$$f(x) = \begin{cases} 4x & \text{if } 0 \le x \le 1/2\\ 4(1-x) & \text{if } 1/2 \le x \le 1 \end{cases} \qquad R = \frac{r_1 + r_2}{2}$$

Weibull(a,b)

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





## **Appendix**

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

pdf

$$f_X(x) = \begin{cases} 1/(b-a) & \text{if } a \le x \le b, \\ 0 & \text{otherwise.} \end{cases}$$

CDF

$$F_X(x) = \begin{cases} 0 & \text{if } x \le a, \\ (x-a)/(b-a) & \text{if } a \le x \le b, \\ 1 & \text{if } x \ge b. \end{cases}$$

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

