

Bandits

CISC 7404 - Decision Making

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Let us review some notation I will use in the course

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If you ever get confused, come back to these slides

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Vectors

$$oldsymbol{x} = egin{bmatrix} x_1 \ x_2 \ dots \ x_n \end{bmatrix}$$

Matrix

$$m{X} = egin{bmatrix} x_{1,1} & x_{1,2} & \dots & x_{1,n} \\ x_{2,1} & x_{2,2} & \dots & x_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{m,1} & x_{m,2} & \dots & x_{m,n} \end{bmatrix}$$

We will represent vectors or matrices of **tensors**

Vector of tensors

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Each x_i could be a vector, matrix, 3x3 tensor, etc

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Same for matrices

Matrix of tensors

$$m{X} = egin{bmatrix} m{x}_{1,1} & m{x}_{1,2} & ... & m{x}_{1,n} \ m{x}_{2,1} & m{x}_{2,2} & ... & m{x}_{2,n} \ dots & dots & dots \ m{x}_{m,1} & m{x}_{m,2} & ... & m{x}_{m,n} \end{bmatrix}$$

Question: What is the difference between the following?

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Capital letters will often refer to **sets**

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$$X = \{1, 2, 3, 4\}$$

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We will represent important sets with blackboard font

 \mathbb{R}

Set of all real numbers

$$\{1, 2.03, \pi, \ldots\}$$

 \mathbb{Z}

Set of all integers

$$\{-2, -1, 0, 1, 2, \ldots\}$$

 \mathbb{Z}_{+}

Set of all **positive** integers

$$\{1, 2, ...\}$$

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$$\max_{x} f(x)$$

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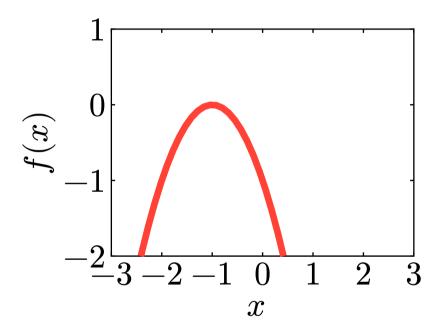
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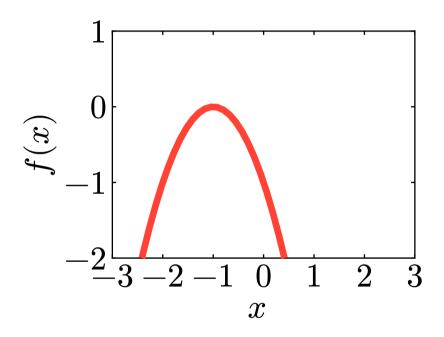
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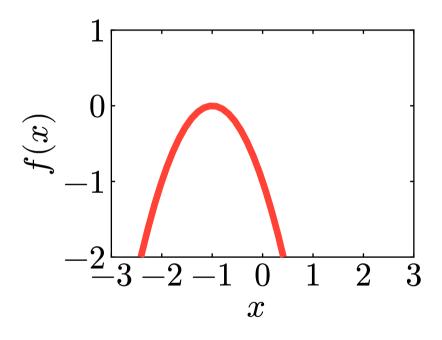
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Set of all boolean vectors of length n

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A bandit steals your money

Here is the bandit we will focus on in this course

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This is a **one-armed** bandit





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Let us see if we can make money playing this game

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But we can combine them to find out

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$$-10 \cdot \frac{199}{200} + 1000 \cdot \frac{1}{200} = -4.95$$

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As we play the game more and more, we converge to the expectation

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Answer: Do not play! If you must, play as little as possible

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After playing enough, the gambler can approximate the expectation!

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If $\mathbb{E}[\mathcal{X}] > 0$ you should gamble

If $\mathbb{E}[\mathcal{X}] < 0$ you should not gamble

We will consider a more interesting problem

You arrive at the Londoner with 1000 MOP and want to win money

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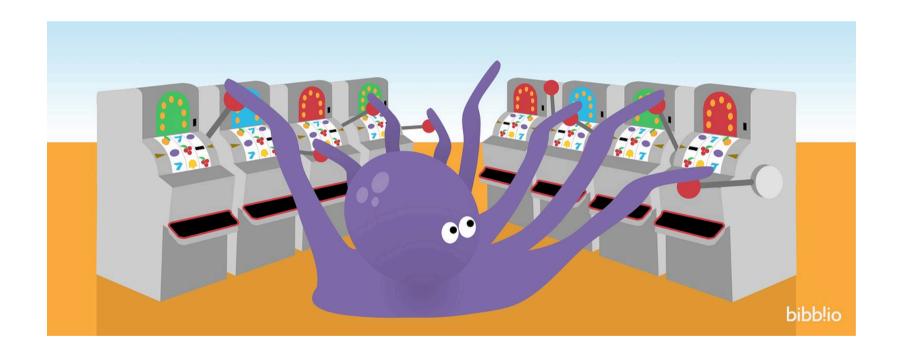
You arrive at the Londoner with 1000 MOP and want to win money



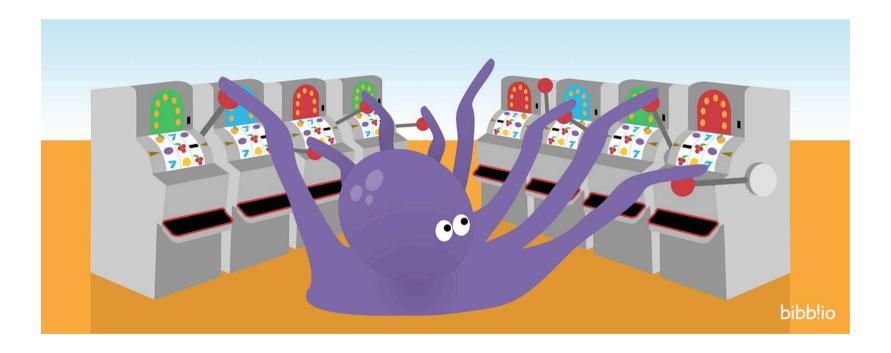
Question: Which machine do you play?

We call this the **multi-armed bandit** problem

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You don't know the expected value of each arm. Which should you pull?

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Medicine A



Medicine B



Medicine C

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Medicine B



Medicine C

We can find the best medicine while healing the most people

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Dog videos



Gaming videos



Study videos

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You are the bandit!



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You are the bandit!

You like a specific type of video, but YouTube does not know what it is

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Study videos

You are the bandit!

You like a specific type of video, but YouTube does not know what it is

YouTube tries to find your favorite video category

Problem: We have *k* bandits, and each bandit is a random variable

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Question: How should we approach this problem?

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- Pick a to approximate bandits
- Pick a to make the most money

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$$\operatorname*{arg\ max}_{a \in 1 \dots k} \mathbb{E}[\mathcal{X}_a]$$

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Use our estimates to make money

It is important you understand this! Any questions?

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Question: How can we achieve both goals at once?

Answer: Sometimes choose a to explore, sometimes choose a to exploit

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if u < 0.5 then $a \sim \operatorname{uniform}(\{1...k\})$

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Answer:

- $\varepsilon \approx 1$ when we trust our estimates $\mathbb{E}[\mathcal{X}]$
- $\varepsilon \approx 0$ when we do not trust our estimates

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- Sometimes it suggests study videos

Let us code some multiarmed bandits!

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