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Outline

From Expectation to Probability

Markov's Inequality

Application to Algorithms

Expectation vs. Probability

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Expectation vs. Probability

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- We will see now how these notions can help in studies of probabilities of events

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- $10n \times 0.4 = 4n$ dollars are spent on the prizes

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- This exceeds the total prize budget of 4n!
- We arrived into contradiction and the problem is solved

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Markov's Inequality

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

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

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

$$\mathsf{E} g \leq \mathsf{E} f$$

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- Thus $Eg = a \times Pr[f \ge a]$

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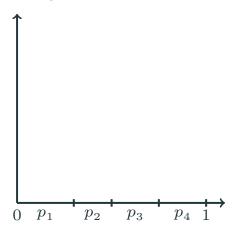
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We have shown Markov's inequality

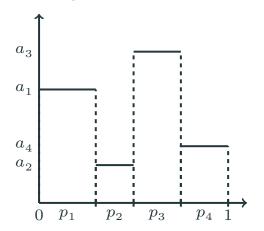
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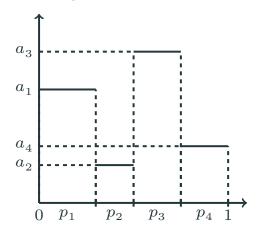
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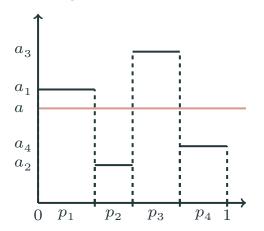
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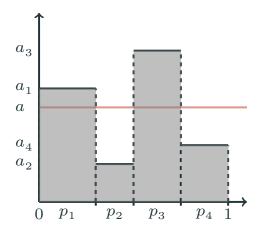
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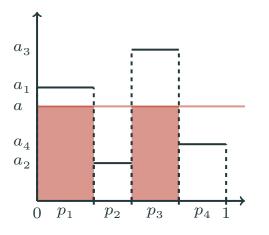


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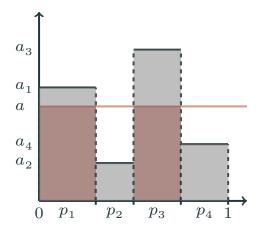


- Ef is the area of the gray region
- $a \times \Pr[f \ge a]$ is the area of a red region

 $\mathsf{E} f \geq a \times \Pr[f \geq a]$

Suppose f obtains values a_1, a_2, a_3, a_4 with probabilities $% \left({a_1} \right) = a_1 + a_2 + a_3 + a_4 + a_4 + a_5 +$

 p_1, p_2, p_3, p_4



- Ef is the area of the gray region
- $a \times \Pr[f \ge a]$ is the area of a red region
- The gray region is large and the inequality follows

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Suppose there is a randomized algorithm that runs on average in time, say, 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 at most 10^{-3}

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· We will apply Markov inequality

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- If it stops, we also stop
- If not, stop and output, say, 0

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

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- Indeed, this probability is $\Pr[f \ge 10^3 n^2]$
- By Markov's inequality it is bounded by

$$\Pr[f \ge 10^3 n^2] \le \frac{\mathsf{E}f}{10^3 n^2} = \frac{n^2}{10^3 n^2} = 10^{-3}$$

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- Allow us to apply many analytic tools to study probability
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- On one side, expectation bears a lot of information of a random variable
- On the other side, expectation has very convenient mathematical properties