

# Chapter 13

## Generating Random Variables for Simulation

We've seen many examples of distributions.  
How do we generate instances of a given  
distribution?

Two methods:

- ❑ Inverse Transform method
- ❑ Accept-Reject method

Both assume we  
already have  
generator for  $U(0,1)$

# Inverse Transform Method

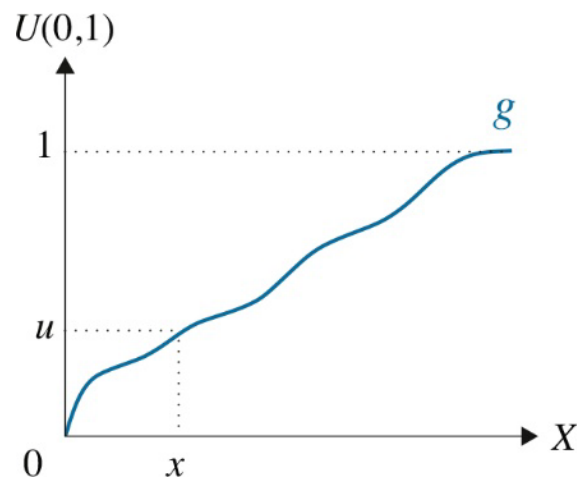
**Requirement:** To generate instances of r.v.  $X$ :

1. Need to know c.d.f. of  $X$ , that is,  $F_X(x) = P\{X \leq x\}$
2. Need to be able to easily invert  $F_X(x)$ , i.e. get  $x$  from  $F_X(x)$

# Inverse Transform Method

Let  $u$  be our random instance of  $U(0,1)$ . Want to map  $u$  to an instance  $x$  of the r.v.  $X$ .

Let  $g^{-1}$  be the mapping that takes  $u$ 's to  $x$ 's:



**Q:** What properties should  $g$  have?

- A value in  $[0, x]$  should be output with probability:  $F_X(x)$
- A value in  $[0, x]$  is output with probability:  $u$

➡  $u = F_X(x)$

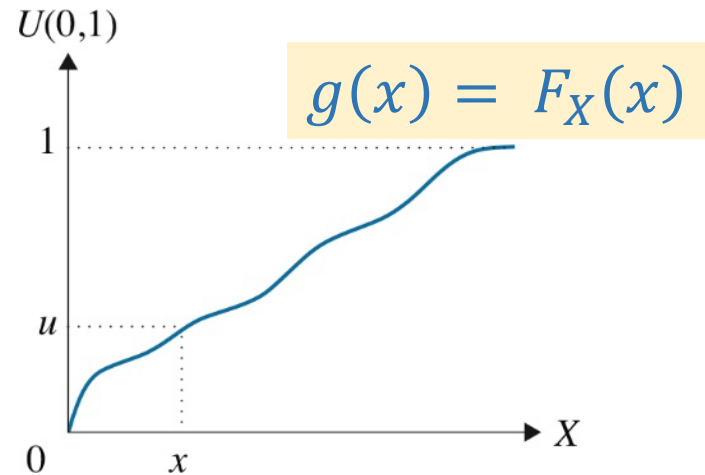
➡  $x = F_X^{-1}(u)$

➡  $g(\cdot) = F_X(\cdot)$

# Inverse Transform Method

Let  $u$  be our random instance of  $U(0,1)$ . Want to map  $u$  to an instance  $x$  of the r.v.  $X$ .

Let  $g^{-1}$  be the mapping that takes  $u$ 's to  $x$ 's:



## Inverse Transform method to generate continuous r.v. $X$ :

1. Generate random  $u \in U(0,1)$ .
2. Return  $x$  such that  $F_X(x) = u$ .

# Inverse Transform Method: Example

**Q:** Generate  $X \sim \text{Exp}(\lambda)$ , given  $u \in U(0,1)$ .

$$F_X(x) = u$$

$$1 - e^{-\lambda x} = u$$

$$e^{-\lambda x} = 1 - u$$

$$-\lambda x = \ln(1 - u)$$

$$\text{Return } x = -\frac{1}{\lambda} \ln(1 - u)$$

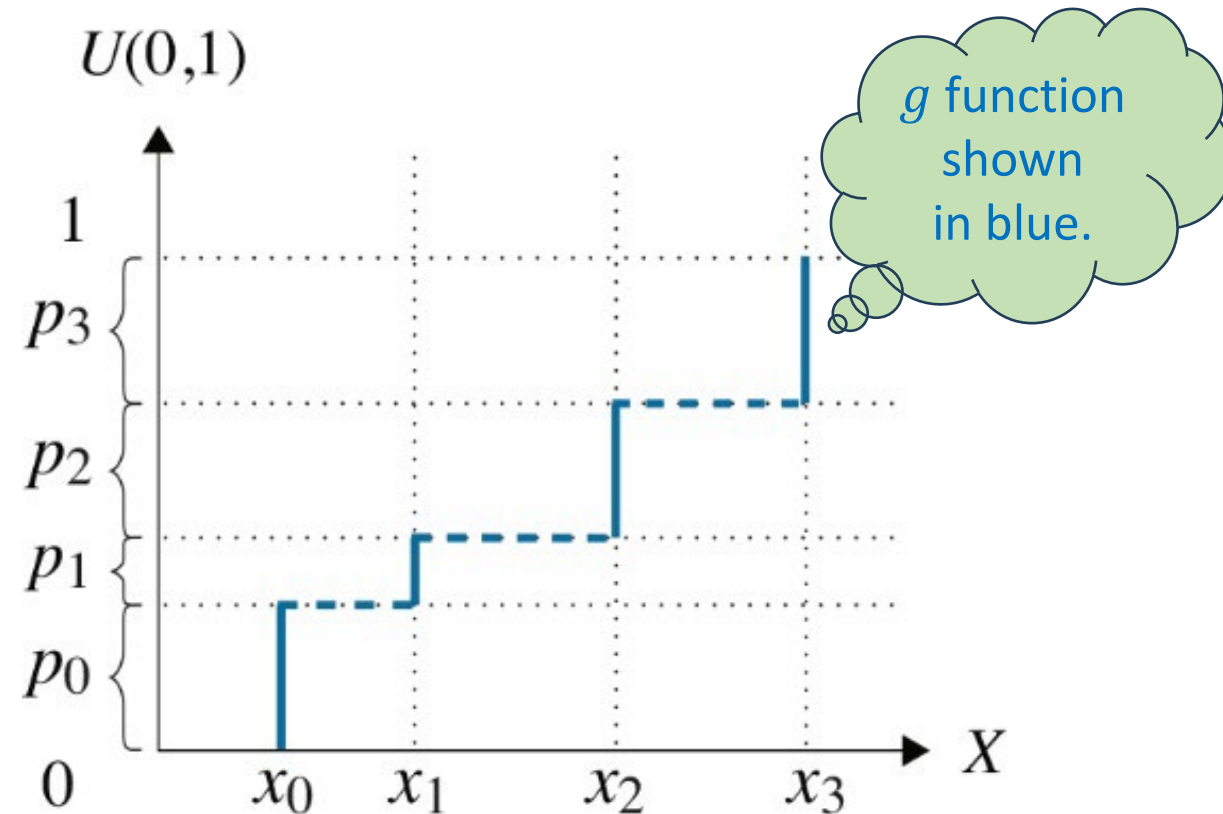
# Inverse Transform Method for Discrete R.V.

Same idea applies to **discrete** r.v.s.

Let  $u$  be our random instance of  $U(0,1)$ . Want to map  $u$  to an instance  $x$  of the r.v.  $X$ .

$$X = \begin{cases} x_0 & \text{w. p. } p_0 \\ x_1 & \text{w. p. } p_1 \\ \vdots & \\ x_k & \text{w. p. } p_k \end{cases}$$

Warning: Need to start by arranging values s.t.  $x_0 < x_1 < x_2 < \dots < x_k$

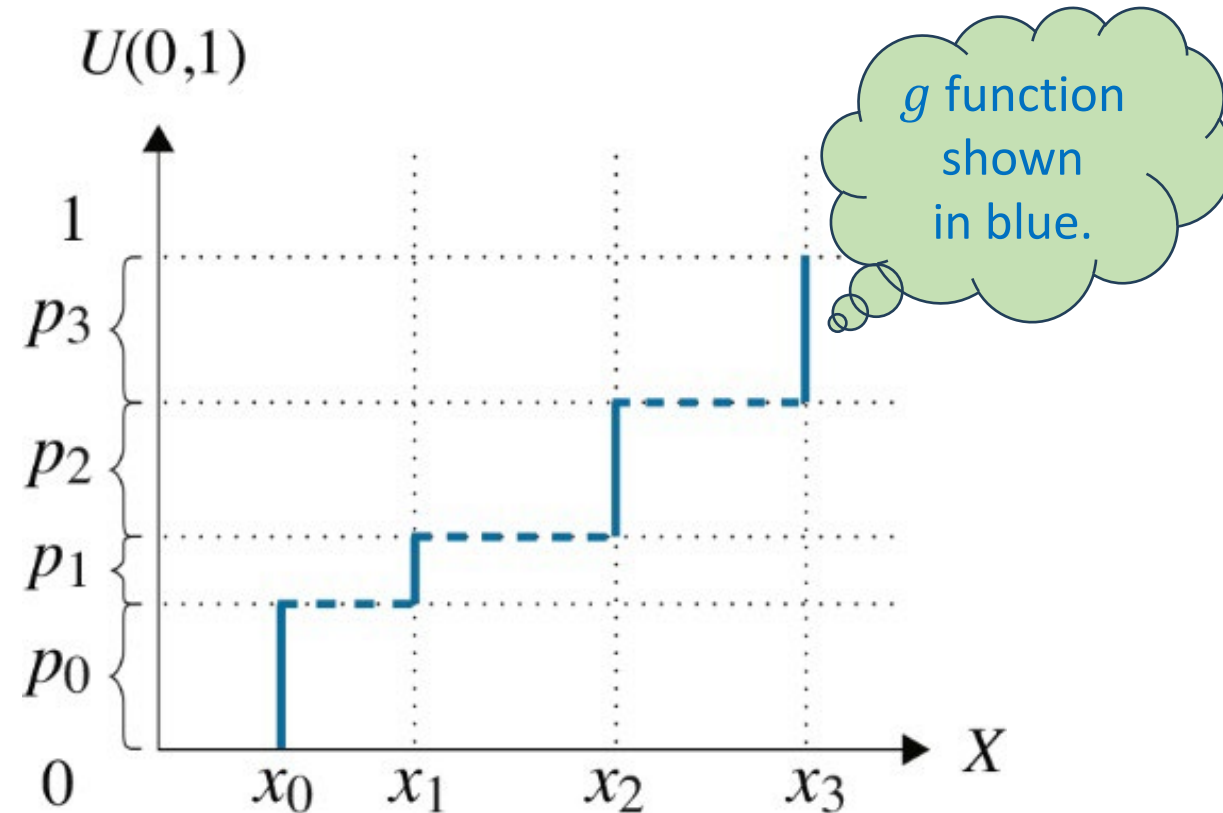


# Inverse Transform Method for Discrete R.V.

## Inverse Transform method to generate discrete r.v. $X$ :

1. Generate random  $u \in U(0,1)$ .
2. If  $0 < u \leq p_0$ , output  $x_0$ .  
If  $p_0 < u \leq p_0 + p_1$ , output  $x_1$ .  
If  $p_0 + p_1 < u \leq p_0 + p_1 + p_2$ , output  $x_2$ .  
If  $\sum_{i=0}^{\ell-1} p_i < u \leq \sum_{i=0}^{\ell} p_i$ , then output  $x_\ell$ ,  
where  $0 \leq \ell \leq k$

To be practical, requires closed-form and invertible expression for  $F_X(x)$ .





# Inverse Transform Method for Discrete R.V.

**Example:** Generate instance  $x$  of  $X$  given instance  $u$  of  $U(0,1)$ .

$$X = \begin{cases} 1 & \text{w.p. } 0.1 \\ 2 & \text{w.p. } 0.1 \\ \vdots & \\ 10 & \text{w.p. } 0.1 \end{cases}$$

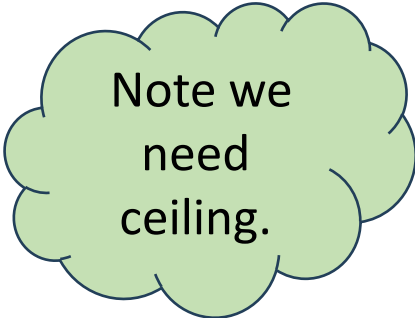
**Solution:** We want to determine  $x$  as a function of  $u$ .

$$p_X(i) = 0.1$$

$$F_X(x) = \mathbf{P}\{X \leq x\} = \sum_{i=1}^x p_X(i) = x \cdot (0.1)$$

$$\text{Set } u = F_X(x) = x \cdot (0.1)$$

$$\text{Hence } x = [10u]$$



Note we need ceiling.

# Inverse Transform Method for Discrete R.V.

## **Example:**

Generate instances of  $X \sim \textit{Geometric}(p)$

## **Solution:**

See Exercise 13.3 in your textbook.

# Interview Question

Generate instance of  $X$  where:

$$X = \begin{cases} 1 & \text{w. p. } p_1 = 10^{-6} \cdot k_1 = 10^{-6} \cdot 0.33 \\ 2 & \text{w. p. } p_2 = 10^{-6} \cdot k_2 = 10^{-6} \cdot 1.63 \\ 3 & \text{w. p. } p_3 = 10^{-6} \cdot k_3 = 10^{-6} \cdot 0.75 \\ & \dots \\ 10^6 & \text{w. p. } p_{10^6} = 10^{-6} \cdot k_{10^6} = 10^{-6} \cdot 1.02 \end{cases}$$

Given:

All of the constants,  
 $k_1, k_2, k_3, \dots, k_{10^6}$   
are between  $(0, 2)$ .

**Question:** Suppose you know all the  $p_i$ 's. How do you generate an instance of  $X$ ?

Inverse Transform:

1. Compute all the partial sums. Store in an array  $A[1]$  to  $A[10^6]$ .
2. Given  $u$ , find *approximate* bin in array:  $x \approx \lceil u \cdot 10^6 \rceil$ ,  
then search around there for partial sum interval that contains  $u$ .

# Interview Question

Generate instance of  $X$  where:

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Given:

All of the constants,  
 $k_1, k_2, k_3, \dots, k_{10^6}$   
are between  $(0, 2)$ .

**Question:** Now suppose you don't know the  $p_i$ 's. Can look up one at a time, but takes time to look up each one, and they change over time.

Can't get partial sums → Can't use Inverse Transform Method  
→ Need new approach!

# Interview Question

Generate instance of  $X$  where:

$$X = \begin{cases} 1 & \text{w. p. } p_1 = 10^{-6} \cdot k_1 = 10^{-6} \cdot 0.33 \\ 2 & \text{w. p. } p_2 = 10^{-6} \cdot k_2 = 10^{-6} \cdot 1.63 \\ 3 & \text{w. p. } p_3 = 10^{-6} \cdot k_3 = 10^{-6} \cdot 0.75 \\ & \dots \\ 10^6 & \text{w. p. } p_{10^6} = 10^{-6} \cdot k_{10^6} = 10^{-6} \cdot 1.02 \end{cases}$$

Given:

All of the constants,  
 $k_1, k_2, k_3, \dots, k_{10^6}$   
are between  $(0, 2)$ .

**Question:** Now suppose you don't know the  $p_i$ 's. Can look up one at a time, but takes time to look up each one, and they change over time.



- Imagine r.v.  $Y$ , where  $p_i = 10^{-6}$ ,  $\forall i$
- Generate instance  $i$  of  $Y$  (easy to do).
- Now look up  $p_X(i)$ . Return  $i$  as instance of  $X$  if  $p_X(i)$  is high compared to  $p_Y(i)$ ?

Intuition  
behind  
Accept/Reject  
method

# A method with fewer requirements ...

**Goal:** To generate instances of r.v.  $X$

## Inverse Transform Method

1. Need to know closed-form  $F_X(x)$
2. Need to know be able to easily invert  $F_X(x)$ , i.e. get  $x$  from  $F_X(x) = u$

## Accept-Reject Method

1. Need the p.m.f. of  $X$  (or p.d.f., if continuous)
2. Need to know how to generate some other r.v.  $Y$ , where  $X$  and  $Y$  take on the same set of values, i.e.,

$$p_X(i) > 0 \iff p_Y(i) > 0$$

# A method with fewer requirements ...

**Goal:** To generate instances of r.v.  $X$



## High-level idea:

We generate an instance of  $Y$ .  
Then with some probability we return that value as our instance of  $X$  (accept).  
Otherwise, we reject that value and try again.

## Accept-Reject Method

1. Need the p.m.f. of  $X$  (or p.d.f., if continuous)
2. Need to know how to generate some other r.v.  $Y$ , where  $X$  and  $Y$  take on the same set of values, i.e.,

$$p_X(i) > 0 \iff p_Y(i) > 0$$

# Accept-Reject Method

## Accept-Reject method to generate discrete r.v. $X$ :

1. Find a discrete r.v.  $Y$  which we already know how to generate, where

$$p_X(i) > 0 \iff p_Y(i) > 0$$

2. Let  $c > 1$  be the smallest constant s.t.

$$\frac{p_X(i)}{p_Y(i)} \leq c, \quad \forall i \text{ s.t. } p_X(i) > 0$$

3. Generate an instance of  $Y$ . Call this instance  $i$ .

4. With probability  $\text{AcceptRatio}(i) = \frac{p_X(i)}{cp_Y(i)}$  accept  $i$ .  
Else, reject  $i$  and return to step 3.

Relative likelihood  
of  $i$  being an instance  
of  $X$  versus an instance  
of  $Y$ .

The  $c$  ensures  
 $\text{AcceptRatio} < 1$ .



# Accept-Reject Method: Correctness

**Claim:**  $P\{X \text{ is set to } i\} = p_X(i)$

**Proof:** Frac of time  $i$  is generated & accepted =  $P\{i \text{ is generated}\} \cdot P\{i \text{ is accepted} \mid i \text{ is generated}\}$

$$= p_Y(i) \cdot \frac{p_X(i)}{c p_Y(i)} = \boxed{\frac{p_X(i)}{c}}$$

Frac of time any value is accepted =  $\sum_i$  Frac of time  $i$  is generated & accepted

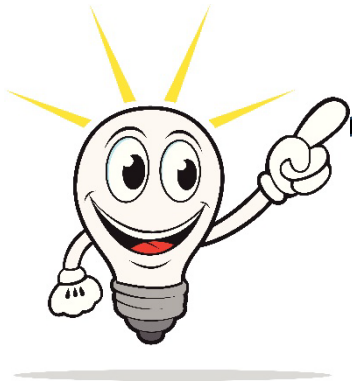
$$= \sum_i \frac{p_X(i)}{c} = \boxed{\frac{1}{c}}$$

$$P\{X \text{ is set to } i\} = \frac{\text{Frac of time } i \text{ is generated and accepted}}{\text{Frac of time any value is accepted}} = \frac{\frac{p_X(i)}{c}}{\frac{1}{c}} = \boxed{p_X(i)}$$

# Accept-Reject Method

**Theorem:** On average, only need to generate  $c$  values of  $Y$  before one is accepted.

**Proof:**



This is easy. Use what we just proved!

$$\text{Frac of time any value is accepted} = \boxed{\frac{1}{c}}$$

$$\Rightarrow E[\# \text{ values generated until one is accepted}] = c$$

# Back to the Interview Question

Generate instance of  $X$  where:

$$X = \begin{cases} 1 & \text{w.p. } p_1 = 10^{-6} \cdot k_1 = 10^{-6} \cdot 0.33 \\ 2 & \text{w.p. } p_2 = 10^{-6} \cdot k_2 = 10^{-6} \cdot 1.63 \\ 3 & \text{w.p. } p_3 = 10^{-6} \cdot k_3 = 10^{-6} \cdot 0.75 \\ & \dots \\ 10^6 & \text{w.p. } p_{10^6} = 10^{-6} \cdot k_{10^6} = 10^{-6} \cdot 1.02 \end{cases}$$

Given: All the constants,  $k_1, k_2, k_3, \dots, k_{10^6}$  are between  $(0, 2)$ .

**Question:** You don't know the  $p_i$ 's. Can look up one at a time, but takes time to look up each one, and they change over time.

**Solution:** Accept-Reject!

1. Let  $Y = [1, 2, 3, \dots, 10^6]$ , each with probability  $10^{-6}$
2. Observe  $\frac{p_X(i)}{p_Y(i)} < c, \forall i$
3. Generate  $i$ : instance of  $Y$ . Look up  $p_X(i)$  only.
4. With probability  $\frac{p_X(i)}{c \cdot p_Y(i)}$ : ACCEPT  $i$ , RETURN  $X = i$ .  
Else, Return to Step 3.

What is  $c$ ?

$c = 2$

How many lookups are required in expectation?

2

# Accept-Reject Method: Continuous

## Accept-Reject method to generate continuous r.v. $X$ with p.d.f. $f_X(t)$ :

1. Find a discrete r.v.  $Y$  which we already know how to generate, where

$$f_X(t) > 0 \iff f_Y(t) > 0$$

2. Let  $c > 1$  be the smallest constant s.t.

$$\frac{f_X(t)}{f_Y(t)} \leq c, \quad \forall t \text{ s.t. } f_X(t) > 0.$$

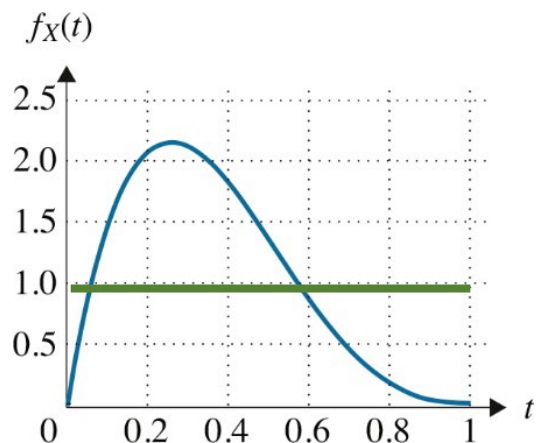
3. Generate an instance  $t$  of  $Y$ .

4. With probability  $\text{AcceptRatio}(t) = \frac{f_X(t)}{cf_Y(t)}$ , accept  $t$  and return  $X = t$ .

Else, reject  $t$  and return to step 3.

# Accept-Reject Method: Example 1

**GOAL:** Generate r.v.  $X$  where  $f_X(t) = 20t(1-t)^3$ ,  $0 < t < 1$



**Q:** Given this image of  $f_X(t)$ , what's a good choice for  $Y$  ?

**A:**  $f_Y(t) = 1$ ,  $0 < t < 1$  has same range and is easy to generate.

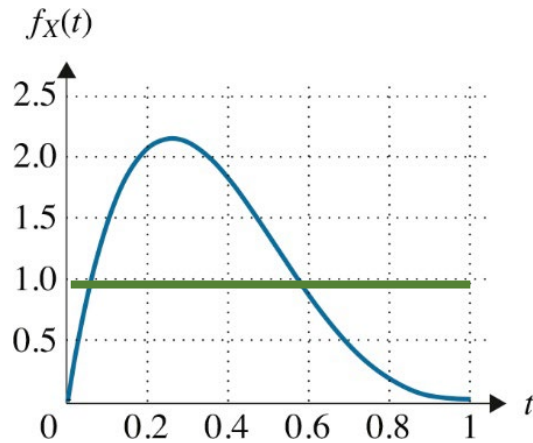
**Q:** Looking at the picture, how high is  $c$  approximately ?

**A:**  $c \approx 2$ .

So only need 2 guesses to attempts on average to generate  $X$ .

# Accept-Reject Method: Example 1

**GOAL:** Generate r.v.  $X$  where  $f_X(t) = 20t(1 - t)^3$ ,  $0 < t < 1$



More precisely:

$$c = \max_t \left\{ \frac{f_X(t)}{f_Y(t)} \right\} = \max_t 20t(1 - t)^3$$

$$\frac{d}{dt} (20t(1 - t)^3) = 0 \quad \Leftrightarrow \quad t = \frac{1}{4}$$

$$c = \frac{f_X\left(\frac{1}{4}\right)}{f_Y\left(\frac{1}{4}\right)} = 20 \left(\frac{1}{4}\right) \left(\frac{3}{4}\right)^3 = \frac{135}{64}$$

# Accept-Reject Method: Example 2

**GOAL:** Generate r.v.  $N$  where  $N \sim Normal(0,1)$



Suffices to generate  $X = |N|$  and then multiply  $X$  by  $-1$  with probability  $\frac{1}{2}$

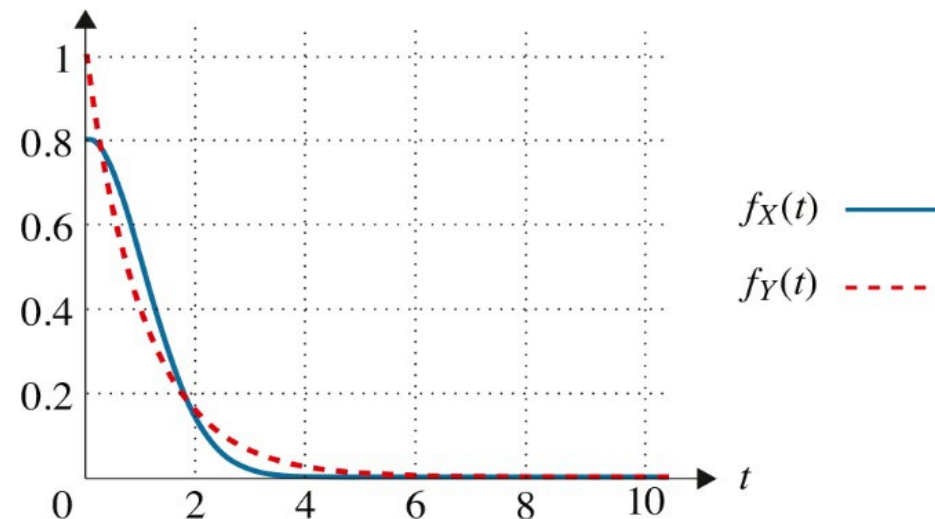
**Q:** What's a good choice for r.v.  $Y$  with range  $0$  to  $\infty$ ?

**A:** Let  $Y \sim Exp(1)$ .  $f_Y(t) = e^{-t}$ ,  $0 < t < \infty$

**Q:** From the picture, how high is  $c$  approximately?

**A:**  $c \approx 1.3$ .

So only need 1.3 guesses to attempts on average to generate  $X$ .



# Accept-Reject Method: Example 2

More precisely:

$$c = \max_t \left\{ \frac{f_X(t)}{f_Y(t)} \right\} = \max_t \sqrt{\frac{2}{\pi}} e^{t - \frac{t^2}{2}}$$

Suffices to maximize  $t - \frac{t^2}{2}$

$$0 = \frac{d}{dt} \left( t - \frac{t^2}{2} \right) = 1 - t \Leftrightarrow t = 1$$

$$c = \frac{f_X(1)}{f_Y(1)} = \sqrt{\frac{2e}{\pi}} \approx 1.3$$

