

Introduction to Discrete Math

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- 1 -

Global Frontier College

- Mathematical Thinking
 - Convincing Arguments, Find Example, Recursion & Induction, Logic, Invariants
- Probability & Combinatorics
 - Basic Counting, Binomial Coeff, Advanced Counting, Probability, **Random Variables**

Probability & Combinatronics – Random Variables

MARKOV INEQUALITY

- **From Expectation to Probability**
- Markov Inequality
- Application to Algorithms

Expectation vs Probability

- We have introduced random variables and their expected values
- We will now see how these notions can help in studies of probabilities of events

Lottery Budget

Problem

A lottery ticket costs 10,000 Won. 40% of the lottery budget goes to prizes. Show that the chances to win 500,000 won or more is $< 1\%$.

- We will use proof by contradiction

Lottery Budget

Problem

A lottery ticket costs 10,000 Won. 40% of the lottery budget goes to prizes. Show that the chances to win 500,000 won or more is < 1%.

- We will use proof by contradiction
- Assume the contrary: the probability to win 500,000 won or more is at least 0.01 $= \geq 0.01$ (1%)
- Denote number of tickets sold as n
- Then the budget of the lottery is $10n$ won
- Therefore, $10n \times 0.4 = 4n$ won are spent on prizes

Handwritten notes:
 $10,000$ (ticket cost)
 $1,000,000$ (lottery budget)
 n (number of tickets sold)
 $10n$ (total budget)
 $4n$ (amount spent on prizes)

Lottery Budget

Problem

A lottery ticket costs 10,000 Won. 40% of the lottery budget goes to prizes. Show that the chances to win 500,000 won or more is $< 1\%$.

- By our assumption, at least $\frac{n}{100}$ tickets win at least 500,000 won
- In total, these tickets win $\frac{n}{100} \times 500 = 5n$ won
- This exceeds the total prize budget of $4n$! ✗
- We arrived into contradiction and the problem is solved

- From Expectation to Probability
- **Markov Inequality**
- Application to Algorithms

Markov's Inequality

Suppose that f is a non-negative random variable. Then for any number $a > 0$, we have

$$\Pr\left[f \geq \frac{Ef}{a}\right]$$

- The inequality allows to use expected value to bound probability of certain events
- For the proof, it is convenient to rewrite the inequality:

$$a \times \Pr[f \geq a] \leq Ef$$

Markov's Inequality

- We need to prove the inequality $\alpha \times \Pr[f \geq \alpha] \leq Ef$
- Consider the following random variable g on the same probability space: for an outcome such that $f = a_i$ we let $\underline{g} = \underline{a}$ on this outcome if $\underline{a} \leq a_i$ and $\underline{g} = 0$ otherwise; that is

$$g = \begin{cases} a & \text{if } f = a_i \geq a \\ 0 & \text{if } f = a_i < a \end{cases}$$

- g is less than or equal to f on each outcome
- So, the average value Eg of g is less than or equal to the average value Ef of f

$$Eg \leq Ef$$

Markov's Inequality

$$g = \begin{cases} a & \text{if } f = a_i \geq a \\ 0 & \text{if } f = a_i < a \end{cases}$$

- What is the expectation of g ?
 - g has only ~~one nonzero value~~
- So Eg is this value multiplied by the sum of probabilities of all outcomes for this value
 - But these outcomes are exactly the outcomes that form event " $f \geq a$ "
- Thus the sum of their probabilities is equal to $\Pr[f \geq a]$
- Thus $Eg = a \times \Pr[f \geq a]$

Markov's Inequality

$$g = \begin{cases} a & \text{if } f = a_i \geq a \\ 0 & \text{if } f = a_i < a \end{cases}$$

- Finally we have, $Eg \leq Ef$.

- And $Eg = a \times \Pr[f \geq a]$

- So

$$Ef \geq Eg = a \times \Pr[f \geq a]$$

- We have Markov's Inequality

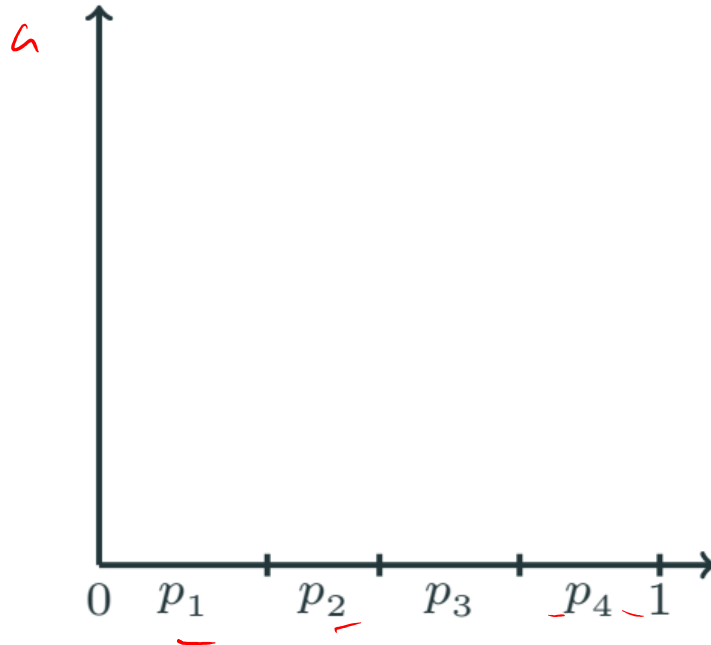
$$a \geq b$$

$$b \leq a$$

Geometric Interpretation

$$Ef \geq Eg = a \times \Pr[f \geq a]$$

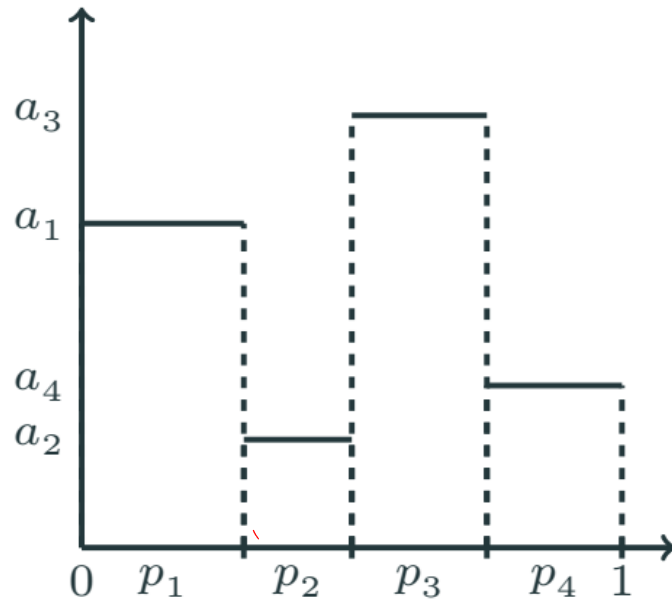
- Suppose f has values a_1, a_2, a_3, a_4 with probabilities p_1, p_2, p_3, p_4



Geometric Interpretation

$$Ef \geq Eg = a \times \Pr[f \geq a]$$

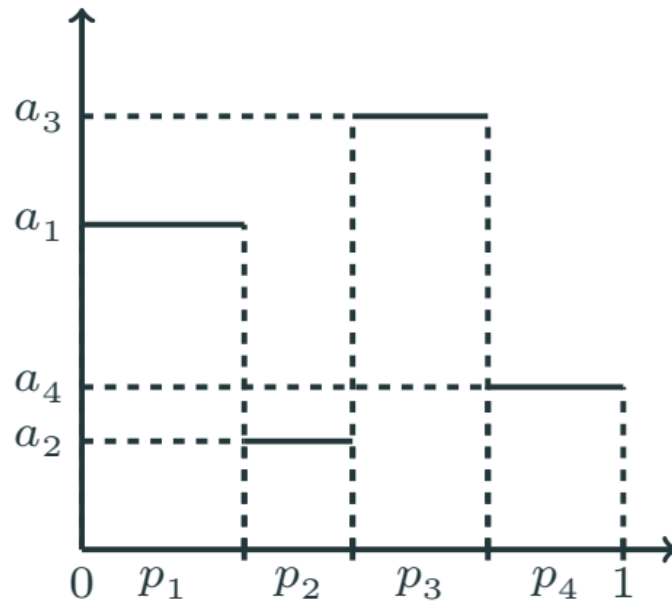
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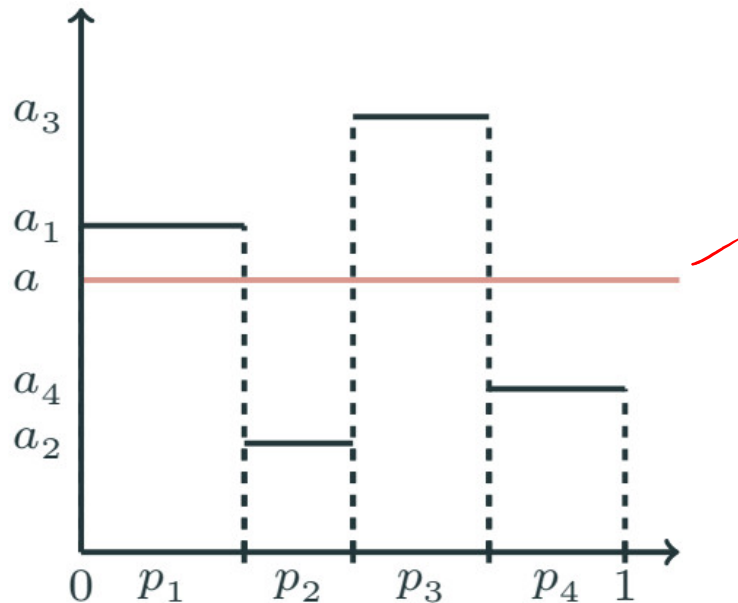
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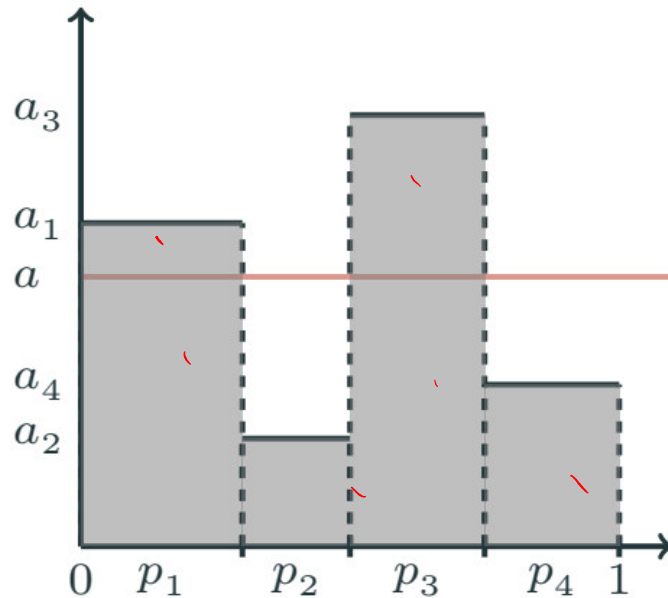
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Geometric Interpretation

$$Ef \geq Eg = a \times \Pr[f \geq a]$$

- Suppose f has values a_1, a_2, a_3, a_4 with probabilities p_1, p_2, p_3, p_4



- Ef is gray region area

$$Ef \geq Eg = a \times \Pr[f \geq a]$$

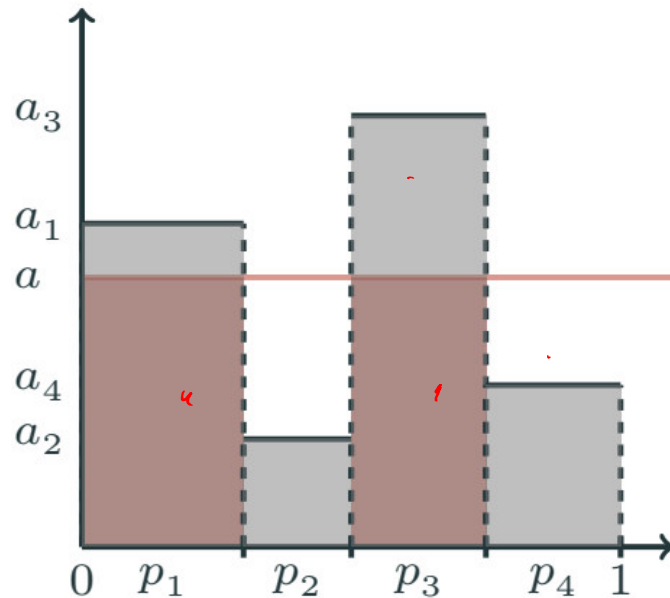
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- Ef is gray region area
- $a \times \Pr[f \geq a]$ is reddish region area

Geometric Interpretation

$$\left\{ \{Ef \geq Eg\} = a \times \Pr[f \geq a] \right\}$$

- Suppose f has values a_1, a_2, a_3, a_4 with probabilities p_1, p_2, p_3, p_4



- Ef is gray region area
- $a \times \Pr[f \geq a]$ is reddish region area
- Gray region is large and inequality follows

- From Expectation to Probability
- Markov Inequality
- **Application to Algorithms**

Application to Algorithms

Problem

Suppose there is a randomized algorithm that runs on average, assume n^2 , where n is the size of input. The algorithm outputs the correct answer. Construct another randomized algorithm that always stops in time cn^2 for some constant c and makes a mistake with probability of at most 10^{-3} .

- We will use **Markov's Inequality**

Application to Algorithms

- Running time of algorithm is a random variable, set it as f
- Recall that $Ef = n^2$
- Here is a new algorithm:
 - Run original algorithm for $10^3 n^2$ steps
 - If it stops, we also stop
 - If not, stop and output, for example “0”

Application to Algorithms

Claim

The probability that the original algorithm does not stop after $10^3 n^2$ number of steps is at most 10^{-3}

- Indeed, the probability is $\Pr[f \geq 10^3 n^2]$
- By Markov's Inequality, it is bounded by

$$\underline{\Pr[f \geq 10^3 n^2]} \leq \frac{Ef}{10^3 n_2} = \frac{\cancel{n}^2}{10^3 \cancel{n}^2} = \underline{10^{-3}}$$

Conclusion

- We studied **random variables**
- Allowed us to study the **quantitative aspects** of randomness
- Allowed us to apply many **analytical tools** to study probability
- **Expected values** is one of the main characteristics of a random variable
- On one side, **expectation** bears a **lot of information** of a random variable
- On the other side, expectation has very convenient **mathematical properties**

Thank you.