# CSCI 202 Research Methods





### A static mode

#### **Predicting profits for furniture sales**

Simulation Model for Special Promotion Furniture Sale
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Simulation Model for Special Promotion Furniture Sale		re Sale		fixed by contract
				input data
Stock ordered (S):	3000			calculated data
Unit cost for stock (C):	\$175.00			
			Distribution	Parameters
			Lower	Upper
Demand within first 8 weeks (V):	2667		500	3500
Sales within first 8 weeks (V):	2667			
Initial price (R):	\$251		200	300
Sales after first 8 weeks (S-V):	333			
Discount (D):	0.2			
Sale price (R*D):	0.5			
Profit (P):	\$144,343			
Note: Google Sheets refresh or	browser reload command			

#### Time has no bearing on this model



#### A static model Predicting profits for furniture sales

## This one is on Moodle: Spreadsheet simulation

#### What is ....

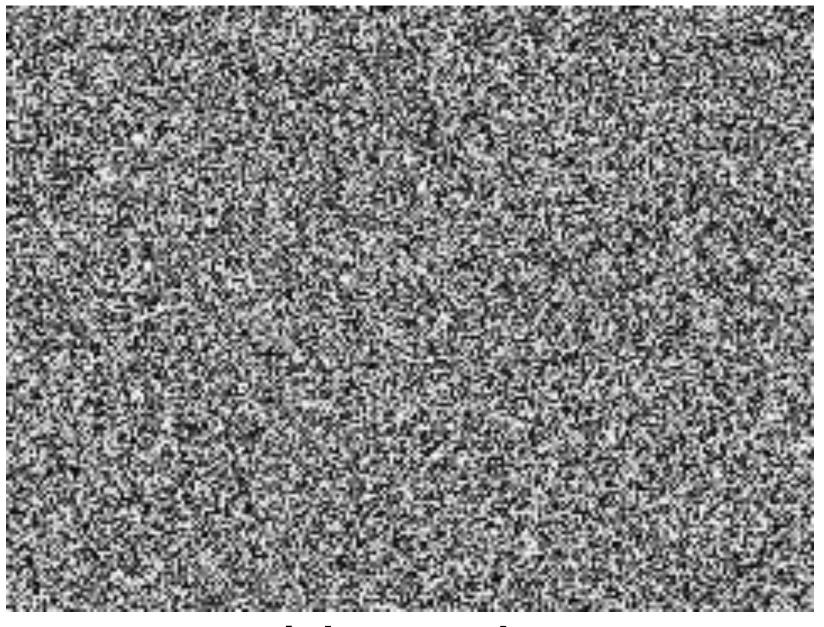
- a random number generator?
- a pseudo-random number generator?

## Why should we need randomness?

#### enerator? Imber generator?

#### "Real" random numbers

source of entropy, you can "mine" random numbers from it.





## There is entropy in nature. If you can identify a

#### white noise

#### **Computers are deterministic...**

If computers are fully deterministic, you need to do some work to get them to give you random numbers...

In Linux, look to /dev/random for random numbers. (You cannot "control" them.)



COMPUTING PRACTICES

Edgar H. Sibley Panel Editor

Practical and theoretical issues are presented concerning the design, implementation, and use of a good, minimal standard random number generator that will port to virtually all systems.

#### **RANDOM NUMBER GENERATORS: GOOD ONES ARE HARD TO FIND**

STEPHEN K. PARK AND KEITH W. MILLER

An important utility that digital computer systems should provide is the ability to generate random numbers. Certainly this is true in scientific computing where many years of experience has demonstrated the importance of access to a good random number generator. And in a wider sense, largely due to the encyclopedic efforts of Donald Knuth [18], there is now a realization that random number generation is a concept of fundamental importance in many different areas of computer science. Despite that, the widespread adoption of good, portable, industry standard software for ran-

goal. Many generators have been written, most of them have demonstrably non-random characteristics, and some are embarrassingly had In fact the current state

siderations developed over a period of several years while teaching a graduate level course in simulation. Students taking this course work on a variety of systems and their choices typically run the gamut from personal computers to mainframes. With Knuth's advice in mind, one important objective of this course is for all students to write and use implementations of a good, minimal standard random number generator that will port to all systems. For reasons discussed later, this minimal standard is a multiplicative linear congruential generator [18, p. 10] with multiplier 16807 and

random number generator (or any other for that matter) to a wide variety of systems is not as easy as it may soon. The issues involved are discussed later in this

#### Pseudo-random number generators (PRNGs)

#### WHAT MAKES A PRNG "GOOD"?





### **Pseudo-random number generator**

#### $Z_i = (aZ_{i-1} + c) \% m$

k = period

 $Z_0 = \text{seed}$ 

LINEAR CONGRUENTIAL GENERATOR

#### a, c, and m = carefully chosen constants

 $\{Z_{0}, Z_{1}, Z_{2}, ..., Z_{k}, Z_{0}, Z_{1}, Z_{2}, ..., Z_{k}, Z_{0}, Z_{1}, Z_{2}, ..., Z_{k,...}\}$ 

### **Pseudo-random number generator**

 $Z_i = (aZ_{i-1} + c) \% m$ 

 $Z_0 = \text{seed}$ 

SAME SEED, SAME SEQUENCE

a, c, and m = carefully chosen constants

 $\{Z_0, Z_1, Z_2, ..., Z_k, Z_0, Z_1, Z_2, ..., Z_k, Z_0, Z_1, Z_2, ..., Z_{k,...}\}$ k = period

LINEAR CONGRUENTIAL GENERATOR

MER GENERATOR

### **Pseudo-random number generator**

#### $Z_i = (aZ_{i-1} + c) \% m$

k = period

 $Z_0 = \text{seed}$ 

LINEAR CONGRUENTIAL GENERATOR

MER GENERATOR

THIS IS VERY BASIC. YOU CAN FIND MUCH BETTER PRNGS OUT THERE.

#### a, c, and m = carefully chosen constants

 $\{Z_{0}, Z_{1}, Z_{2}, ..., Z_{k}, Z_{0}, Z_{1}, Z_{2}, ..., Z_{k}, Z_{0}, Z_{1}, Z_{2}, ..., Z_{k,...}\}$ 



### What is a random variate?

- Bernoulli (discrete)
- •Binomial (discrete)
- •Geometric (discrete)
- Equilikely (discrete)
- Uniform (continuous)
- •Normal or Gaussian (continuous)
- Exponential (continuous)

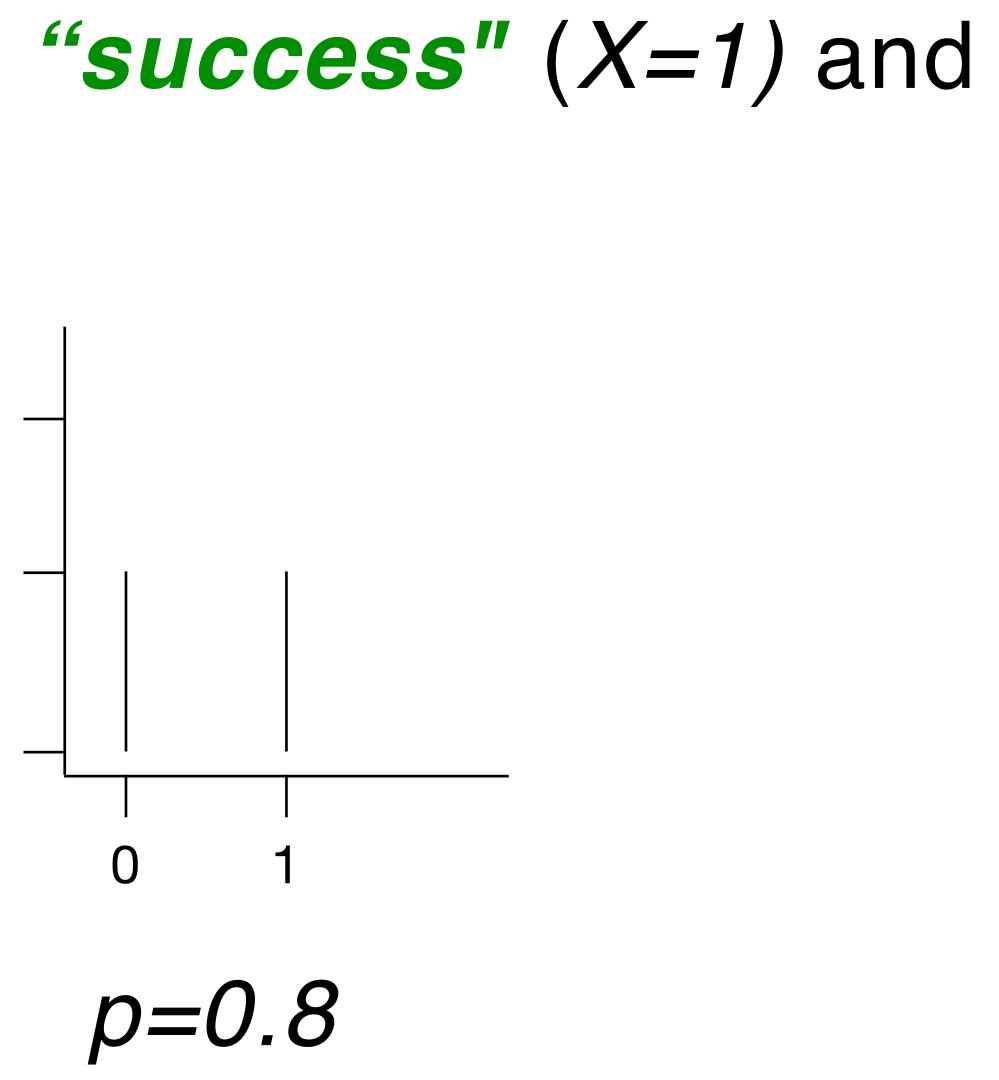


#### **Bernoulli(p)**

# Two possible outcomes: "*success*" (*X*=1) and *failure* (*X*=0).

#### $Pr{X=1} = X$ 1.0 - $Pr{X=0} = 1-X$ 0.5 -

Range = [0, 1] 0.0

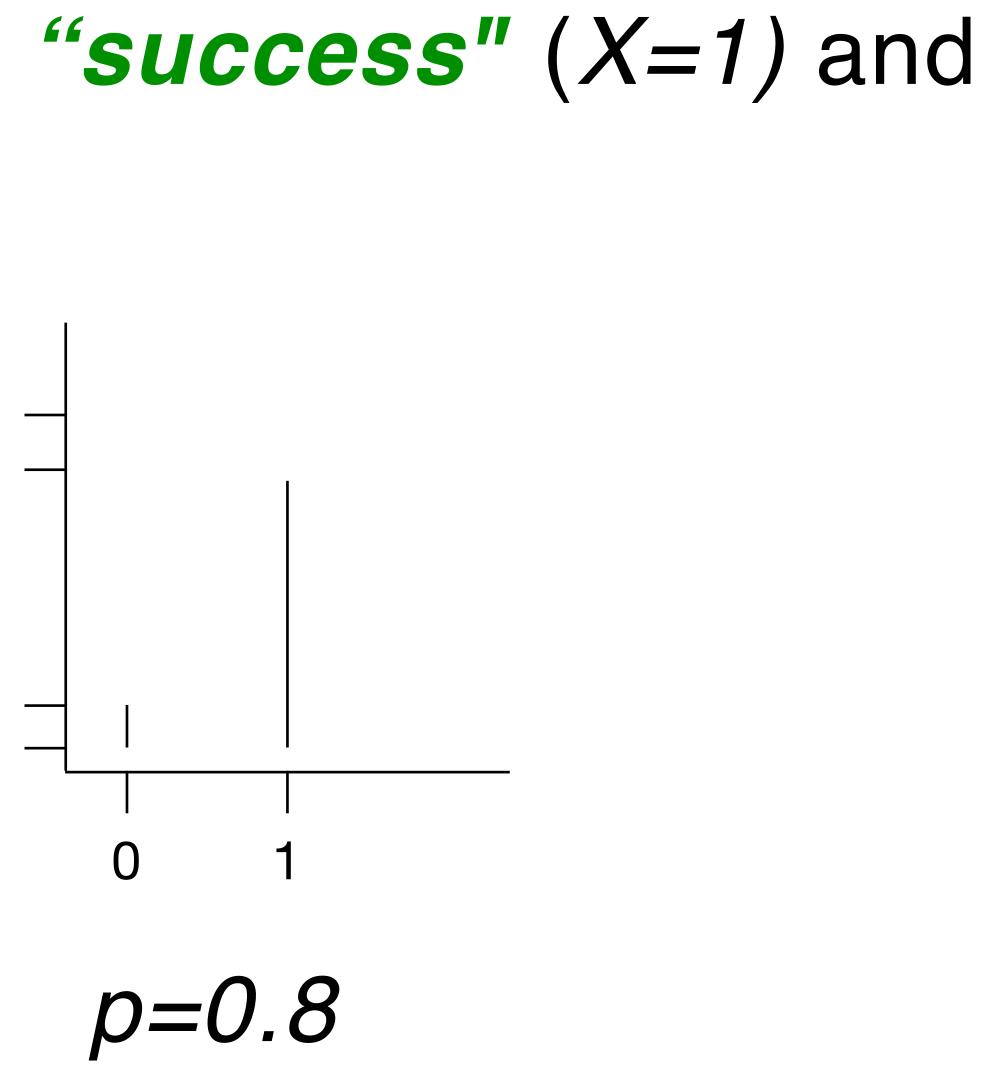


#### **Bernoulli(p)**

# Two possible outcomes: "*success*" (*X*=1) and *failure* (*X*=0).

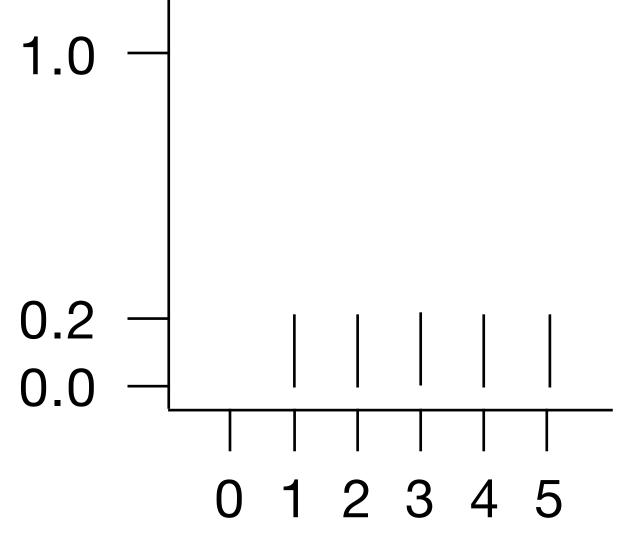
 $Pr{X=1} = x$ 1.0 -0.8 - $Pr{X=0} = 1-x$ 

0.2 Range = [0, 1]

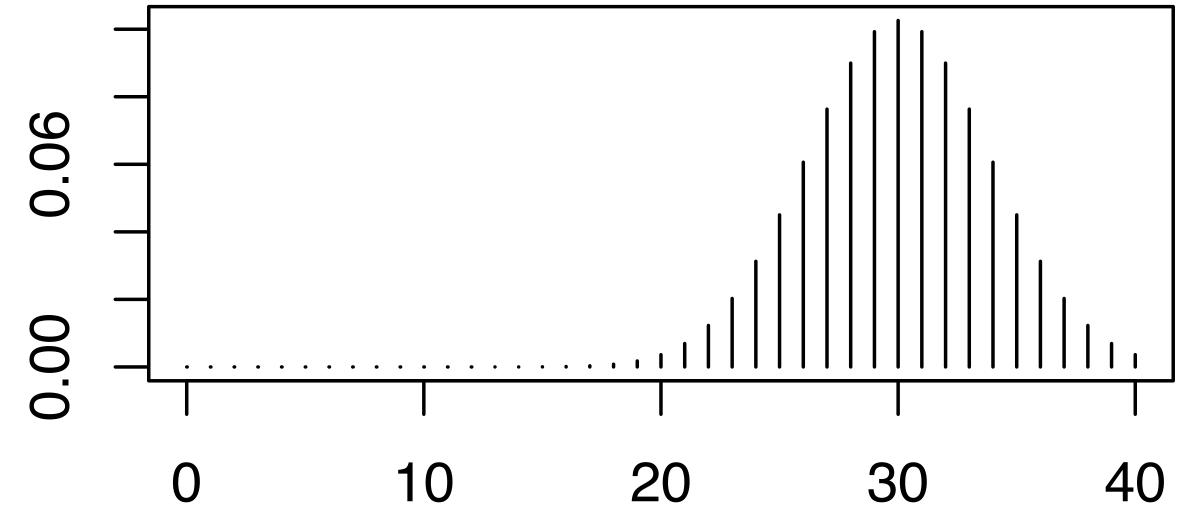


## Equilikely(a,b) Possible values are $\{a, a+1, a+2, ..., b\}$ Range = [a,b]

 $Pr{X=i} = ?$ 



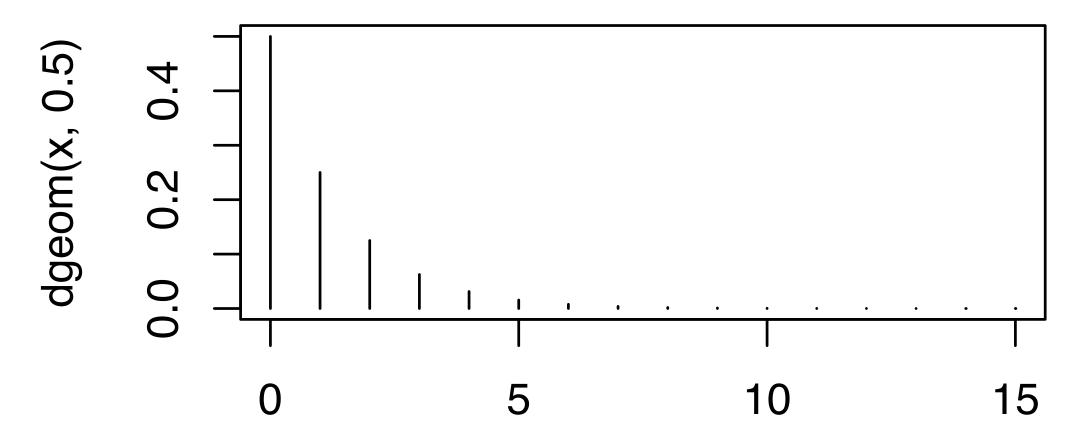
Binomial(n, p) Repeat a Bernoulli(p) experiment n times and count the number of successes. What is the range of *Binomial(n,p)*?  $Pr{X=x} = ?$ 1/2) dbinom(x, 60, 00. SSS....SFF.... F 00.



### **Geometric(p)**

Repeat a Bernoulli(p) experiment until you have a first successes; count the number of failures before you see that success. What is the range of *Geometric(p)*?

$$Pr{X=x} = ?$$



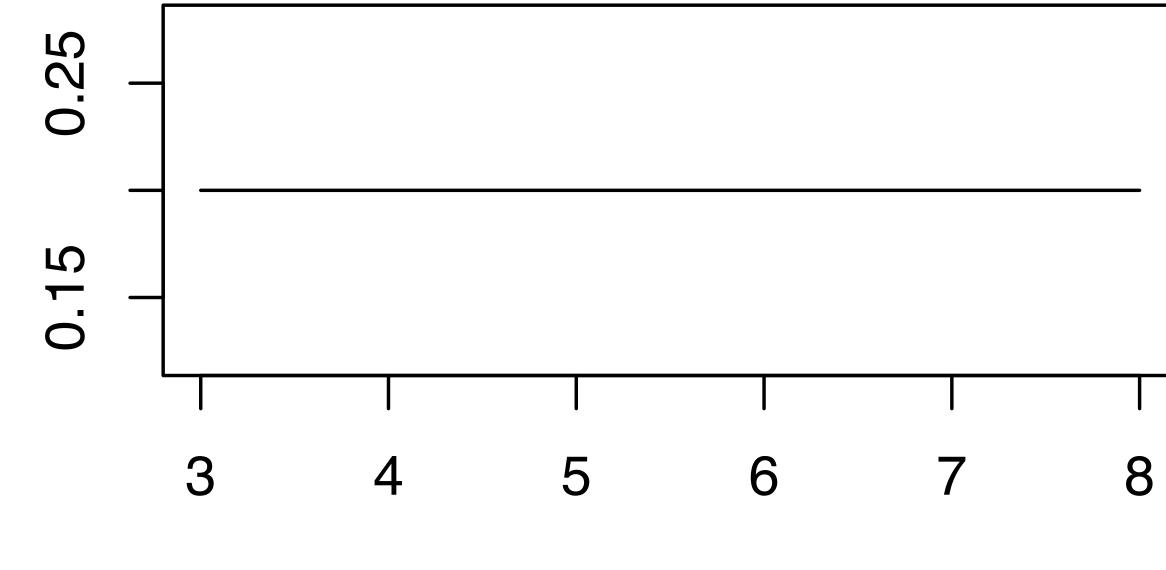
Χ

## Uniform(a,b)

- a = start
- b = end

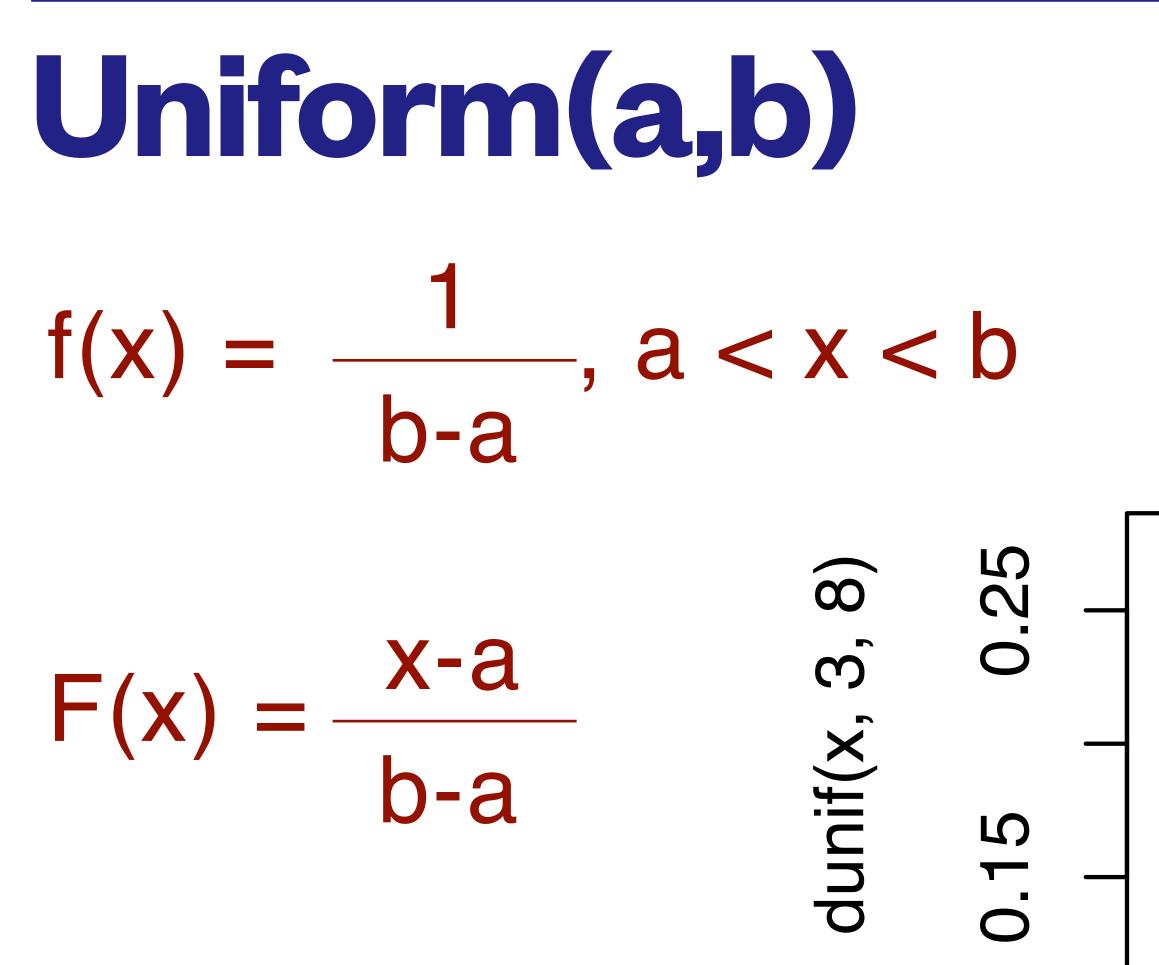
dunif(x, 3, 8)

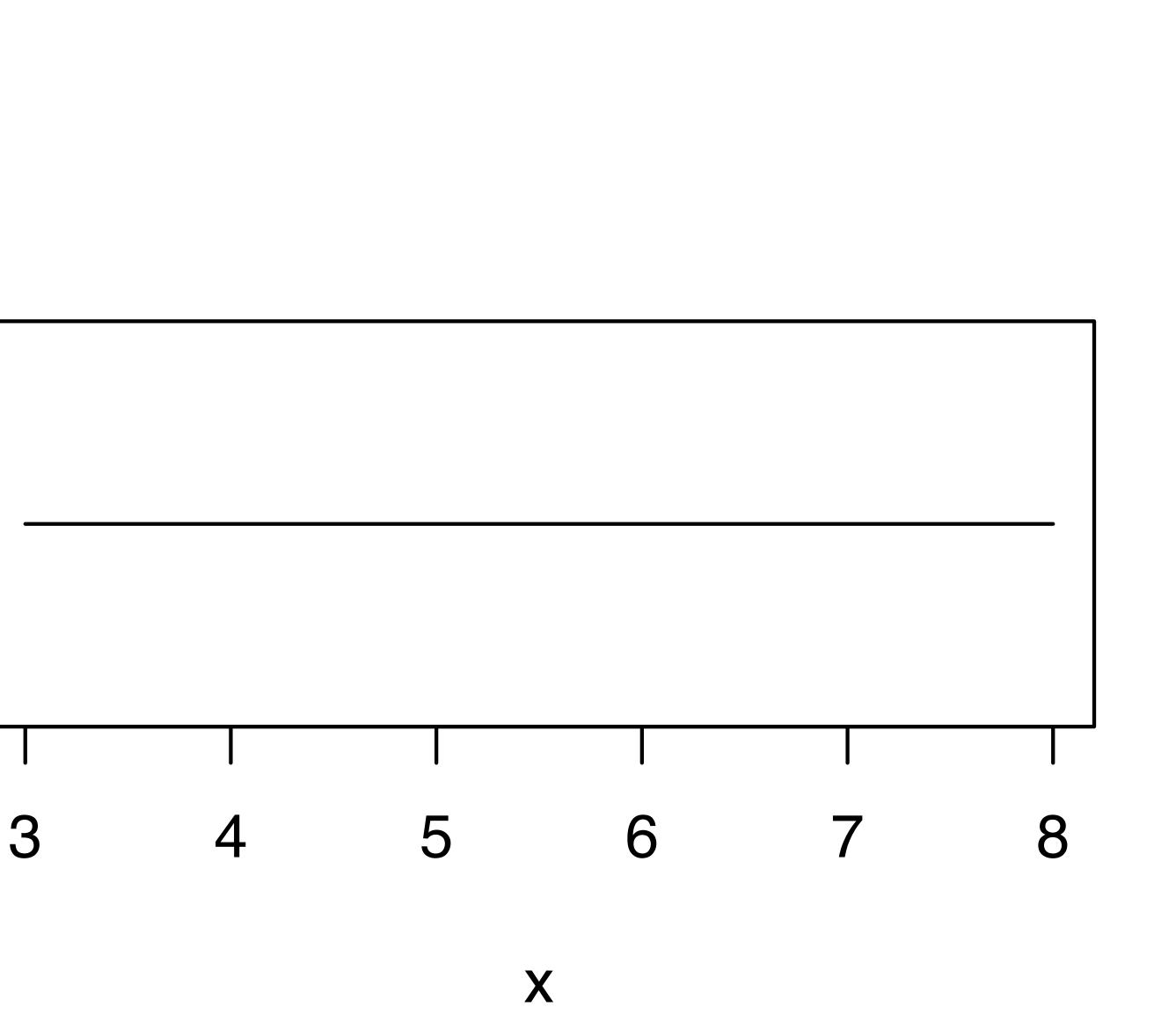
# f(x) = probability density function F(x) = cumulative distribution function



Χ







## Normal(m,s)

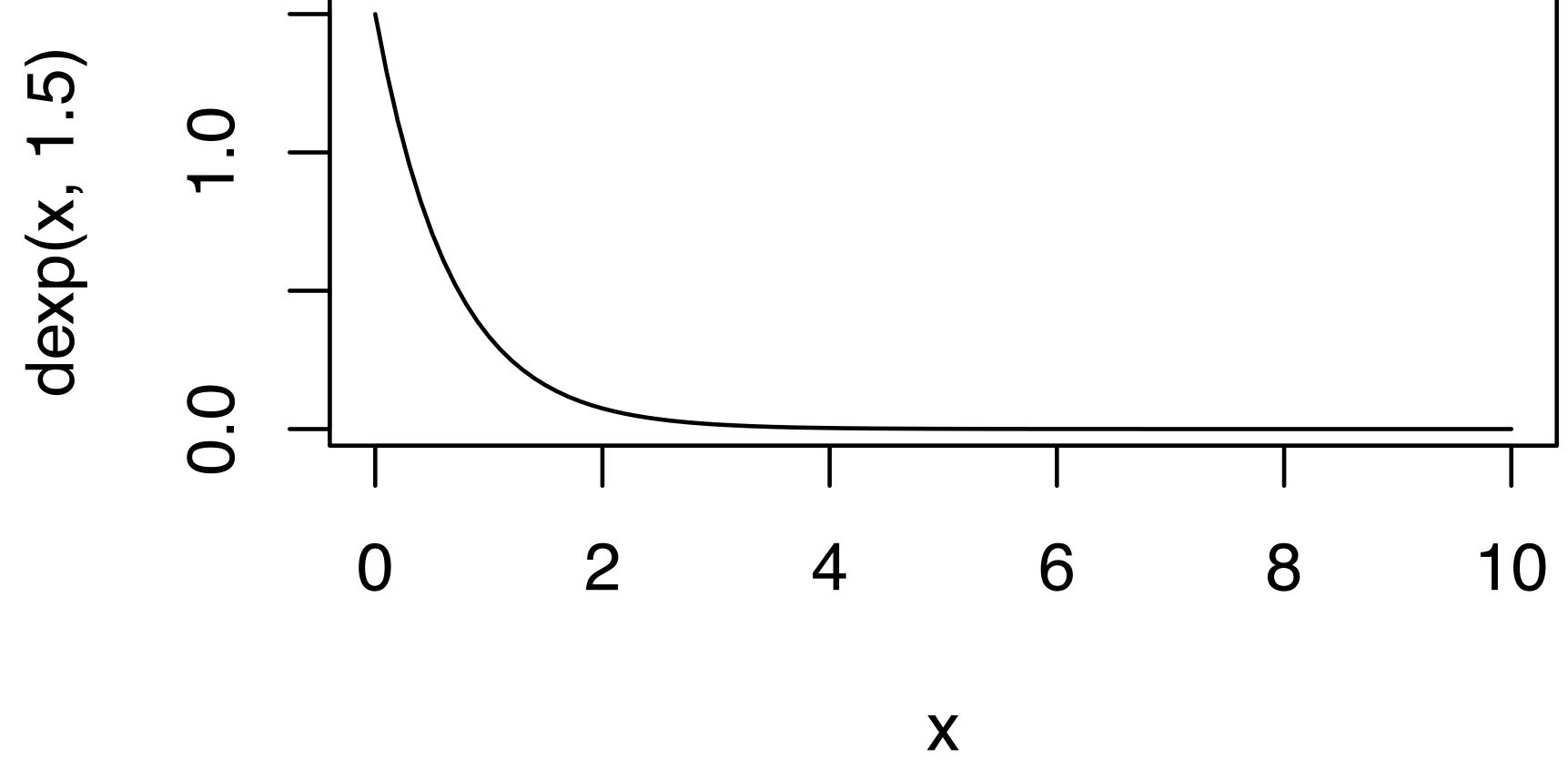
m = mean

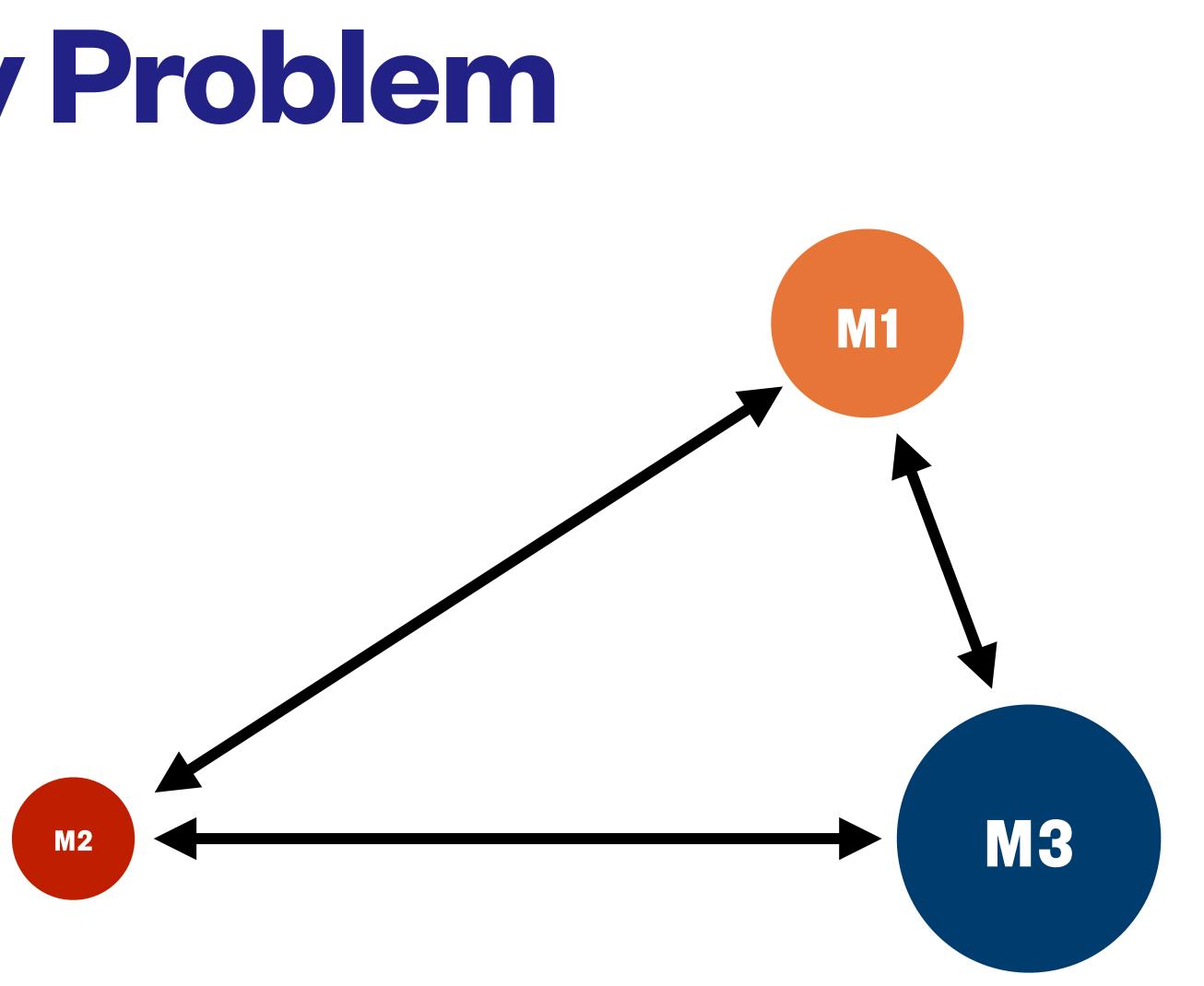
s = standard deviation

What is the range of Normal(m,s)?

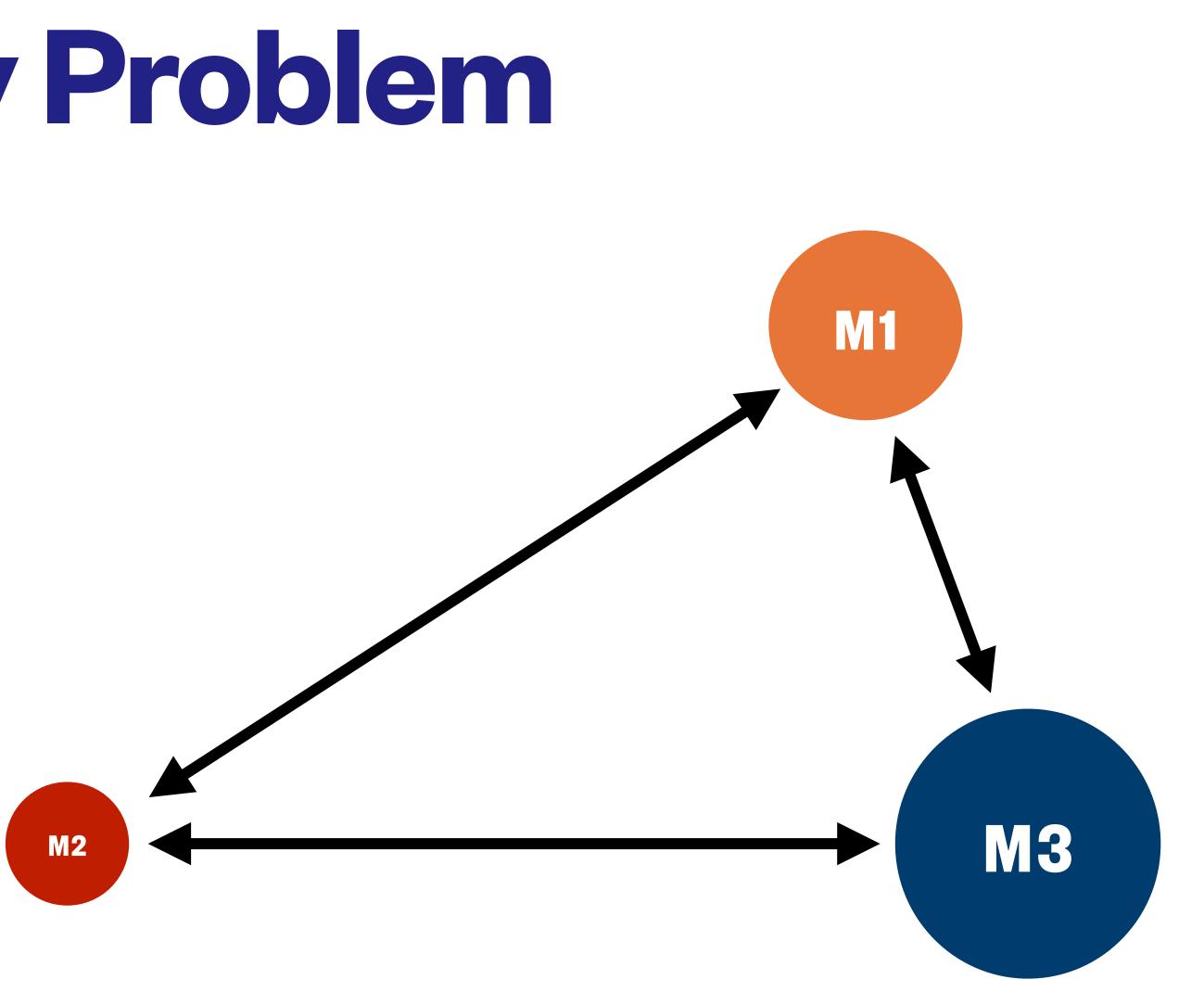
0.8 0.6 dnorm(x, 2, 0.5) 0.4 0.2 0.0 0 2 З 1

# Exponential(r) r = rate

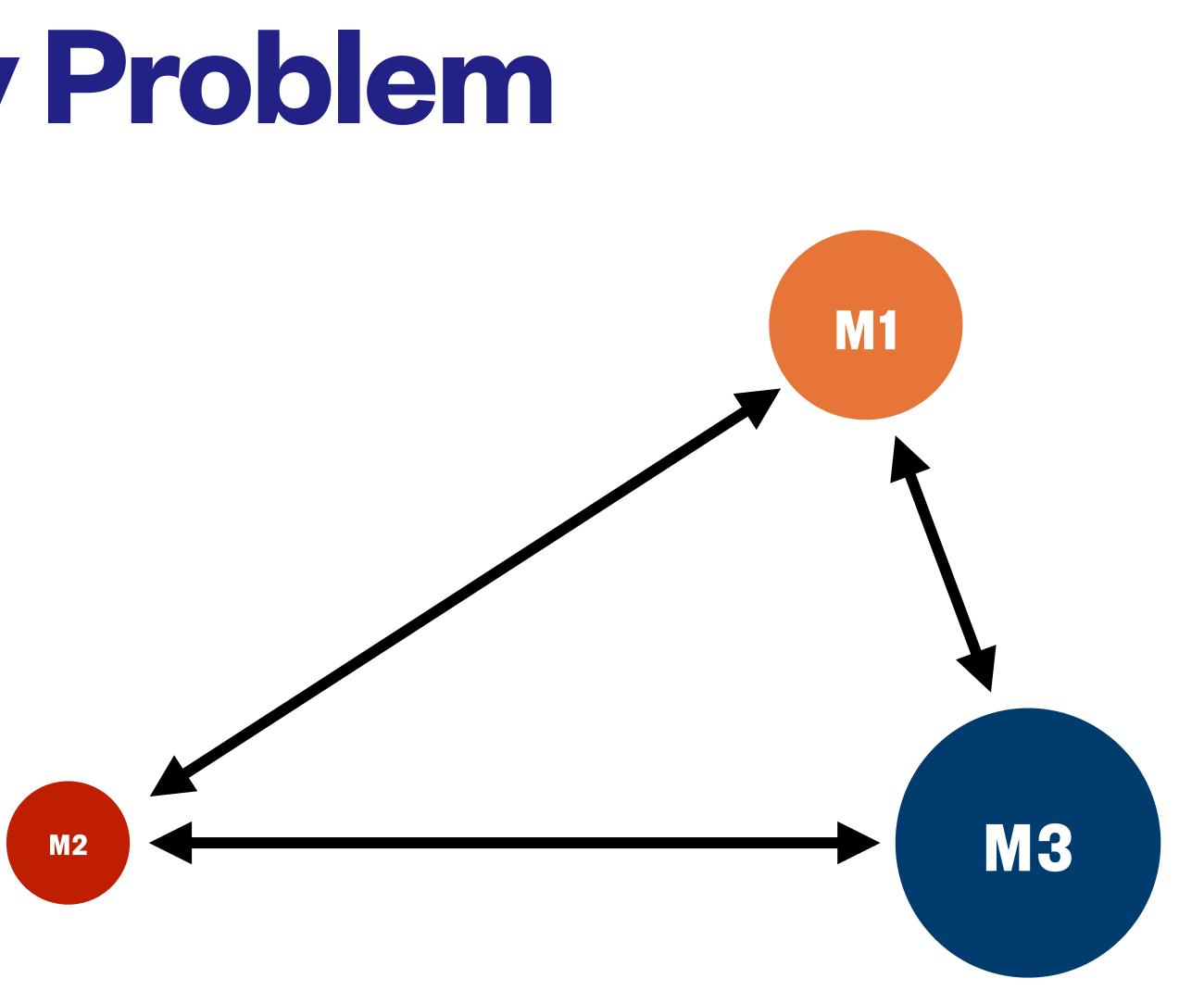




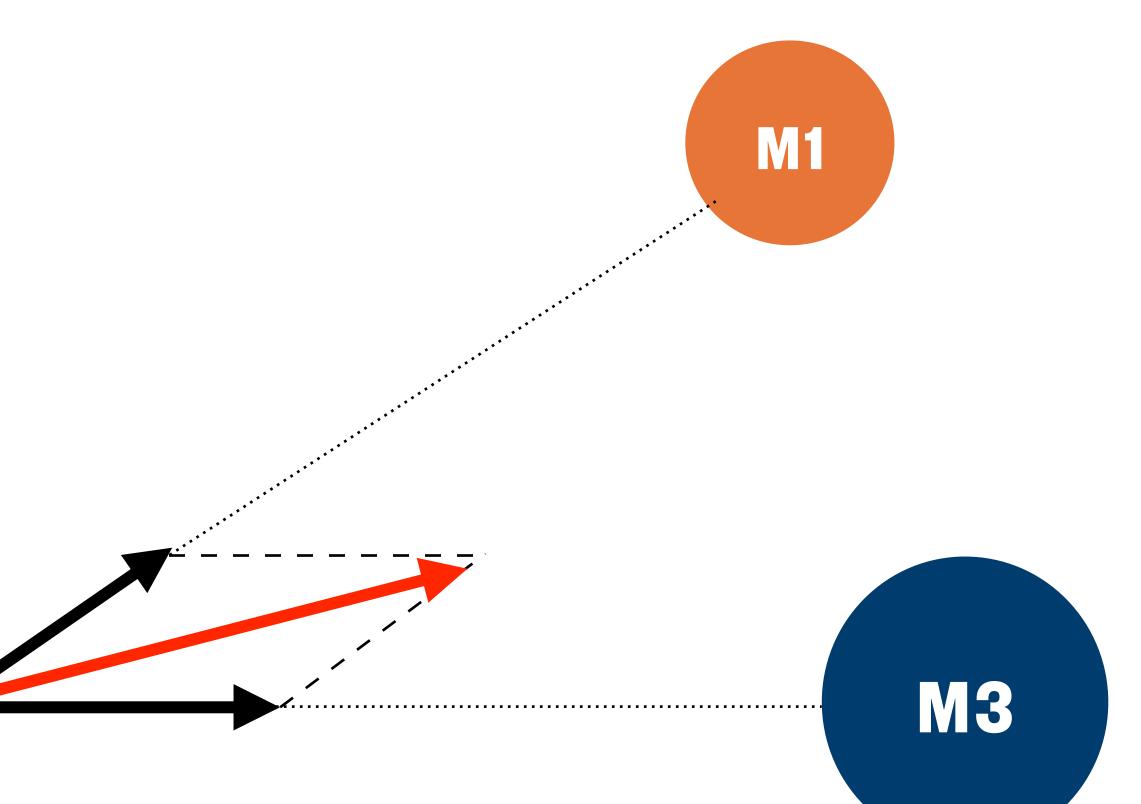
# What do you gain by simulating this?



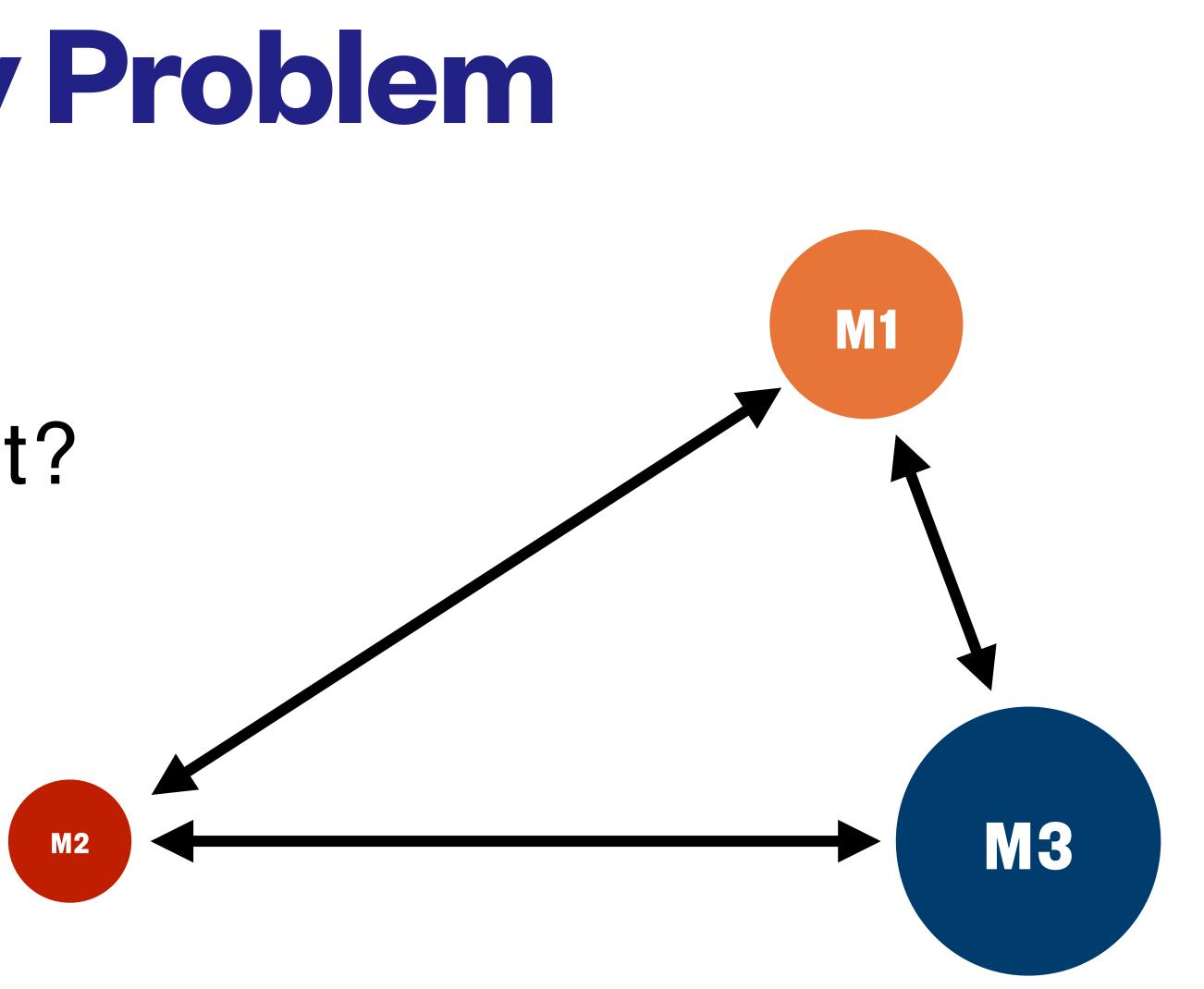
# What does the model need to capture?



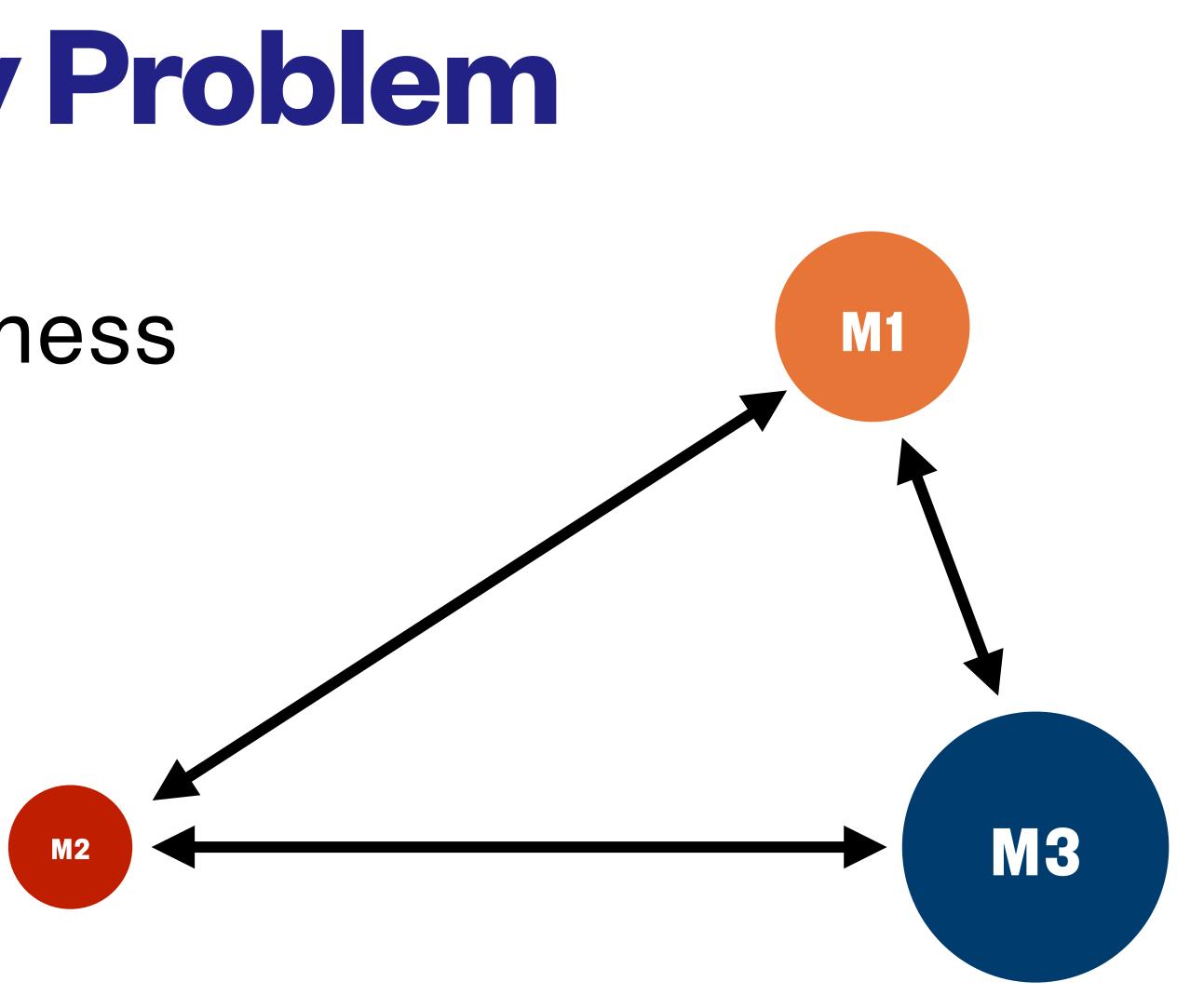
M2



# What defines each simulation experiment?

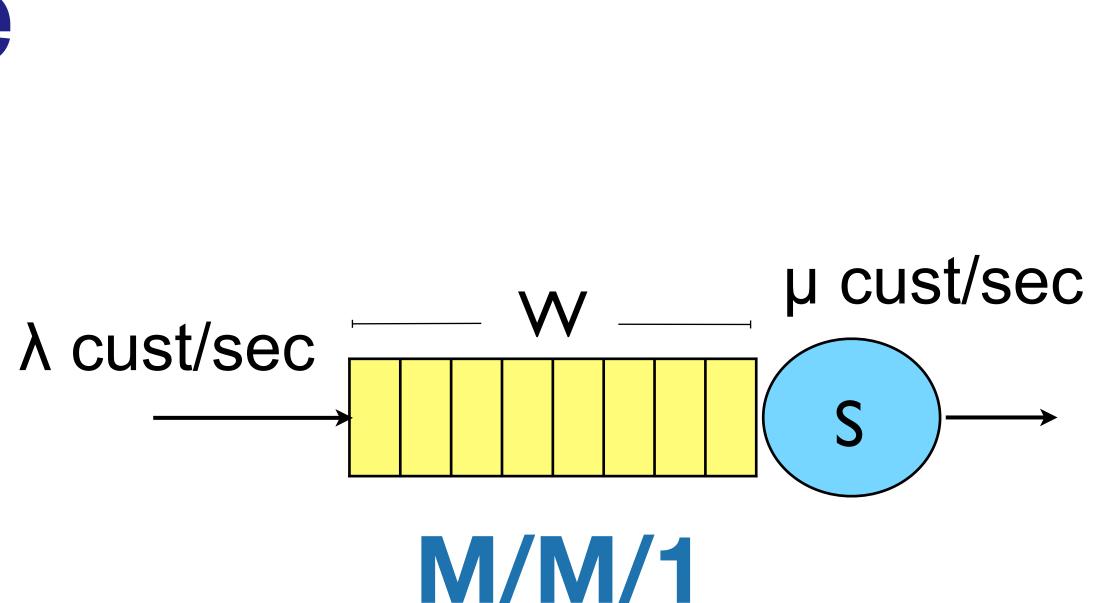


# Where is the randomness in this model?



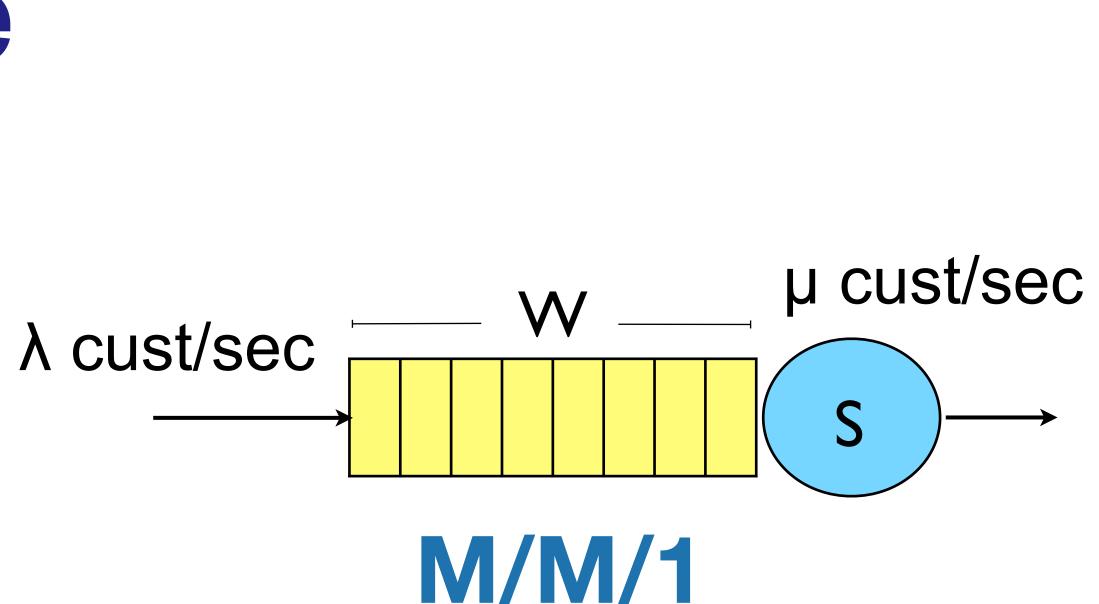
## The M/M/1 Queue

# What in the world is captured in this abstraction?



## The M/M/1 Queue

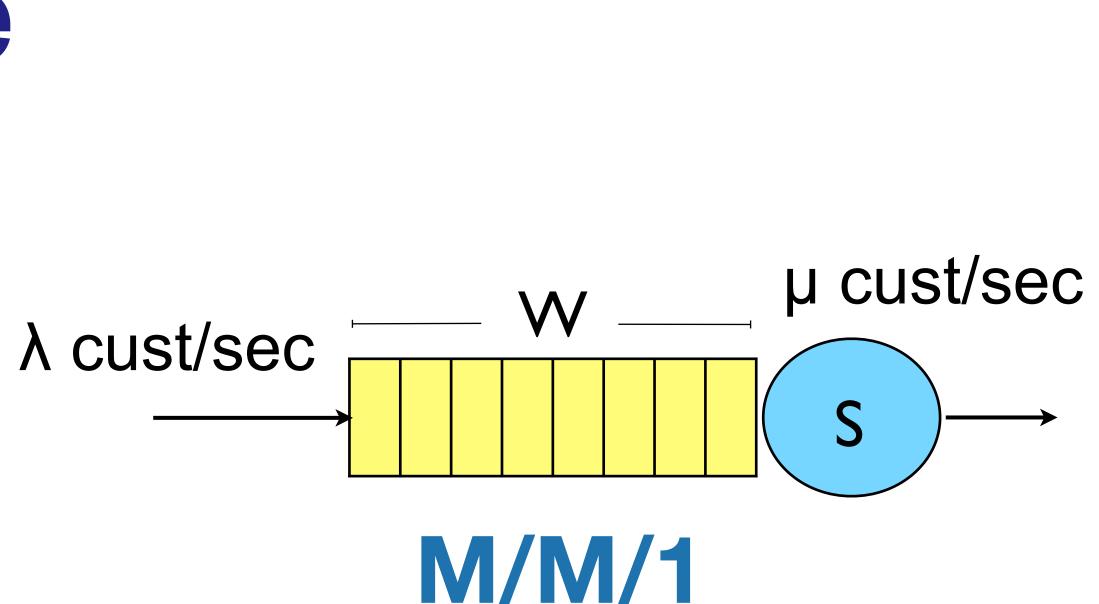
# What do you gain by simulating this?



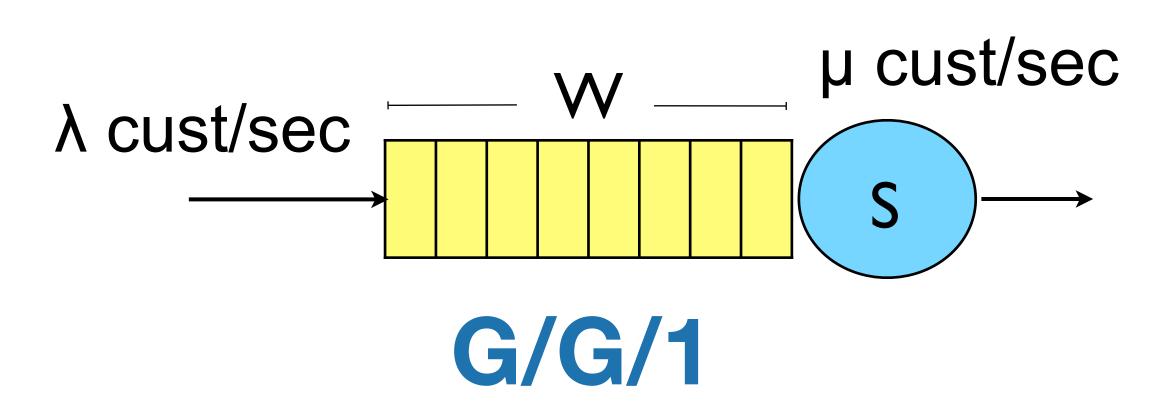
## The M/M/1 Queue

# What do you gain by simulating this?

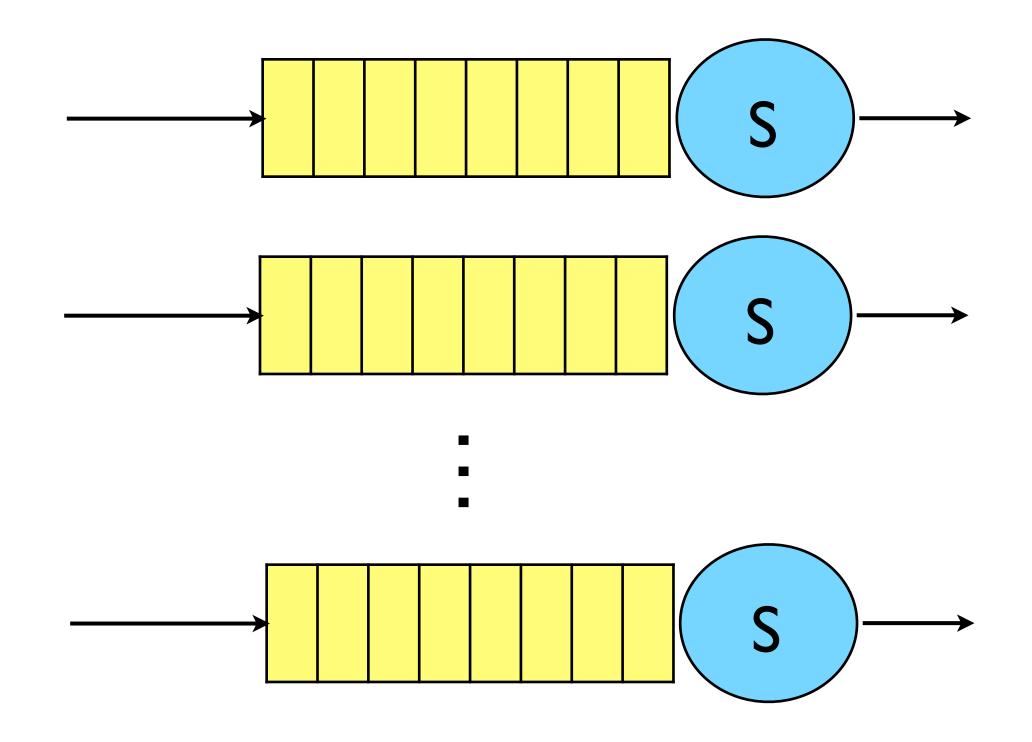
$$L = \frac{\frac{\lambda}{\mu}}{1 - \frac{\lambda}{\mu}}$$
$$W = \frac{1}{\mu - \lambda}$$



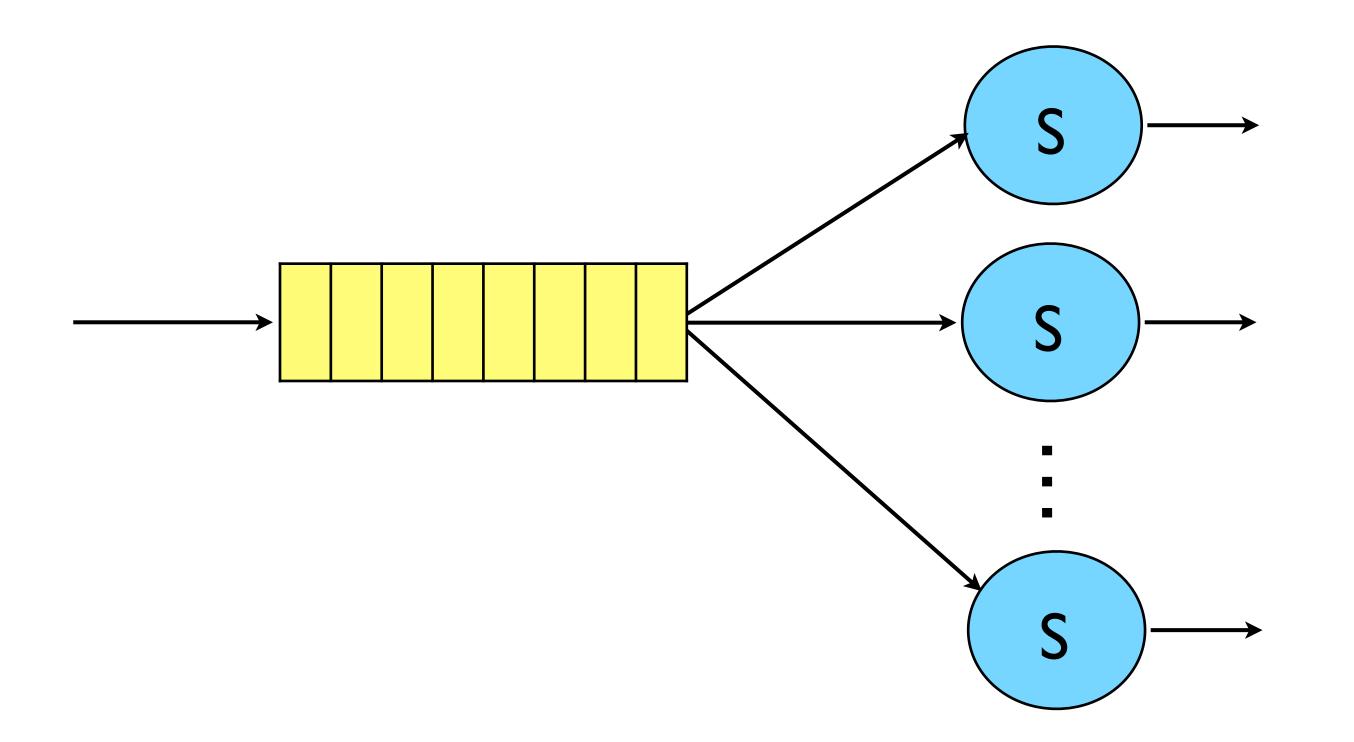
#### **Related Models**

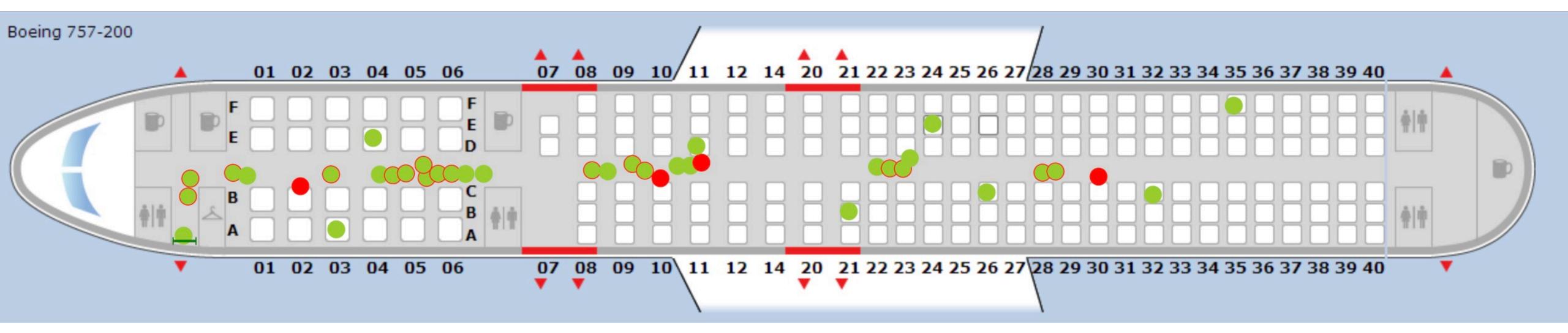


#### **Related Models**

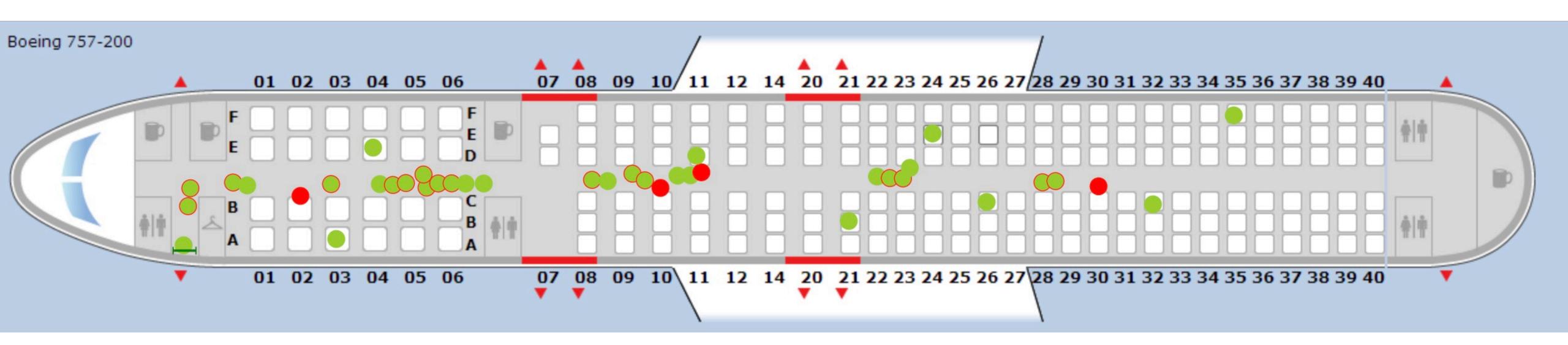


#### **Related Models**

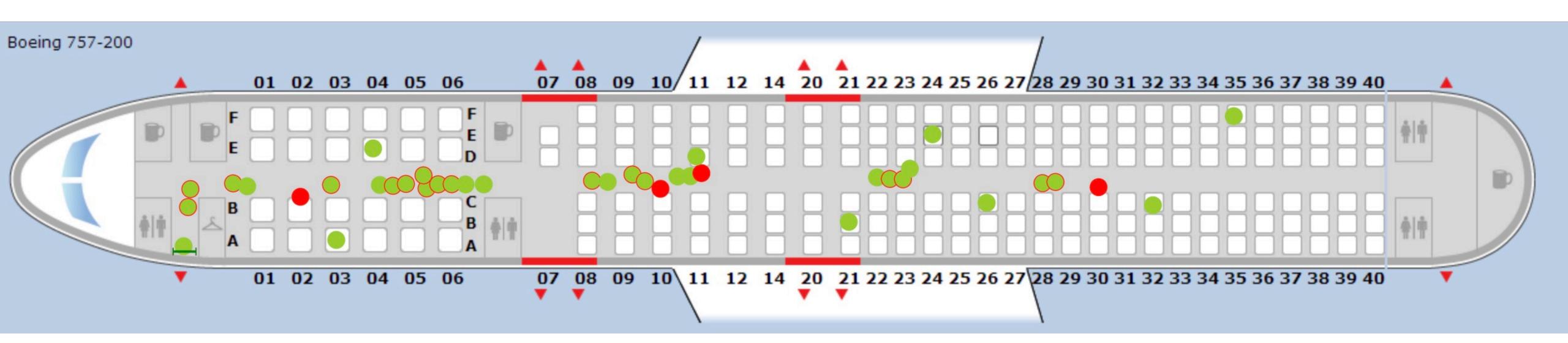




#### What do you gain by simulating this?



#### Where is the randomness in this model?



#### What do you learn from each simulation experiment?

