

$$X \rightarrow \begin{cases} \text{Pr}[X=1] = 0.8 \\ \text{Pr}[X=0] = 0.2 \end{cases}$$

Mean :  $\mu = 0.8 \times 1 + 0.2 \times 0$   
 $= 0.8$

$$\begin{array}{r} 0.64 \\ 0.2 \\ \hline 1.28 \end{array} \quad \begin{array}{r} 0.128 \\ 0.04 \\ \hline 0.168 \end{array}$$

$$\sigma^2 = ? \quad \sigma^2 = \sum (x - \mu)^2 \cdot f(x)$$

$$= (1 - 0.8)^2 \cdot 0.8 + (0 - 0.8)^2 \cdot 0.2$$

$$= (0.2)^2 \times 0.8 + 0.64 \times 0.2$$

$$= 0.04 \times 0.8 + 0.128$$

$$\sigma^2 = \sum x^2 f(x) - \mu^2 = 0.16$$

$$= 1 \cdot 0.8 + 0 \cdot 0.2$$

$$= 0.8 - 0.64$$

$$= 0.8 - 0.64 = 0.16$$

$$\sigma = \sqrt{0.16} = 0.4$$

$$\sigma = 0.4$$

$$\sigma^2 = \sum (x - \mu)^2 f(x) = \boxed{\sum x^2 f(x) - \mu^2}$$

$$\begin{aligned} \sigma^2 &= \sum_x (x - \mu)^2 f(x) \\ \sigma^2 &= \sum_x x^2 f(x) - \mu^2 \end{aligned}$$

Proof:  $\sigma^2 = \sum_x (x - \mu)^2 f(x)$

$$\begin{aligned} &= \sum_x (x^2 - 2\mu x + \mu^2) f(x) \\ &= \sum_x [x^2 f(x) - 2\mu x f(x) + \mu^2 f(x)] \end{aligned}$$

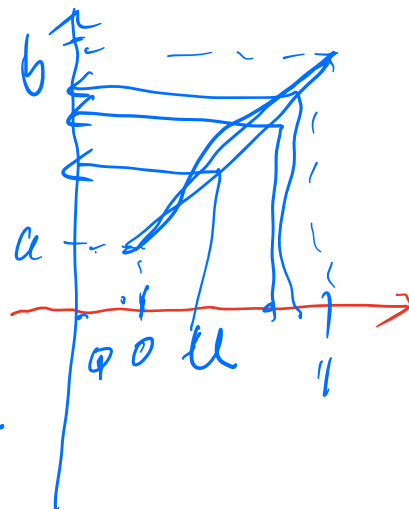
$$= \sum_x x^2 f(x) - 2\mu \left( \sum_x x f(x) \right) + \sum_x \mu^2 f(x)$$

$$= \sum_x x^2 f(x) - 2 \cdot \mu \cdot \mu + \mu^2 \left( \sum_x f(x) \right)$$

$$= \sum x^2 f(x) - 2\mu^2 + \mu^2$$

$$\sigma^2 = \sum x^2 f(x) - \mu^2$$

$$u \in (0, 1) \\ \downarrow \quad \downarrow \quad \downarrow \\ x \in (a, u, b)$$



$$x = a + (b - a) \cdot u$$

$$u = 0 \Rightarrow x = a$$

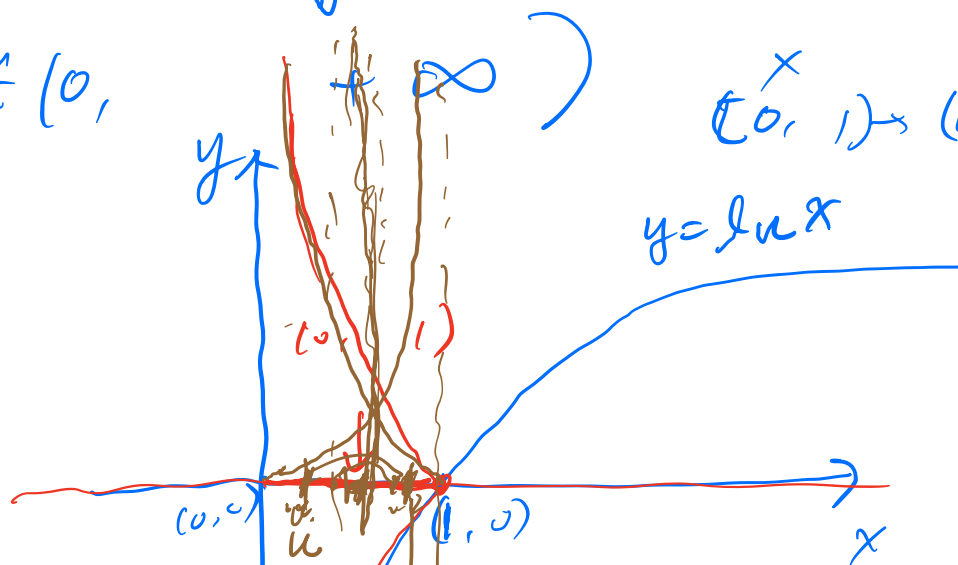
$$u = 1 \Rightarrow x = b$$

$$u \in (0, 1) \rightarrow ?$$

$$x \in (0, \infty)$$

$$y = -\ln x$$

$$(0, 1) \rightarrow (0, +\infty)$$

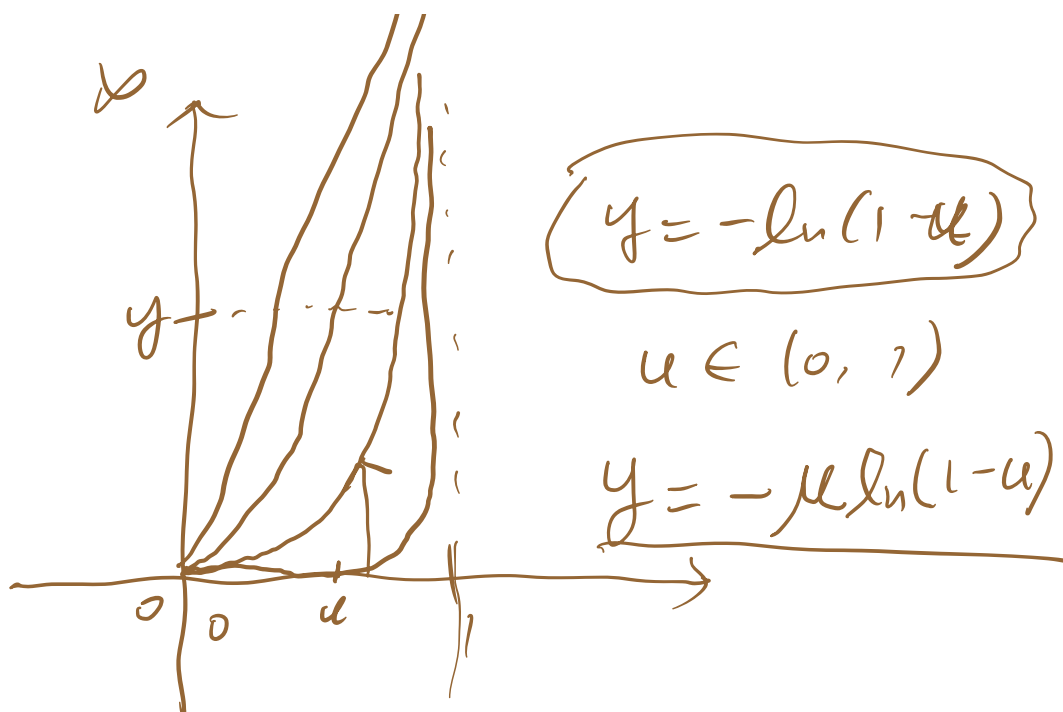


$$u = 0.2 \\ \downarrow \quad \textcircled{0.8}$$

$$u = 0.3 \\ \downarrow \quad 0.7$$

$$y = -\ln x$$

$$y = -\ln(1 - u)$$



# Exponential Random Variates

The transformation  
is monotone  
increasing, one-to-  
one, and onto

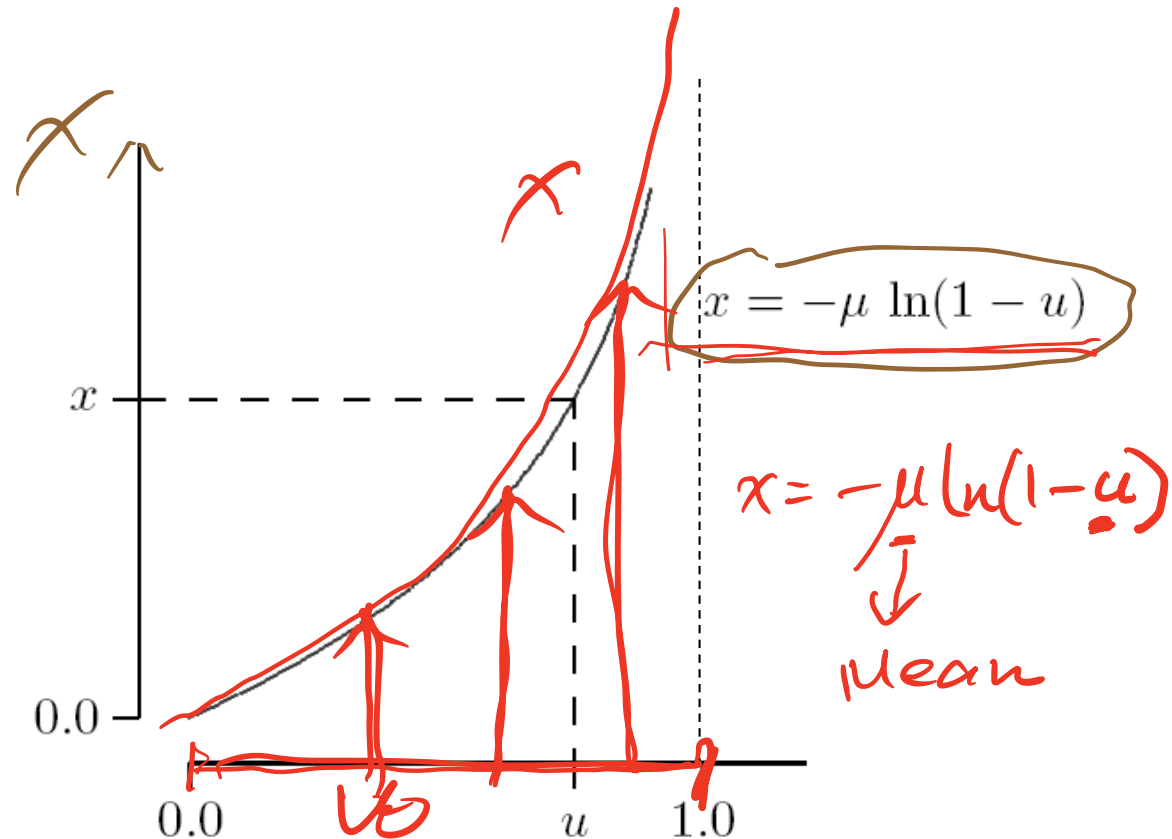
$$0 < u < 1$$

$$\Leftrightarrow 0 < (1 - u) < 1$$

$$\Leftrightarrow -\infty < \ln(1 - u) < 0$$

$$\Leftrightarrow 0 < -\mu \ln(1 - u) < \infty$$

$$\Leftrightarrow 0 < x < \infty$$



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

$$u \in (0, 1)$$

# Exponential Random Variate

- Parameter  $\mu$ :  
 $x = -\mu \ln(1 - u)$   
 $\bar{x} = \mu$  (sample mean)  
 $\sigma^2 = \mu^2$  (variance)  
 $\sigma = \mu$  (standard deviation)

- Continuous uniform - Parameters  $a$  and  $b$ :

$$x = a + (b - a) * \text{Random}();$$

$$\bar{x} = \frac{a + b}{2}$$

$$\sigma^2 = \frac{(b - a)^2}{12}$$

$$x = a + (b - a) \cdot u$$

$\downarrow$   
 $(0, 1)$

# Generating an Exponential Random Variate

```
double Exponential(double  $\mu$ )    /* use  $\mu > 0.0$ 
*/
{
     $-\mu$ 
    return ( $-\mu$  * log(1.0 -  $u$  Random()));
}
```

- The parameter  $\mu$  specifies the sample mean.
- In the single-server service node simulation, use Exponential( $\mu$ ) interarrival times:

$$a_i = a_{i-1} + \text{Exponential}(\mu); i = 1, 2, 3, \dots, n.$$

# Exponential Interarrival Time

```
double Exponential(double m)           // generate an Exponential random variate, m > 0.0
{
    return (-m * log(1.0 - Random()));
}
```

```
double Uniform(double a, double b) // generate a Uniform random variate, a < b
{
    return (a + (b - a) * Random());
}
```

```
double GetArrival(void)                // generate the next arrival time
{
    static double arrival = START;
    arrival += Exponential(2.0);
    return (arrival);
}
```

*arrival = arrival + exponential(2.0)*

```
double GetService(void)                // generate the next service time
{
    return (Uniform(1.0, 2.0));
}
```



# Geometric Random Variates

- The Geometric( $p$ ) random variate is the discrete analog to a continuous Exponential( $\mu$ ) random variate

Let  $x = \text{Exponential}(\mu) = -\mu \ln(1 - u)$

$y = \lfloor x \rfloor$  and let  $p = \Pr(y \neq 0)$ ,

It can be proved that

$$p = \Pr(y \neq 0) = \exp(-1/\mu)$$

Use  $p$  as a parameter for the Geometric Random Variate,

$$y = \lfloor \ln(1 - u) / \ln(p) \rfloor$$

where  $p = \Pr(y \neq 0)$ .

$\mu$ : mean  $\rightarrow$  return  $\log(-\mu \ln(1-u))$   
 $x = -\underline{\mu} \ln(1-u)$

$$y = \lfloor x \rfloor = \lfloor -\mu \ln(1-u) \rfloor$$

$$\underline{x \in (0, +\infty)}$$

$$\hookrightarrow \underline{y \in \{0, 1, 2, 3, 4, \dots\}}$$

$$\lfloor 0.5 \rfloor = 0$$

$$\lfloor 5.756 \rfloor = 5$$

$$\lfloor 101.00001 \rfloor = 101$$

$$\lfloor 99.0000 \rfloor = 99$$

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

$$y = \lfloor x \rfloor \quad x \in (0, +\infty)$$

$$\Pr(y \neq 0) \quad \boxed{x \geq 1}$$

For what value of

$$\underline{x} \quad y \neq 0? \quad x = 0.6$$

$$y = \underline{\lfloor x \rfloor} \quad \lfloor x \rfloor = \lfloor 0.6 \rfloor$$

$$y \neq 0 \quad = 0$$

$$\Pr[\underline{y \neq 0}] = \Pr[x \geq 1]$$

$$= \Pr[\underline{-\mu \ln(1-u) \geq 1}]$$

$$= \Pr[\underline{u \geq 1 - e^{-\frac{1}{\mu}}}]$$

$$\frac{-\mu \ln(1-u) \geq 1}{e^{\ln(1-u)} \leq e^{-\frac{1}{\mu}}}$$

$$e^{\ln(1-u)} \leq e^{-\frac{1}{\mu}}$$

$$1-u \leq e^{-\frac{1}{\mu}}$$

$$u \geq 1 - e^{-\frac{1}{\mu}}$$

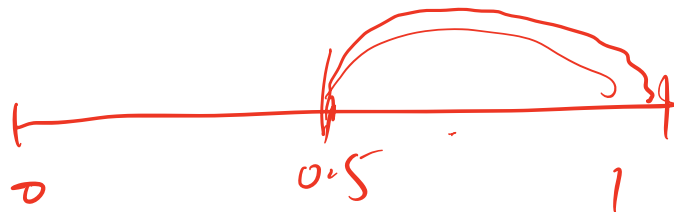
$$\Pr[u \geq \underbrace{1 - e^{-\frac{1}{\mu}}}_r]$$

$$\underbrace{e^{-\frac{1}{\mu}}}_{<1} = \frac{1}{\underbrace{e^{\frac{1}{\mu}}}_{>1}}$$

$$0 < r < 1$$

$$\Pr[u \geq r] \propto r < 1$$

Assume that  $r = 0.5$



$$\Pr[u \geq r] = 1 - r$$

$$1 - [1 - e^{-\frac{1}{\mu}}] = e^{-\frac{1}{\mu}}$$

$$P = \Pr[y \neq 0] = e^{-\frac{1}{\mu}}$$

$$P = e^{-\frac{1}{\mu}}$$

$$\ln P = \ln e^{-\frac{1}{\mu}} = -\frac{1}{\mu} \frac{\ln e}{1}$$

$$\ln p = -\frac{1}{\mu}$$

$$\mu = \left( -\frac{1}{\ln p} \right)$$

$$y = \lfloor x \rfloor = \left\lfloor -\mu \ln(1-a) \right\rfloor$$

$$= \left\lfloor \frac{\ln(1-a)}{\ln p} \right\rfloor$$

$$P = \Pr[y \neq 0]$$

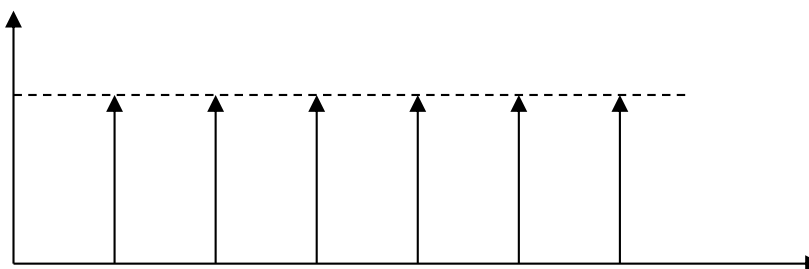
# Discrete Uniform

- Parameters  $a$  and  $b$

$$P(n) = \frac{1}{b - a + 1} \quad \text{for } n = a, a + 1, \dots, b - 1, b$$

$$\bar{x} = \frac{a + b}{2} \quad (\text{mean})$$

$$\sigma^2 = \frac{(b - a)(b - a - 1)}{12} \quad (\text{variance})$$



# Geometric Random Variates

- With parameter  $p$ :  $y = \lfloor \ln(1 - u) / \ln(p) \rfloor$

$$P(n) = P^n (1 - P), \quad n = 0, 1, 2, \dots$$

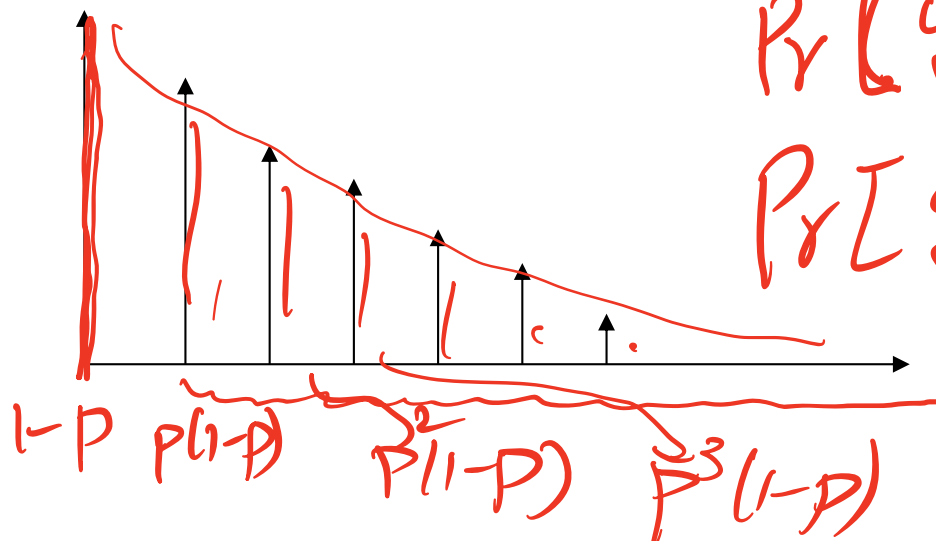
$$\Pr[y = n]$$

$\bar{x}$

$$\bar{x} = \frac{p}{1 - p} \quad (\text{mean})$$

$$\sigma^2 = \frac{p}{(1 - p)^2} \quad (\text{standard deviation})$$

pdf



$$\Pr[y = 0] = 1 - p$$

$$\Pr[y \neq 0] = p$$



# Generating a Geometric Random Variate

```
long Geometric(double p) /* use  $0.0 < p < 1.0$  */  
{  
    return ((long) (log(1.0 - Random()) / log(p)));  
}
```

*Handwritten notes:*  
- A red arrow points from the parameter  $p$  in the function signature to  $Pr(Y \neq 0)$ .  
- The expression  $Random()$  is circled in red, with  $U \sim U(0, 1)$  written below it.

- The mean of a Geometric( $p$ ) random variate is  $p/(1 - p)$ .
- If  $p$  is close to zero then the mean will be close to zero.
- If  $p$  is close to one, then the mean will be large.

$$\frac{p}{1-p}$$

# Example of Service Times

- Assume that jobs arrive at random with a steady-state arrival rate of 0.5 jobs per minute.
- Assume that Job service times are composite with two components:

- The number of service tasks is

$$1 + \text{Geometric}(0.9)$$

$$0.9$$

- The time (in minutes) per task is

$$\text{Uniform}(0.1, 0.2)$$

$$Pr(y \neq 0) = 0.9$$

# Get Service Method

```
double GetService(void)  
{
```

```
    long k;
```

```
    double sum = 0.0;
```

```
    long tasks = 1 + Geometric(0.9);
```

```
    for (k = 0; k < tasks; k++)
```

```
        sum += Uniform(0.1, 0.2);
```

```
    return (sum);
```

```
}
```

10  
0, 1, 2, 3, ...