NCAIR 2024 Annual Conference Institutional Innovation

#### Enrollment Projection with Monte Carlo Simulation

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# What is a Monte Carlo Simulation?

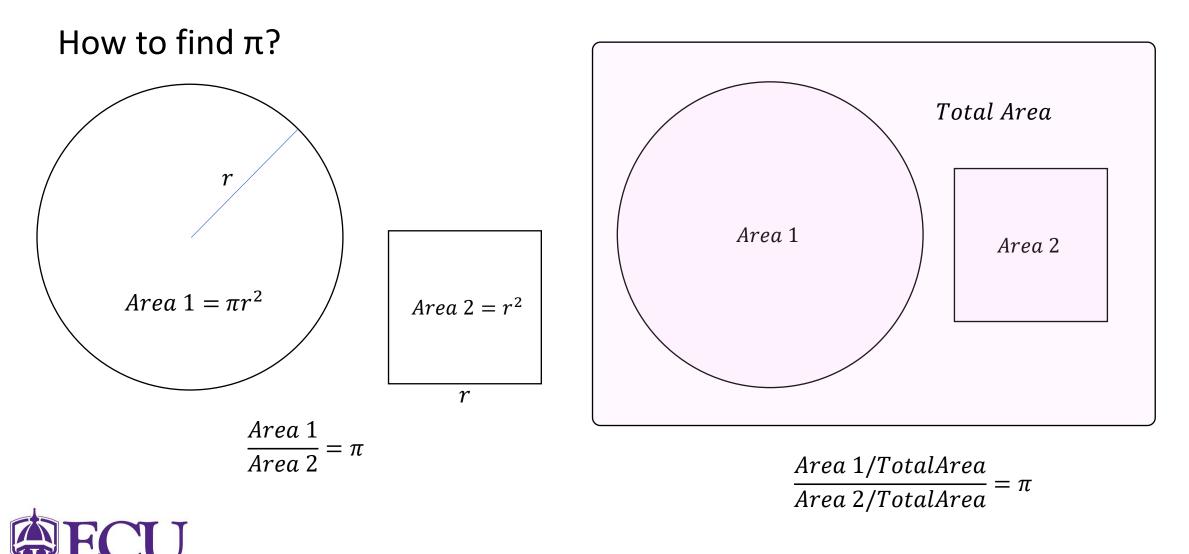
Monte Carlo: a casino in Monaco

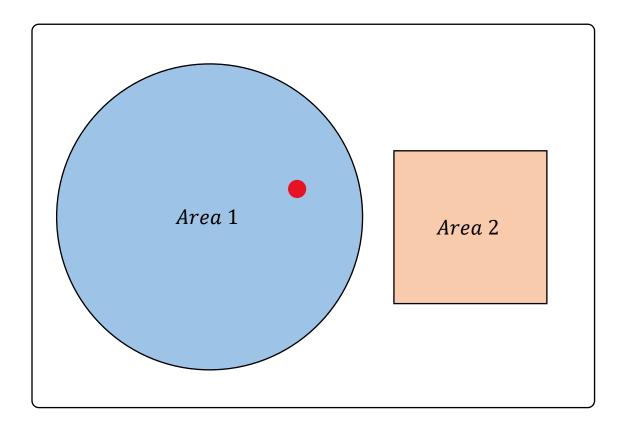


"Monte Carlo Simulation, also known as the Monte Carlo Method or a multiple probability simulation, is a mathematical technique, which is used to estimate the possible outcomes of an uncertain event."



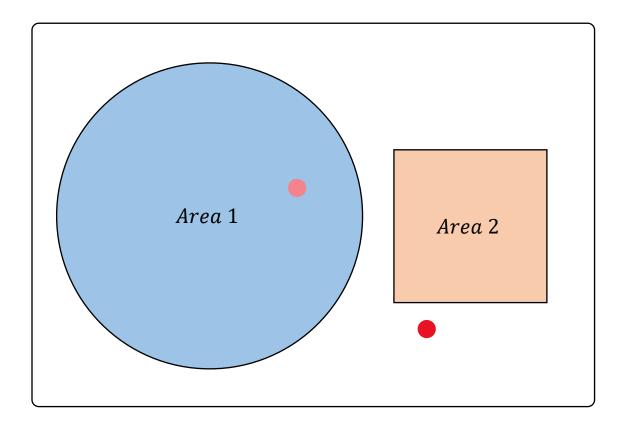
https://www.ibm.com/topics/monte-carlo-simulation





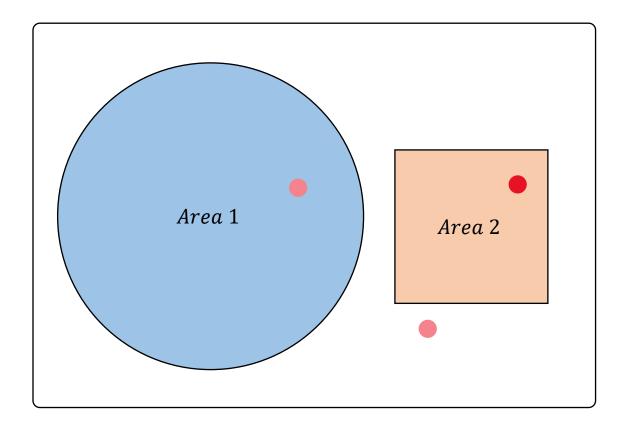
For each single dot, the probability of falling within the circle is  $\frac{Area\ 1}{Total\_Area}$ , the probability of falling within the square is  $\frac{Area\ 2}{Total\_Area}$ 





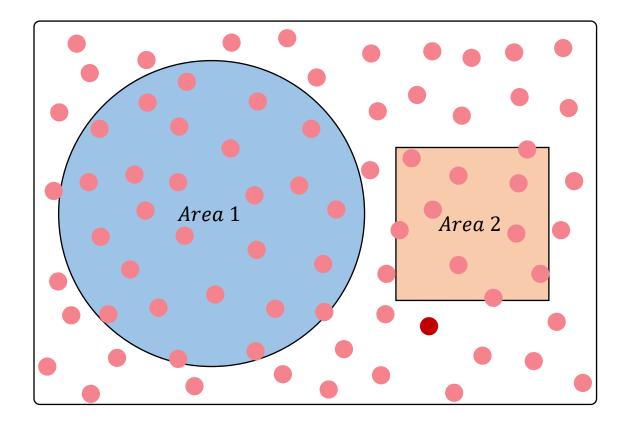
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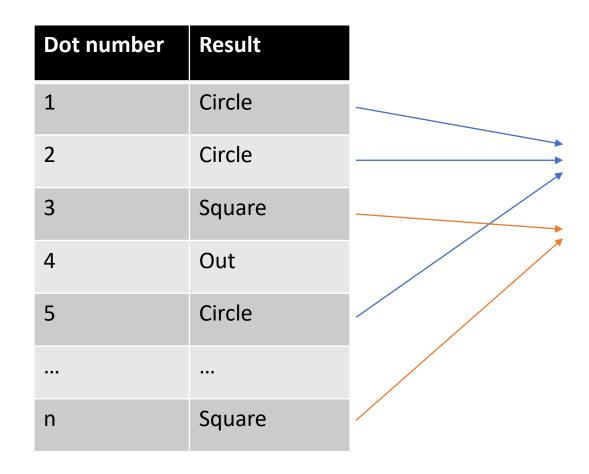
For each single dot, the probability of falling within the circle is  $\frac{Area\ 1}{Total\_Area}$ , the probability of falling within the square is  $\frac{Area\ 2}{Total\_Area}$ 

Since  $\frac{Area \ 1/Total\_Area}{Area \ 2/Total\_Area} = \pi$ 

In the long run,

 $\frac{\# of \ dots \ within \ the \ circle}{\# of \ dots \ within \ the \ square} = \pi$ 





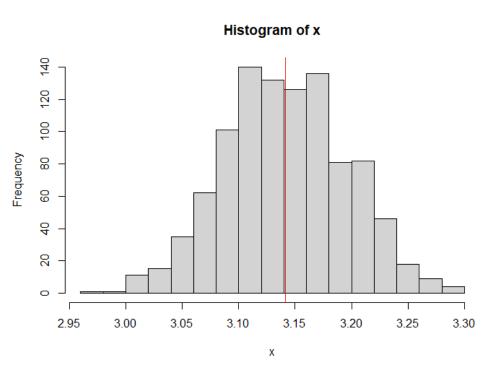
When n is getting larger, according to Law of Large Numbers:

 $\frac{\# of \ dots \ within \ the \ circle}{\# of \ dots \ within \ the \ square} \approx \pi$ 

 $P_{1}(Circle) = P_{2}(Circle) = \dots = P_{n}(Circle)$  $P_{1}(Square) = P_{2}(Square) = \dots = P_{n}(Square)$  $P_{1}(Out) = P_{2}(Out) = \dots = P_{n}(Out)$ 



Dot number	Simulation 1 result	Simulation 2 result	 Simulation n result
001	Circle	Circle	 Square
002	Circle	Square	 Out
003	Square	Out	 Circle
004	Out	Circle	 Circle
005	Circle	Circle	 Square
		•••	 
500	Circle	Out	 Circle
	$\pi_1$	$\pi_2$	 $\downarrow$ $\pi_n$

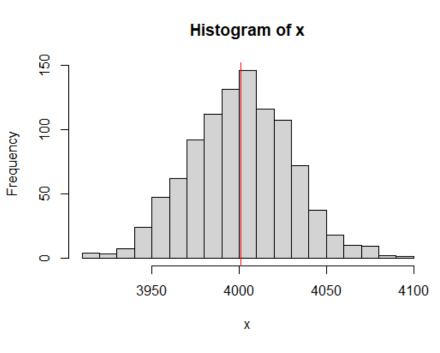


#### Simulation of 1000 times

Min. 1st Qu. Median Mean 3rd Qu. Max. 2.976 3.104 3.144 3.144 3.180 3.284

# Similarly...

Student ID	Simulation 1 result	Simulation 2 result	••••	Simulation n result
001	Retained	Retained		Not Retained
002	Not Retained	Retained		Retained
003	Retained	Retained		Retained
004	Retained	Not Retained		Retained
005	Retained	Retained		Not Retained
500	Not Retained	Retained		Retained
				$\downarrow$
	# Retained	# Retained		# Retained



#### Simulation of 1000 times

Min. 1st Qu. Median Mean 3rd Qu. Max. 3913 3982 4002 4001 4021 4094

#### Data set

- Dependent Variable • Retained in Fall (Y/N)
- Independent Variables
  - $\circ \ {\rm Age}$
  - $\circ \,\, \text{Gender}$
  - o Race/Ethnicity
  - Full Time/Part Time
  - Cumulative GPA
  - 0 ...

Step 1: Use predictive model to get the predicted retention probability for each student.

Student ID	Probability of Retention
0001	$P_{001}(Retention)$
0002	$P_{002}(Retention)$
0003	$P_{003}(Retention)$
0004	$P_{004}(Retention)$
0005	$P_{005}(Retention)$
5000	$P_{500}(Retention)$



	Col A	Col B	Col C
Row 1	Student ID	Probability of Retention	Simulation 1
Row 2	0001	0.389937	0
Row 3	0002	0.757576	1
Row 4	0003	0	0
Row 5	0004	0.111111	1
	0005	0.15	0
Row 5001	5000	0.142857	0

Step 2: Run simulation for all students.

=IF(\$B2>=RAND(),1,0) =IF(\$B3>=RAND(),1,0)

...

=IF(\$B501>=RAND(),1,0)



Step 3: Run as many simulations as you want

	Col A	Col B	Col C	Col D		Col AAA
Row 1	Student ID	Probability of Retention	Simulation 1	Simulation 2	•••	Simulation n
Row 2	0001	0.389937	0	1		0
Row 3	0002	0.757576	1	1		1
Row 4	0003	0	0	0		0
Row 5	0004	0.111111	1	0		0
	0005	0.15	0	0		1
Row 5001	5000	0.142857	0	1		0



Step 4: Calculate the sum for each simulation as expected total retention number

	Col A	Col B	Col C		Col D	•••	Col AAA
Row 1	Student ID	Probability of Retention	Simulation	1	Simulation 2	2	Simulation n
Row 2	0001	0.389937	0				
Row 3	0002	0.757576	1		1		1
Row 4	0003	0	0		0		0
Row 5	0004	0.111111	1		0		0
	0005	0.15	0		0		1
Row 5001	5000	0.142857	0				0
	-					01)	

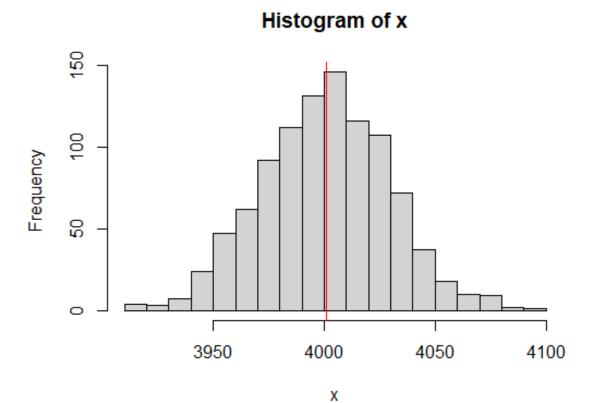


**=SUM(C2:C5001)** =SUM(D2:D5001)

Step 5: Descriptive analysis of all expected retention numbers

Simulation of 1000 times

Min. 1st Qu. Median Mean 3rd Qu. Max. 3913 3982 4002 4001 4021 4094





• Step 1: Use predictive model to get the predicted retention probability for each student

Example: binary logistic regression

```
# build the logistic regression
Im_ug <- glm(RETAINED ~., data = UG_train, family = "binomial")</pre>
```

# ....
# validation part omitted

This gives you the probability of Response = 1

# predict the retention by applying the model on the new data
projection\_24 <- predict(lm\_ug, newdata = UG\_23\_DATA, type = "response")</pre>



• Step 2: Run Monte Carlo Simulation

# set total number of simulations N=3000

Store the predicted total numbers in the vector

runif(): generate a random number in a uniform distribution

lapply(): apply a function
over a List or Vector

# build up a vector
total\_ug <- NULL</pre>

```
# create the function
pred_fun <- function(x){
    ifelse(x >= runif(1),1,0)
    }
```

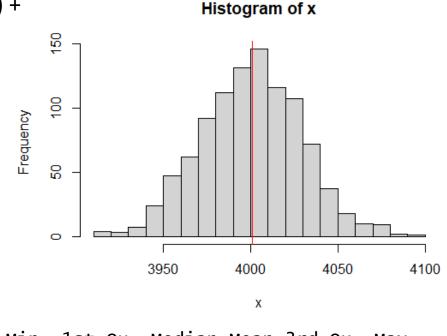
# run the simulation N times and store the sum into the vector
for (i in 1:N){
 df\_pred <- lapply(projection\_24,pred\_fun)
 total\_ug[i] = sum(unlist(df\_pred))</pre>

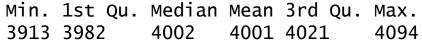


#### • Step 3: Check the result and make plot

```
ggplot(data = as.data.frame(total_ug), mapping = aes(x = total_ug)) +
geom_histogram(bins = 50) +
labs(title = "Retention projection of 2024 undergraduates") +
xlab("Estimated Retention") +
theme_light()
```

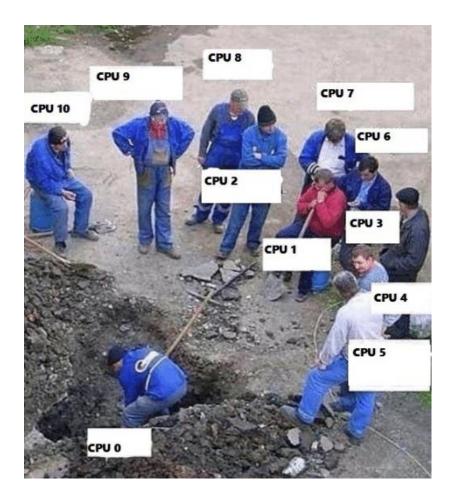
```
summary(total_ug)
quantile(total_ug, c(0.025,0.975))
```







### Performance Enhancement



By default, R runs only on a single thread on the CPU.

#### How to enhance the performance?

• Upgrade equipment



- Improve R code: vectorized functions
- Use parallel processing programming

#### Performance Enhancement – vectorized function

• Speed up Step 2: Monte Carlo Simulation (original code)

system.time()

user system elapsed 161.68 0.11 161.92

The problem: we evaluate this function too many times.

30,000 students X 3,000 simulations = 90,000,000 times!



```
# set total number of simulations
N=3000
```

# build up a vector
total\_ug <- NULL</pre>

```
# create the function
pred_fun <- function(x){
    ifelse(x >= runif(1),1,0)
    }
```

# run the simulation N times and store the sum into the list
for (i in 1:N){
 df\_pred <- lapply(projection\_24,pred\_fun)
 total\_ug[i]= sum(unlist(df\_pred))
 }</pre>

#### Performance Enhancement – vectorized function

• Speed up Step 2: Monte Carlo Simulation (improved code)

system.time()						
user system elapsed						
0.61	0.03	0.64				

# set total number of simulations

N=3000

# build up a vector total ug <- NULL

The ">=" function and "runif()" function only be called ONCE in each iteration of the loop. It saves you lots of time!

# run the simulation N times and store the sum into the list
for (i in 1:N){
 total\_ug[i] <- sum(projection\_24 >= runif(length(projection\_24)))
}



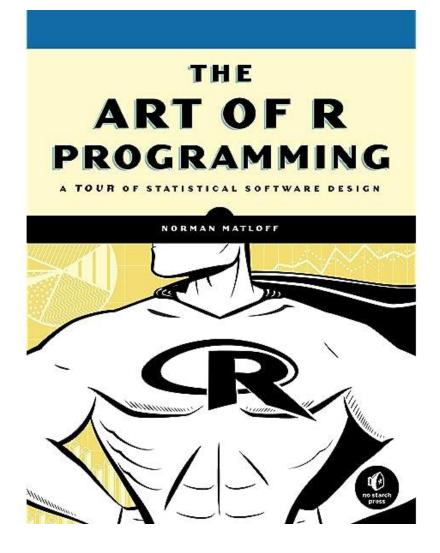
#### R Programming Resources

# Parallel and high performance computing with R

https://youtu.be/NWgOkKorFH4

#### **Parallel Programming with R**

https://youtu.be/O8PiX9ofXDI











#### Thank you for attending the 2024 NCAIR Annual Conference!

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