Introduction to Data Science



Last Update: 18 March 2020

Chapter 5 Prescriptive Analysis

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Outline Predictive Analysis

1. Prescriptive Analysis

- 2. Optimization
 - Optimization Problem
 - Defining an optimization problem
 - Applying predictive analytics to solve optimization problem
- 3. Simulation
 - What is simulation?
 - Applying Simulation to Prescriptive Analysis
 - Monte Carlo simulation

Prescriptive Analysis

The process of modelling a real-life or hypothetical situation on a computer.



Simulation is useful because it can be used to <u>test new ideas</u> about business decisions and actions to mitigate risk pertaining to a process or system, or how modifications will affect an existing process or system.



Simulation is often used in applications related to safety of infrastructure as well as safety, quality, and design of products.





Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters



The aim is to experiment with different queuing systems to understand their customers better.

- For instance, there are 20 counters in total.
- Each open counter costs \$20 to operate.
- Customers who wait more than 10 minutes are 'annoyed' and there's a penalty of \$1 per customer.

What is the optimal number of counters?

Applying Simulation to Prescriptive Analysis Example: Optimum number of open Check-out counters

What is the optimal number of counters?

Option 1: Perform experiment of different numbers of counters on real situation.

Option 2: Simulate experiment of different numbers of counters on computer.

Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters

Assumption:

Each counter has its queue and arriving customer joins one queue <u>randomly</u>.

Random under a certain distribution

Gamma distributions are assumed for inter-arrival time and service time for customers.

for inter-arrival time

Service time

Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters

Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters

Total Cost = $\sum_{i=1}^{M} 20 + \sum_{j=1}^{N} (t_j^{wait} > 10)$

- Each open counter costs \$20 to operate.
- Customers who wait more than 10 minutes are 'annoyed' and there's a penalty of \$1 per customer.

Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters

Applying Simulation to Prescriptive Analysis

Example: Optimum number of open Check-out counters

Queue Simulation

Source: <u>https://towardsdatascience.com/every-data-scientist-needs-to-read-these-simulation-stories-7be0531e782f</u>

Prescriptive Analysis

As can be seen,

Random Generator is invoked in simulation systems.

For example:

Prescriptive Analysis

Utilize the historical data!

Recall to descriptive analysis,

we can know some characteristics/behavior of random variables by perform descriptive analysis on historical data.

Then, we can simulate a system under the behavior.

Prescriptive Analysis

	Arrival time	Service time	End service time	Counter No.
x ₁	9:30	9:05	9:20	1
x ₂	9:33	9:21	9:30	1
\mathbf{x}_n	16:35	16:45	17:00	4

17:00

Arrival time

10:00 09:00 00:00 We can generate random numbers under the distribution.

Monte Carlo simulation

Computerized mathematical technique that allows people to account for risk in quantitative analysis and decision making.

Monte Carlo Casino, Monaco.

Pattern of Monte Carlo simulation process

Define a domain of possible inputs

Generate inputs randomly from a probability distribution over the domain

Perform a deterministic computation on the inputs

3

Aggregate the results

Example: Approximating the value of π

See animation at https://academo.org/demos/estimating-pi-monte-carlo/

Applying Monte Carlo simulation to decision making

Example: Monty Hall Problem

Suppose you're on a game show, and you're given the choice of three doors: Behind one door is a car; behind the others, goats.

- You pick a door, say No. 1,
- The host, who knows what's behind the doors, opens another door, say No. 3, which has a goat.
- He then says to you, "Do you want to pick door No. 2?" Is it to your advantage to switch your choice?

Applying Monte Carlo simulation to decision making

Example: Monty Hall Problem

Three initial configurations of the game. In two of them, the player wins by switching away from the choice made before a door was opened.

Source https://en.wikipedia.org/wiki/Monty_Hall_problem#/media/File:Monty_Hall_ Problem - Standard_probabilities.svg

Applying Monte Carlo simulation to decision making

Simulating Monty Hall with Monte Carlo

3 Define a domain of Generate inputs randomly from Perform a deterministic Aggregate the results possible inputs a probability distribution over computation on the inputs the domain **N Doors** – each door has Random to put car behind a Perform *M* times Calculate Prob. of either car or goat, w.r.t. door. success if we do not Count the number of success only one door has car. switch doors. Random the initial door. on switch, S_{switch}. **Initial door** – select 1 from Calculate Prob. of Count the number of success Random your decision. N doors. success if we switch on non-switch, S_{stav}. doors. **Decision** – stay or switch.

Applying Monte Carlo simulation to decision making Simulating Monty Hall with Monte Carlo

After we simulate 10000 games of Monty Hall with 3 doors We got the results as follows: Number of success on switch: 666566 66.6566% Number of success on non-switch: 333434 33.3434%

We can see that the success percentages don't vary much, and it tells us that if we make the switch then the chances of winning are 2 out of 3 times.

Further Study

Papers:

- Katerina Lepenioti, Alexandros Bousdekis, Dimitris Apostolou, Gregoris Mentzas, Prescriptive analytics: Literature review and research challenges, International Journal of Information Management, Volume 50, 2020, Pages 57-70, <u>https://doi.org/10.1016/j.ijinfomgt.2019.04.003</u>.
- Lepenioti K., Bousdekis A., Apostolou D., Mentzas G. (2019) Prescriptive Analytics: A Survey of Approaches and Methods. In: Abramowicz W., Paschke A. (eds) Business Information Systems Workshops. BIS 2018. Lecture Notes in Business Information Processing, vol 339. Springer, Cham

Website

- <u>https://towardsdatascience.com/every-data-scientist-needs-to-read-these-simulation-stories-7be0531e782f</u>
- <u>https://towardsdatascience.com/a-zero-math-introduction-to-markov-chain-monte-carlo-methods-dcba889e0c50</u>