

# Course Introduction & Stochastic Simulation

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# Something about Me



- **Student life at Texas**
  - major: operations management
  - minor: statistics
- ***Teach mental gymnastics at NCCU***
  - management science
  - decision science
  - advanced quantitative methods



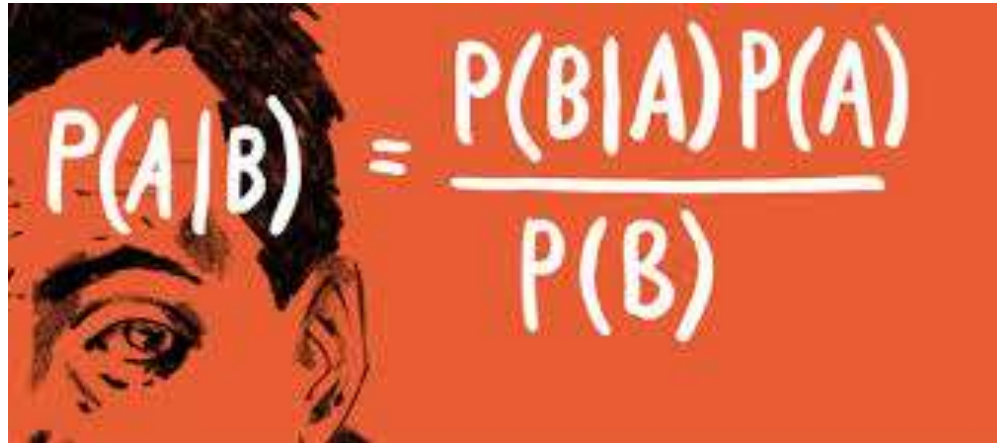
Howard Chuang

- **Personal facts**
  - a nice...**liberal** but **demanding(O)** professor
  - a fan of hiking, driving, biking, & traveling
    - Penghu, Taitung, & Kyoto are the best!
  - a lover of coffee, coke, & beer
  - wish to be a **Texas Poker** player after retirement



- **Dream big? Start small, think deep, & be good**




$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

**Let's go over the Syllabus!**

**fear NOT & work HARD**



# Decision Science in Action

time      Vbike / Uber      traffic      speed

- For key performance outcome  $Y=f(X, Z; c)$ 
  - analyze how two sets of variables – controllable  $X$  & non-controllable  $Z$  – and parameters  $c$  affect  $Y$ 
    - low-dim or high-dim  $X$ ? deterministic or stochastic  $Z$ ?
    - need to estimate  $c$  from subjective/objective data
  - probability distributions for stochasticity/uncertainty
    - well-defined or unknown  $f(\cdot)$ ?
    - machine learning/deep learning for unknown  $f(\cdot)$
  - $f(\cdot)$  may be 1 or  $>1$  equations for a problem or system
  - finally, try to optimize or improve  $X$



Predictive modelling

- Estimate the target for new observations

$$y = f(x)$$



Explanatory modelling

- Describe the effect that a change of certain inputs has on the target

$$y = f(x)$$



Optimisation

- Find the inputs that give optimal performance
- $f$  is known

$$y = f(x)$$



# Simulation for Outcome Estimation



- **Monte Carlo (蒙地卡羅) simulation** is 用電腦做重複抽樣  
– for estimating the value of an *unknown quantity* using **repeated random sampling**
  - Ulam & Von Neumann(馮紐曼) at the Manhattan project (曼哈頓計畫) in 1940s for solitaire (接龍)

ex:

21點

strategy

{ 1 張  
2 張  
3 張





# Simulation for Outcome Estimation



- **Monte Carlo (蒙地卡羅) simulation** is
  - for estimating the value of an *unknown quantity* using **repeated random sampling**
    - Ulam & Von Neumann(馮紐曼) at the Manhattan project (曼哈頓計畫) in 1940s for solitaire (接龍)
  - **Experiment**: Simulate a system or process with *uncertain/random/stochastic* outcomes for  $S$  times. Every time we keep records of key metrics (e.g., cost, profit) or the occurrence of an event ( $E$ ). As we have a **sampling distribution of  $S$  observations**, we can do simple calculation to obtain
    - sample average
    - $P(E)$  (i.e., probability of  $E$  occurring)
    - sample variance
  - **Law of Large Numbers**: When  $S$  approaches infinity, sample estimates are close to population true values



# Simulation for Outcome Estimation

- Toss a FAIR coin (擲銅板)

- event, E: Head shows up

- what is  $P(E)$ ?  $\frac{1}{2}$

- simulate the process for S times

- $S=10, 100, 10000$ , or...?

- if **output variability (變異度)** is **high**, **confidence (信心)** in estimate will be **low** ↓ 每一次的結果差異大

- A coin-tossing game (銅板賭局)

- KoP & Han toss a coin 50 rounds, Head KoP+\$1, Tail koP-\$1, KoP begins with \$0 in his pocket

- How likely KoP breaks even? What is E (event) here?

- can't we just do the math? **why computer simulation?**

- What is the # of rounds KoP has > \$0 in the game?

- What is KoP's highest fortune in the game?

- Will KoP's max cumulative earning > \$10?



# Simulation for Outcome Estimation

HW:  
Sep 13

- **Toss a FAIR die (擲骰子)**
  - KoP & Han in turn throw a die. In each round, one of three possible outcomes – b, s, & t – may occur
    - b: KoP bigger, s: KoP smaller, t: KoP & Han tie
  - KoP & Han do this for R rounds in a game. If the # of b > # of s in R rounds, KoP will be the game-winner
- **Define E: KoP wins the game**
  - **what is the exact  $P(E)$  mathematically?**
  - **can you use simulation to estimate  $P(E)$ ?**
    - show  $P(E)^{\text{exact}}$  &  $P(E)^{\text{sim}}$  for  $R=[5, 10, 15, \dots, 95, 100]$
    - what's the impact of S on approximation error?

