

CS5340: Tutorial 1

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1. Two Numbers Game

- Team 1:
 - Pick 2 **different** numbers between 0 and 10 (inclusive).
 - Write each number on a piece of paper each.
 - Turn the papers face down.
- Team 2:
 - Objective is to pick the **larger number**.
 - Pick one of the pieces of paper.
 - Have a peek at the number.
 - **Decide:** do you keep this number or *switch*?
- **Question:** Can Team 2 win more than 50% of the time?

Yes!

Team 2 (Cheating Strategy)

- Team 2 has a **spy** in Team 1
- Team 1 picks two numbers L and H
- Spy tells you a number between L and H
 - If L and H are next to one another, then spy tells you L
- Example:
 - Team 1 picks 3 and 8
 - Spy says 5
 - Team 2 picks randomly.
 - Team 2: Do you switch?
- Team 2 has a 100% of winning!



From Just now: Academic Honesty

- Please be academically honest.
 - “Give credit where credit is due”

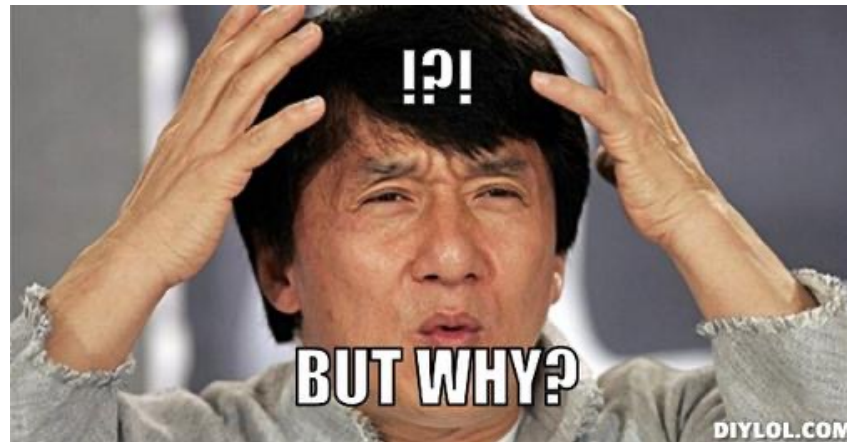


Do ***NOT*** cheat

- Strict Plagiarism policy:
 - If you cheat, we will report you to the disciplinary board
 - If found guilty, you will get an F (University Policy)

Team 2 Strategy

- Randomly pick $z \in [0, 10)$
- Take a peek at one of the numbers, lets call this x
- If $x \leq z$, switch, else stick with x

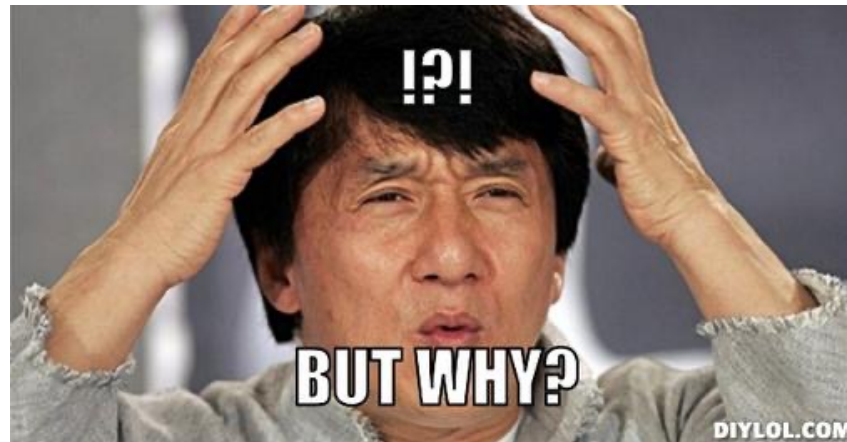


In Groups

- Introduce yourselves.
- Split into Team 1 and Team 2 and play this game.
- Discuss and analyze why the randomized strategy works.
 - Use the techniques learnt in the lecture (tree diagram)
- What is the probability Team 2 wins using this strategy?
- If you are done, move on to the other problems in the tutorial.

Team 2 Strategy

- Randomly pick $z \in [0, 10)$
- Take a peek at one of the numbers, lets call this x
- If $x \leq z$, switch, else stick with x



Strategy Analysis

- Let $L < H$ be the numbers chosen by Team 1.
- 3 possible cases:
 - **Case just-right:** $L \leq Z < H$:
 - Team 2 wins always!
 - $p(\text{win}|\text{justright}) = 1$ and $p(M) \geq \frac{1}{10}$
 - **Case too-high:** $H \leq Z$:
 - Team 2 switch. Only wins if picked L
 - $p(\text{win}|\text{toohigh}) = \frac{1}{2}$
 - **Case too-low:** $Z < L$:
 - Team 2 stays. Only wins if picked H
 - $p(\text{win}|\text{toolow}) = \frac{1}{2}$

$$p(\text{win}) \geq \left(1 \times \frac{1}{10}\right) + \frac{1}{2} \left(1 - \frac{1}{10}\right)$$
$$p(\text{win}) \geq \frac{11}{20}$$

More complete derivation in:
MIT Math for CS, Chapter
18.3.3

https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-042j-mathematics-for-computer-science-spring-2015/readings/MIT6_042JS15_Session31.pdf

Strategy Analysis (general case)

- Team 1 chooses 2 numbers $[0, \dots, n]$

- Team 2 picks half-integers:

$$\frac{1}{2}, \frac{3}{2}, \dots, \frac{2n-1}{2}$$

- Analysis is similar to before

- 3 cases:

- $p(z \text{ just right}): (H - L)/n$
- $p(z \text{ too low}): L/n$
- $p(z \text{ too high}): (n - H)/n$

Strategy Analysis (general case)

Choices
of z

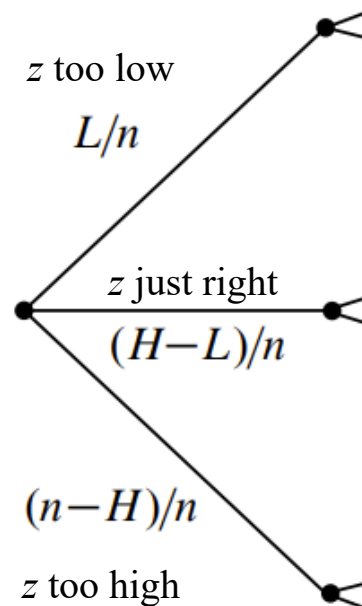
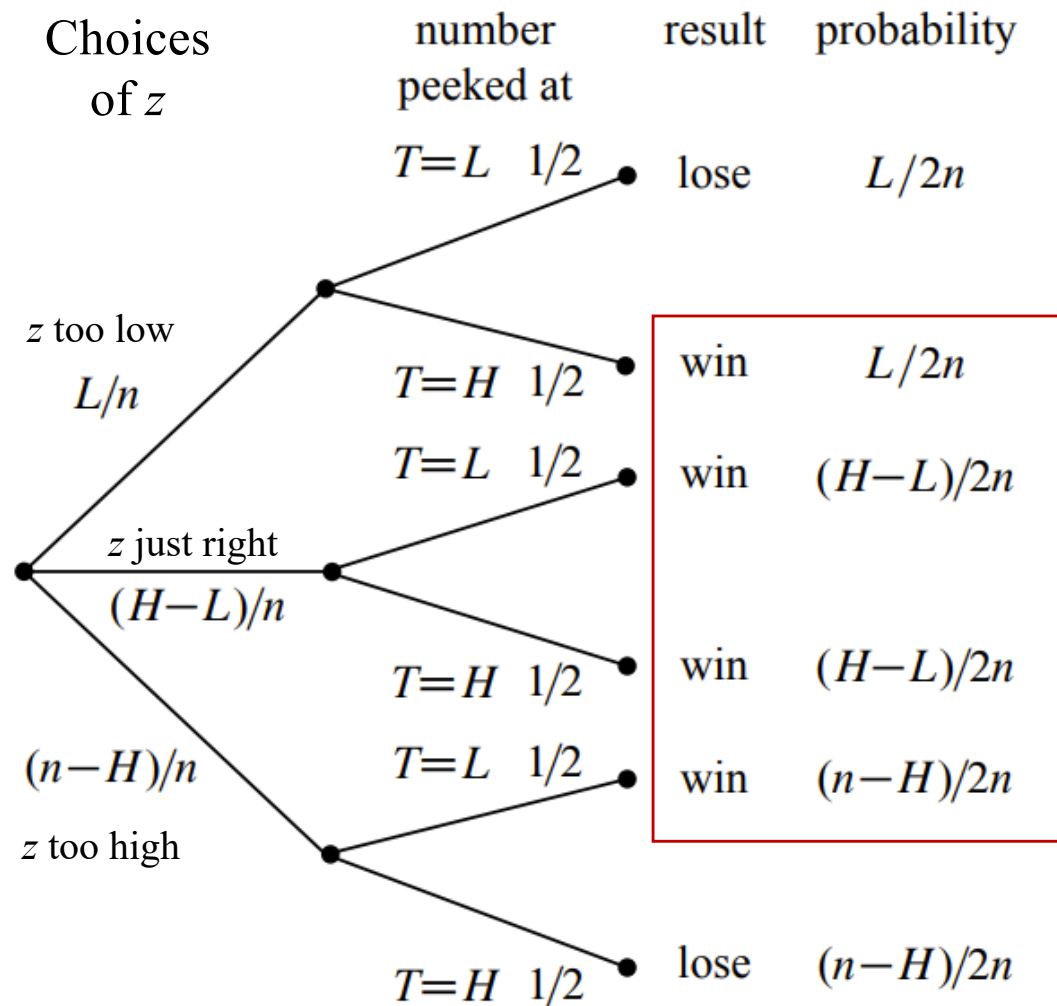


Figure 18.3 The tree diagram for the numbers game.

Strategy Analysis (general case)



Probability of win:

$$\begin{aligned}
 p(\text{win}) &= \frac{L}{2n} + 2 \frac{H-L}{2n} + \frac{n-H}{2n} \\
 &= \frac{1}{2} + \frac{H-L}{2n} \\
 &\geq \frac{1}{2} + \frac{1}{2n}
 \end{aligned}$$

Source: https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-042j-mathematics-for-computer-science-spring-2015/readings/MIT6_042JS15_Session31.pdf

<https://www.quantamagazine.org/solution-information-from-randomness-20150722/>

More rigorous:

<https://arxiv.org/pdf/1608.01899.pdf>

Figure 18.3 The tree diagram for the numbers game.

Questions?

<https://pollev.com/haroldsohsoo986>



2. Legal Reasoning

Setup:

- Blood found at a scene.
- Blood type present is type S.
- Type S blood is found in 1% of the population.

Prosecutor: *“there is a 1% chance the defendant would have blood type S if he was innocent. Thus, there is a 99% chance he is guilty!”*

Is the prosecutor correct? Justify your answer.

What is your decision?

- You are the Jury. Do you vote to convict the Defendant?
- Put in your vote onto polleverywhere



2. Legal Reasoning

The argument is wrong. Let:

- A = “*person has blood type S*”
- B = “*person is innocent*”

Prosecutor has quoted $p(A|B)$ (or $1 - p(A|B)$).
However, what is relevant is $p(B|A)$

$$p(\text{innocent}|S) = \frac{p(S|\text{innocent}) p(\text{innocent})}{p(S)}$$

This is known as the **prosecutor's fallacy**

Questions?

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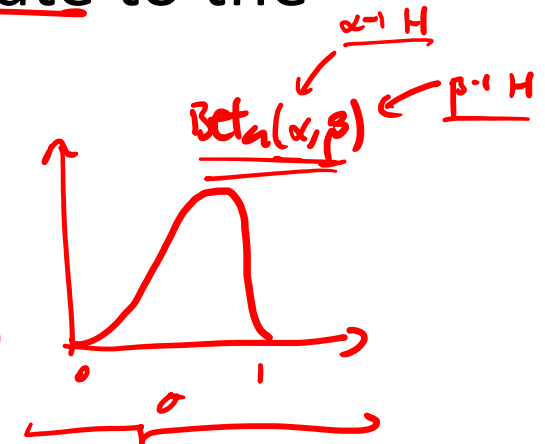
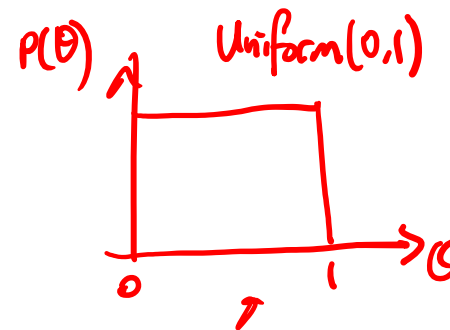


3. Conjugate Analysis

Show that the Beta Distribution is conjugate to the Binomial Distribution.

$$\underbrace{p(\theta|x)}_{\text{posterior}} = \frac{\underbrace{p(\theta)}_{\text{Prior}} \underbrace{p(x|\theta)}_{\text{Likelihood}}}{\underbrace{\int p(\theta) p(x|\theta) d\theta}_{\text{Marginal}}}$$

α, β



$$p(\pi) = \frac{1}{B(\alpha, \beta)} \pi^{\alpha-1} (1-\pi)^{\beta-1}$$

$$B(\alpha, \beta) = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha + \beta)}$$

$$\Gamma(x) = \int_0^{\infty} t^{x-1} e^{-t} dt$$

$$\Gamma(n) = (n-1)!$$

$$\binom{n}{x} = \frac{n!}{x! (n-x)!} = \frac{\Gamma(n+1)}{\Gamma(x+1) \Gamma(n-x+1)}$$

$$p(x|\pi) = \binom{n}{x} \pi^x (1-\pi)^{n-x}$$

$$p(\pi) = \frac{1}{B(\alpha, \beta)} \pi^{\alpha-1} (1-\pi)^{\beta-1} \quad \text{prior.}$$

$$p(x|\pi) = \binom{n}{x} \pi^x (1-\pi)^{n-x} \quad \text{likelihood}$$

$$\begin{aligned} \underline{\underline{p(\pi|x)}} &\propto p(x|\pi) p(\pi) = \binom{n}{x} \pi^x (1-\pi)^{n-x} \cdot \frac{1}{B(\alpha, \beta)} \pi^{\alpha-1} (1-\pi)^{\beta-1} \\ &= \underbrace{\binom{n}{x} \frac{1}{B(\alpha, \beta)}}_{\substack{\text{const} \\ \text{w.r.t } \pi}} \pi^{\underline{x+\alpha-1}} (1-\pi)^{\underline{n-x+\beta-1}} \\ &\quad \left[\begin{array}{l} \alpha' = \underline{x+\alpha} \\ \beta' = \underline{n-x+\beta} \end{array} \right] \end{aligned}$$

3. Conjugate Analysis $p(\mu | \underline{X})$ $p(\mu | x_i)$

Show that the Normal distribution is conjugate to the Normal distribution with unknown mean but known variance.

$$\begin{aligned}
 \frac{N(\mu | \underline{\mu}, \sigma^2)}{p(\mu | x_i)} &\propto \frac{p(\mu) p(x | \mu)}{N(x | \mu, \sigma^2)} \\
 &= \frac{1}{\sqrt{2\pi\sigma_0^2}} \exp\left(-\frac{(\mu - \mu_0)^2}{2\sigma_0^2}\right) \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right) \\
 &= \underbrace{\frac{1}{\sqrt{2\pi\sigma_0^2}} \cdot \frac{1}{\sqrt{2\pi\sigma^2}}}_{\text{const wrt } \mu} \exp\left(-\frac{1}{2} \left[\frac{(\mu - \mu_0)^2}{\sigma_0^2} + \frac{(x - \mu)^2}{\sigma^2} \right]\right) \approx c \exp\left(-\frac{1}{2} \frac{(\mu - \mu_1)^2}{\sigma_1^2}\right)
 \end{aligned}$$

Diagram: A graph showing two normal distributions. The wider one is labeled $N(\mu | \mu_0, \sigma_0^2)$ with arrows pointing to μ_0 and σ_0^2 (labeled 'Known'). The narrower one is labeled $N(x | \mu, \sigma^2)$ with an arrow pointing to σ^2 (labeled 'unknown').

$$\begin{aligned}
 \textcircled{A} \quad \frac{\mu^2 - \frac{2\mu\mu_0 + \mu_0^2}{\sigma_0^2} + \frac{x^2 - \frac{2x\mu + \mu^2}{\sigma^2}}{2}}{2} &= \left[\frac{1}{\sigma_0^2} + \frac{1}{\sigma^2} \right] \mu^2 - \left[\frac{2\mu_0}{\sigma_0^2} + \frac{2x}{\sigma^2} \right] \mu + \left[\frac{\mu_0^2}{\sigma_0^2} + \frac{x^2}{\sigma^2} \right] \\
 &= \frac{1}{\sigma_1^2} \mu^2 - \frac{1}{\sigma_1^2} \cdot 2\mu_1 \mu + \frac{1}{\sigma_1^2} \mu_1^2
 \end{aligned}$$

$$\underline{\left[\frac{1}{\sigma_1^2} + \frac{1}{\sigma^2} \right] \mu^2} - \underline{\left[\frac{2\mu_0}{\sigma_0^2} + \frac{2x}{\sigma^2} \right] \mu} + \text{const}$$

$$= \underline{\frac{1}{\sigma_1^2} \mu^2} - \underline{\frac{1}{\sigma_1^2} 2\mu_1 \mu} + \text{const}$$

$$X = \{x_1, \dots, x_n\}$$

\Downarrow

$$\frac{\mu_0}{\sigma_0^2} + \frac{x_1}{\sigma^2} = \frac{\mu_1}{\sigma_1^2}$$

$$\underline{\underline{\sigma_1^2 = \left[\frac{1}{\sigma_0^2} + \frac{1}{\sigma^2} \right]^{-1}}} \Rightarrow \underline{\underline{\mu_1 = \left[\frac{\mu_0}{\sigma_0^2} + \frac{x_1}{\sigma^2} \right] \left[\frac{1}{\sigma_0^2} + \frac{1}{\sigma^2} \right]^{-1}}}$$

$$\sigma_2^2 = \left[\frac{1}{\underline{\sigma_1^2}} + \frac{1}{\sigma^2} \right]^{-1} = \left[\frac{1}{\left[\frac{1}{\sigma_0^2} + \frac{1}{\sigma^2} \right]} + \frac{1}{\sigma^2} \right]^{-1} = \left[\frac{1}{\sigma_0^2} + \frac{1}{\sigma^2} + \frac{1}{\sigma^2} \right]^{-1} = \left[\frac{1}{\sigma_0^2} + \frac{2}{\sigma^2} \right]^{-1}$$

μ_2

4. Variance of a Sum

Show that

$$V[X + Y] = V[X] + V[Y] + 2\text{Cov}[X, Y] \quad \text{||} \leftarrow \text{show.}$$

where $\text{Cov}[X, Y] = E[XY] - E[X]E[Y]$ ✓

and $\underline{V[X]} = E[X^2] - E[X]^2$ ✓

$$\begin{aligned} V[X+Y] &= E[(X+Y)^2] - E[X+Y]^2 \\ &= E[X^2 + 2XY + Y^2] - E[X+Y]^2 \\ &= E[X^2] + 2E[XY] + E[Y^2] - (E[X] + E[Y])^2 \\ &= \underline{E[X^2]} + 2\underline{E[XY]} + \underline{E[Y^2]} - \underline{E[X]^2} - \underline{E[Y]^2} - 2\underline{E[X]E[Y]} \\ &= V[X] + V[Y] + 2\text{Cov}[X, Y]. \end{aligned}$$

Questions?

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Please Prepare for Next Week

- Complete Tutorial 1 (online).
- Watch the next batch of videos and do tutorial 2 (out tomorrow)