

Optimization and Simulation

Introduction to simulation

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Outline

- 1 Modeling
- 2 Causal effects
- 3 Uncertainty
- 4 Beyond the mean
- 5 Simulation

Modeling

System

- A system can be seen as a black box, modeled by

$$z = h(x, y, u)$$

- Example: a car
- x captures the state of the system (e.g. speed, position of other vehicles)
- y captures external influences (e.g. wind)
- u captures possible human controls on the system (e.g. acceleration/deceleration)
- z represents indicators of performance (e.g. oil consumption).

Modeling

Decompose the complexity

- The model h is usually decomposed to reflect the interactions of the subsystems
- For example,
 - a car-following model captures the target speed of the driver,
 - an engine model derives the actual consumption as a function of the acceleration.

Simulation

- Captures the causal effects.
- Captures the uncertainty.

Simulation

Definition

the act of imitating the behavior of some situation or some process by means of something suitably analogous

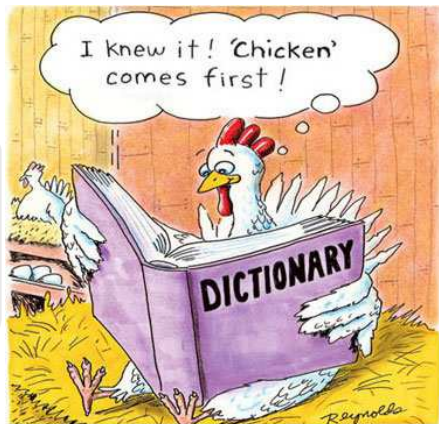
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Modeling

Causal effects

- Very important to identify the causal effects
- Failure to do so may generate wrong conclusions



Example: improving safety

Accidents in Kid City

- The mayor of Kid City has commissioned a consulting company
- Objective: assess the effectiveness of safety campaigns
- They propose to use simulation

Example: improving safety

Accidents in Kid City



Example: improving safety

Accidents in Kid City: 



Example: improving safety

Accidents in Kid City



Example: improving safety

Accidents in Kid City



Example: improving safety

Accidents in Kid City: 



Example: improving safety

Two major flaws

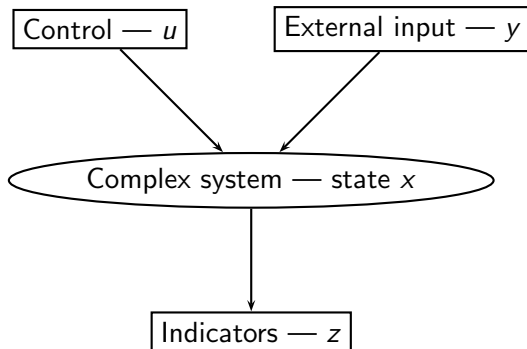
- Causal effects are not modeled
- Simulation performed with only one draw

Simulation: what it is not



Simulation: what it is not

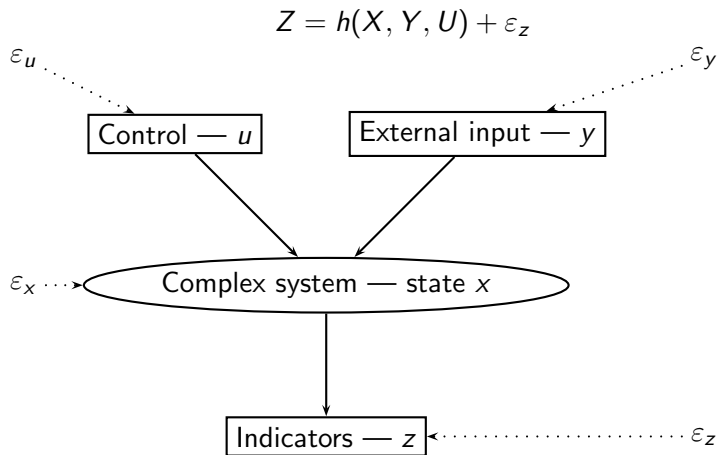
$$z = h(x, y, u)$$



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Simulation

Propagation of uncertainty

$$Z = h(X, Y, U) + \varepsilon_Z$$

- Given the distribution of X , Y , U and ε_Z
- what is the distribution of Z ?

Derivation of indicators

- Mean
- Variance
- Modes
- Quantiles

Simulation

Sampling

- Draw realizations of X, Y, U, ε_Z
- Call them $x^r, y^r, u^r, \varepsilon_Z^r$
- For each r , compute

$$z^r = h(x^r, y^r, u^r) + \varepsilon_Z^r$$

- z^r are draws from the random variable Z



Statistics

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"Numbers don't lie. That's where we come in."

Indicators

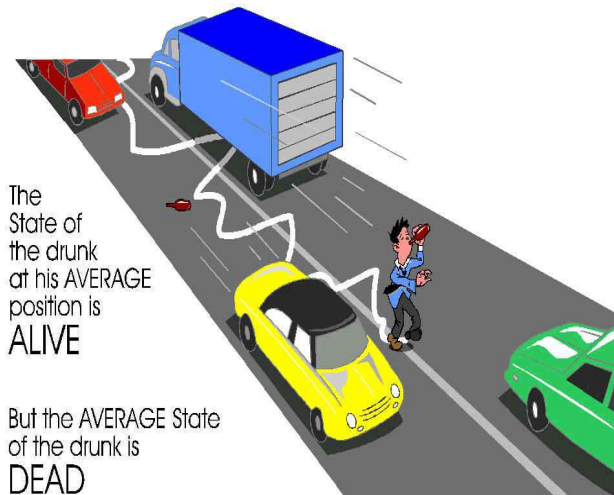
- Mean: $E[Z] \approx \bar{Z}_R = \frac{1}{R} \sum_{r=1}^R z^r$
- Variance: $\text{Var}(Z) \approx \frac{1}{R} \sum_{r=1}^R (z^r - \bar{Z}_R)^2$.
- Modes: based on the histogram
- Quantiles: sort and select

Important: there is more than the mean

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The mean



[Savage et al., 2012]

The mean

The flaw of averages

[Savage et al., 2012]

$$E[Z] = E[h(X, Y, U) + \varepsilon_z] \neq h(E[X], E[Y], E[U]) + E[\varepsilon_z]$$

... except if h is linear.

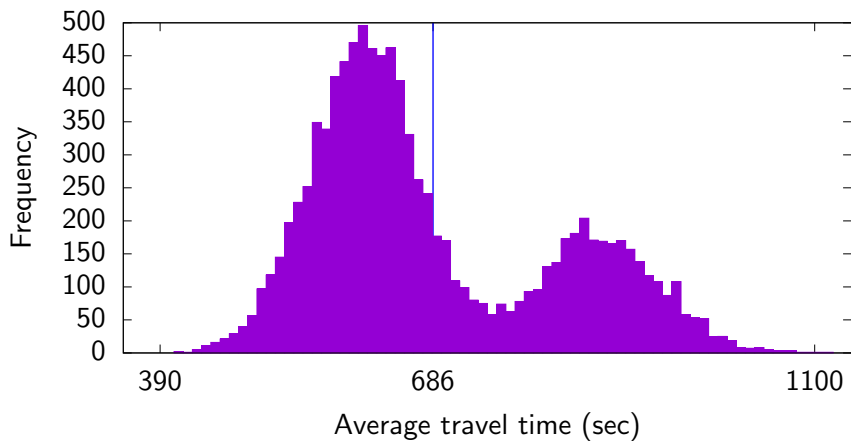
There is more than the mean



Example

- Intersection with capacity 2000 veh/hour
- Traffic light: 30 sec green / 30 sec red
- Constant arrival rate: 2000 veh/hour during 30 minutes
- With 30% probability, capacity at 80%.
- Indicator: Average time spent by travelers

There is more than the mean



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Pitfalls of simulation

Few number of runs

- Run time is prohibitive
- Tempting to generate partial results rather than no result

Focus on the mean

- The mean is useful, but not sufficient.
- For complex distributions, it may be misleading.
- Intuition from normal distribution (mode = mean, symmetry) do not hold in general.
- Important to investigate the whole distribution.
- Simulation allows to do it easily.

Challenges

- How to generate draws from Z ?
- How to represent complex systems? (specification of h)
- How large R should be?
- How good is the approximation?

Pseudo-random numbers

Definition

- Deterministic sequence of numbers
- which have the appearance of draws from a $U(0, 1)$ distribution

Typical sequence

$$x_n = ax_{n-1} \text{ modulo } m$$

- This has a period of the order of m
- So, m should be a large prime number
- For instance: $m = 2^{31} - 1$ and $a = 7^5$
- x_n/m lies in the $[0, 1[$ interval

Outline of the lectures

- Drawing from distributions
- Discrete event simulation
- Data analysis
- Variance reduction
- Markov Chain Monte Carlo

Reference

[Ross, 2006]

Bibliography



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Simulation.

Elsevier, fourth edition.



Savage, S., Danziger, J., and Markowitz, H. (2012).

The Flaw of Averages: Why We Underestimate Risk in the Face of Uncertainty.

Wiley.