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Content Optimization with Multi-Arm Bandits & How to Title Presentations

Guest Talk CSCI 4830 Special Topics in Computer Science Data Driven Design 2021-03-18

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Me

- Call me Brad
- Director of Data Science at CU Boulder
- I use institutional data to inform decision-making
- Excited to chat with you today and beyond
- First time teaching on Zoom, if you have questions, please just interrupt.



Today's Discussion

- Quick Introduction
- Apartment Hunting in Springfield
- Learning vs. Earning in User Experience
- Frequentists and Bayesians
- Questions
- Open Ended Data Science Questions



Finding an Apartment in Springfield





Constraints





























Question: When Do You Stop Looking and Put Down a Deposit?





The challenge:

- If you deposit too early, you may miss out on a better apartment
- If you keep looking, you'll never find an apartment





Some Knowledge Here:



This is a well-known computer science problem called the **explore-exploit tradeoff**



Computers can scan through nearly limitless choices. How do we maximize the expected value from a selection?



Multi-Arm Bandits (MAB)



- Multi-arm bandits are algorithms that seek to maximize an expected payout
- They are intended for the most rapid move toward the "exploit" decision
- There are a variety of strategies for algorithms to switch from explore to exploit





Which machine has highest expected reward?





























THIS IS $EPSILON(\in)$













Let's exploit that knowledge and win some money!





Source Code: https://gist.github.com/robinvanemden/30969b48a44c2742a18ae148 61793741#file-epsilon_greedy_animation-r









Are there other ways I could explore then exploit?



Multi-Arm Bandit Solutions

- Changes to Explore/Exploit Ratio (Epsilon)
 - Do all exploring first
 - Gradually reduce the explore (Epsilon) rate over time
- Changes to Bandit Selection
 - Contextual bandits that choose the "correct" machine based on gambler's information
- Different Algorithmic Approach
 - Thompson Sampling (Bayesian)



An Oversimplified Description of Statistical Approaches

Frequentist

Assumes that with a large enough sample size, the point estimate (usually mean) gives the best guess at the whole population

Bayesian

 Assumes that the "right" answer is actually a range of probabilities from a distribution which can change and be updated





Source Code: https://gist.github.com/robinvanemden/30969b48a44c2742a18ae148 61793741#file-thompson_sampling_animation-r



Experiments in Web Design





Which content has highest expected engagement?



A/B Split Test vs. Multi-Arm Bandit



Develop a randomized statistical test to understand which among multiple options is "significantly" better

Pro: You get a definitive answer and your organization learns

Con: It takes time and requires a robust sample size

Deploy an algorithm that programmatically selects the current "winner" Pro: Your users are getting the benefit of the current, "best" version and the highest engagement rate

Con: The results might not provide a clear organizational answer



Key Takeaways



Randomized experimentation is critical. You should deploy as many *useful* experiments as your organization can effectively absorb

There are multiple approaches, and the main question is "Are you trying to *understand* or to *maximize*?"



Understanding the mechanics of these methods is helpful, but you don't need to be a statistician to use Google or Optimizely





Questions



Data Science Questions



Sources

Zillow

Simpsons images are protected by copyright and used here under fair use/educational guidelines

Interactive Bandits: <u>https://pavlov.tech/2019/03/02/animated-multi-armed-bandit-policies/</u>

Original R Code for Interactive Bandits: https://gist.github.com/robinvanemden/30969b48a44c2742a18ae14861793741

Updated R Code for Interactive Bandits: <u>https://gist.github.com/bradweiner/96593a8e7a34c03c0a92db33fdc64f75</u>

Algorithms to Live By Christian, B., & Griffiths, T. (2016). Algorithms to live by: The computer science of human decisions.

* Note that the interactive bandits in lecture were slightly repurposed, but otherwise copied from Github gist from Robin van Emden



Stable Location

This presentation can be found at:

https://bradweiner.info/files/mab_presentation.pdf

Bandit Simulations:

https://bradweiner.github.io/multi arm bandit presentation/

Github Repo: https://github.com/bradweiner/multi arm bandit presentati on



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