

EM384: Analytical Methods for Engineering Management

Lesson 35: Monte Carlo Simulation Applications in Python

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Lesson Objectives

Lesson Objectives

- Design and implement simple Monte Carlo simulations in python.
- Visualize the results of monte carlo simulation in python.
- Create and interpret an empirical cumulative distribution function (ECDF) from the results of the model.

Monte Carlo Simulation in Python

Monte Carlo Simulation

Given a system...



And ideas of how to model the system...



And some data that describes the behavior of the system...



- Determine how you plan to measure the performance of the system (system performance measures) → **will be your model output(s)**.
- Develop a **base case model of the system** that estimates system performance measures. You can start with simple diagram, essential mathematical relationships, and logic.
- **Identify components of the model susceptible to random variation.** Collect data that represents this random variation. If data is not available, estimate the random variation.
- Fit the observed data to known random variable distributions. These random variables will be used to create overall system behavior variance.
- Develop **pseudo-code** that iterates through multiple iterations of your model. Each iteration will incorporate a random outcome of the variables identified in the previous step.
- **Implement your pseudo code in Python.**

EM384 Standards for Monte Carlo Simulation in Python

Before starting: **Sketch out your model using pseudo-code!** (this is a general structure of your code using words and abbreviations). Then, in Python:

1. Import the required **libraries** at the top.
2. Define your fixed parameters by assigning values to **variables**. Pick names that make sense.
3. Define an **empty list (or array)** to in which save the results of your Monte Carlo Simulation.
4. Create a iteration structure (e.g. **FOR loop**) to iterate through your model. Inside your iteration structure:
 - 4.1 Define values that are **random variables**.
 - 4.2 Use calculations and logic to obtain the result(s) for the **current iteration**.
 - 4.3 **ADD the result of your current(s) iteration to your results list.**
5. Create a **histogram** of your results list data.
6. Create the **ECDF function** of your results list data
7. Use the ECDF to answer questions related to probability of events.

Practical Exercises

You recently purchased a car wash for your gas station. The wash component takes triangular(min=23,mode=30,max=35) and the dry component takes N(mean = 15 seconds, std dev = 5 seconds).

- Let x_1 = wash time, Triangular(min=23,mode=30,max=35)
- Let x_2 = dry time, Normal(mean = 15 seconds, std dev = 5 seconds)
- Let $x_1 + x_2 = y$ = complete car wash cycle time

What is the first and 99th percentiles for expected car wash times?

PE2 Struggling Deli

Your deli experiences a high volume of orders between 11am and 1pm for the lunch rush. Your restaurant specializes in salads, wraps, and burgers. About 30% of the meals are burgers, 50% are wraps, and 20% are salads. Your cook prepares meals with the following time distributions:

- Burgers take Normal(mean = 3, std dev = 1) minutes. Burger profit = \$1.50
- Wraps take Normal(mean = 2, std dev = 0.5) minutes. Wrap profit = \$0.75
- Salads take Normal(mean = 1.5, std dev = 0.25) minutes. Salad profit = \$0.50

Questions:

1. You have one cook that manages the orders in a first come, first serve sequence. How much profit do you expect to make during a typical lunch period?
2. What is the probability of making less than \$50?
3. What is the probability of making more than \$55?

Conclusion

Homework:

- Review Block 4 readings and previous Python tutorial videos.

Next Lesson:

- Design and implement simple Monte Carlo simulations in python.
- Visualize the results of monte carlo simulation in python.
- Create and interpret an empirical cumulative distribution function (ECDF) from the results of the model.