

Session 83: Estimating Deviations

- Markov's Inequality
- Chebyshev's Inequality
- Examples

Markov's Inequality

Let X be a non-zero and non-negative random variable:

$$\exists s X(s) > 0 \wedge \forall s X(s) \geq 0$$

Let $p(X \geq a)$ denote the probability that the variable attains a value larger than a

$$\text{Then } \forall M > 0 \ p(X \geq M \cdot E(X)) \leq \frac{1}{M}$$

Denote $A = \{s \in S \mid X(s) \geq a\}$

Then $P(X \geq a) = P(A)$

Example

Rolling a dice, where rolling a 6 is considered as success (Bernoulli trial)

When performing 6 trials the expectation value is to obtain 1 success.

Estimate the probability of having at least 3 successes: (in 6 trials)

$$p(X \geq 3) = p(X \geq 3E(X)) \leq \frac{1}{3} \quad (M = 3)$$

Proof :

$$\frac{1}{M} = \frac{E(X)}{M \cdot E(X)} = \frac{\sum_{x \in X(S)} x \cdot p(X=x)}{M \cdot E(X)} \stackrel{*}{\geq} \frac{\sum_{\substack{x \in X(S) \\ x \geq M \cdot E(X)}} x \cdot p(X=x)}{M \cdot E(X)}$$
$$\geq \frac{\sum_{\substack{x \in X(S) \\ x \geq M \cdot E(X)}} M \cdot E(X) p(X=x)}{M \cdot E(X)} =$$

$$\sum_{\substack{x \in X(S) \\ x \geq M \cdot E(X)}} p(X=x) = p(X \geq M \cdot E(X))$$

* here we use that X is non-negative : if not, for $M = \frac{1}{2}$

e.g. if $S = \{0, 1\}$ and $X(0) = X(1) = -\frac{1}{2}$, then $\sum_{x \in X(S)} x \cdot p(X=x) = -1$, $\sum_{\substack{x \in X(S) \\ x \geq -\frac{1}{2}}} x \cdot p(X=x) = 0$

$$p(0) = p(1) = -\frac{1}{2}$$
$$E(X) = -1, M \cdot E(X) = -\frac{1}{2}$$

Chebyshev's Inequality

Let X be a random variable on a sample space S with probability function p . If r is a positive real number, then

$$p(|X(s) - E(X)| \geq r) \leq V(X)/r^2$$

Example

Rolling a dice, where rolling a 6 is considered as success (Bernoulli trial)

When performing 6 trials the expectation value is to obtain 1 success.
The variance is $npq = 6 \cdot 1/6 \cdot 5/6 = 5/6$

Estimate the probability of having at least 3 successes:

$$p(|X(s) - E(X)| \geq 2) \leq V(X)/2^2 = \frac{5}{6 \cdot 2^2} = \frac{5}{24} \approx \frac{1}{5}$$

Note: this is a smaller probability than $\frac{1}{3}$ obtained with the Markov inequality

Proof : Let A be the event $A = \{s \in S \mid |X(s) - E(X)| > r\}$

$$V(X) = \sum_{s \in S} (X(s) - E(X))^2 p(s)$$

$$= \sum_{s \in A} (X(s) - E(X))^2 p(s) + \sum_{s \in S \setminus A} (X(s) - E(X))^2 p(s)$$

$$\geq \sum_{s \in A} \underbrace{(X(s) - E(X))^2}_{\geq r} p(s) \geq \sum_{s \in A} r^2 p(s) = r^2 p(|X(s) - E(X)| \geq r)$$

Note : $p(|X(s) - E(X)| \geq r) = \sum_{s \in S} p(s) - \sum_{s \in A} p(s) = P(A)$

Example

We can also compute the exact probability of having 3 or more successes using the binomial distribution

$$b(3 : 6, 1/6) + b(4 : 6, 1/6) + b(5 : 6, 1/6) = \frac{6 \cdot 5 \cdot 4}{1 \cdot 2 \cdot 3} \frac{5^3}{6^6} + \frac{6 \cdot 5}{1 \cdot 2} \frac{5^2}{6^6} + \frac{6}{1} \frac{5}{6^6} \approx 0.06$$

Thus also the estimate using Chebyshev's Inequality was not very sharp

Summary

- Markov's Inequality
 - based on expectation value
- Chebyshev's Inequality
 - based on variance

Geometric Distribution

- assume you have a biased coin, probability of tail T is p

Question: What is the expected number of tosses, till a tail shows up

Sample Space: $S = \{T, HT, HHT, \dots\}$, it is infinite!

Probabilities: $p(T) = p$ T in first roll

$p(HT) = (1-p)p$ T in 2nd roll

$p(H^k T) = (1-p)^{k-1} \cdot p$

Define $X(s)$ as the random variable, that gives the number of rolls needed to obtain T

The $p(X = k) = (1-p)^{k-1} p$ geometric distribution

Properties of the geometric distribution , Expected Value

$$\begin{aligned} E(X) &= \sum_{r \in X(S)} r \cdot p(X=r) = \sum_{j=1}^{\infty} j \cdot p(X=j) = \\ &= \sum_{j=1}^{\infty} j \cdot (1-p)^{j-1} p = p \sum_{j=1}^{\infty} j \cdot (1-p)^{j-1} \end{aligned}$$

$$= p \cdot \frac{1}{(1 - (1-p))^2} = \frac{p}{p^2} = \frac{1}{p}$$

Note : $\frac{1}{(1-x)^2} = \sum_{j=0}^{\infty} C(2+j-1, j) \cdot x^j = \sum_{j=0}^{\infty} \binom{j+1}{j} x^j = \sum_{j=0}^{\infty} (j+1) x^j$

$$\begin{aligned} &= \sum_{j=1}^{\infty} j \cdot x^{j-1} \end{aligned}$$

Example: If $p = \frac{1}{4}$ it takes an expected 4 rolls to obtain tail.

Properties of the geometric distribution , Variance

$$V(X) = E(X^2) - E(X)^2 = E(X(X-1) + X) - E(X)^2 = E(X(X-1)) + E(X) - E(X)^2$$

$$\begin{aligned} E(X(X-1)) &= \sum_{j=1}^{\infty} j(j-1) p(X=j) = \sum_{j=1}^{\infty} j(j-1)(1-p)^{j-1} \cdot p = \\ &= \sum_{j=0}^{\infty} (j+1)j(1-p)^j \cdot p = 2p(1-p) \sum_{j=0}^{\infty} \frac{j(j+1)}{2}(1-p)^{j-1} \\ &= 2p(1-p) \frac{1}{p^3} = \frac{2(1-p)}{p^2} \end{aligned}$$

$$\begin{aligned} \text{Note: } \frac{1}{(1-x)^3} &= \sum_{j=0}^{\infty} C(3+j-1, j) x^j = \sum_{j=0}^{\infty} \binom{j+2}{j} x^j = \sum_{j=0}^{\infty} \frac{(j+2)(j+1)}{2} x^j \\ &= \sum_{j=1}^{\infty} \frac{(j+1)j}{2} x^{j-1} = \sum_{j=0}^{\infty} \frac{(j+1)j}{2} x^{j-1} \end{aligned}$$

$$V(X) = \frac{2(1-p)}{p^2} + \frac{1}{p} - \frac{1}{p^2} = \frac{2-2p+p-1}{p^2} = \frac{1-p}{p^2}$$

Using dice to generate random numbers

Assume you have a fair coin, and would like to generate random numbers from the set $\{0, \dots, n-1\}$

Simple Case if $n = 2^j$ for some $j \geq 1$

flip the coin j times, to generate the digits of a binary number.

Problem: if $n = 3$

- one toss gives 2 choices
- two tosses give 4 choices ?

Possible Solution

While True

toss coin 2 dimes

if result in $\{00, 01, 10\}$

return result

(i.e. ignore one of the possible outcomes)

Expected Number of tosses to obtain the random number :

$$P(X=1) = \frac{3}{4}$$

$$P(X=2) = \frac{1}{4} \cdot \frac{3}{4}$$

$$P(X=k) = \left(\frac{1}{4}\right)^{k-1} \cdot \frac{3}{4} \Rightarrow \text{geometric distribution with } p = \frac{3}{4}$$

It is not guaranteed that the algorithm terminates !

What is the expected number of tosses?

$X(s)$: number of attempts

$$P(X = k) = (1-p)^{k-1} p$$

each attempt requires 2 coin tosses : $T(s) = 2X(s)$

$$E(T) = E(2 \cdot X) = 2 E(X) = 2 \cdot \frac{1}{p} = 2 \cdot \frac{4}{3} = \frac{8}{3}$$

On average we need $\frac{8}{3}$ tosses!

How likely is it to have many losses? : analyzing variance

$$V(X) = \frac{1-p}{p^2}, p = \frac{3}{4}, \text{ therefore } V(X) = \frac{\frac{1}{4}}{\frac{9}{16}} = \frac{4}{9}$$

We use Chebychev's inequality to estimate how likely are large deviations.

$$P(|X(s) - E(X)| \geq r) \leq \frac{V(X)}{r^2}, E(X) = \frac{4}{3}$$

$$\text{Now } X(s) \geq k \Leftrightarrow X(s) - \frac{4}{3} \geq k - \frac{4}{3} \Leftrightarrow |X(s) - E(X)| \geq k - \frac{4}{3}$$

Therefore

$$\begin{aligned} P(X(s) \geq k) &= P\left(|X(s) - E(X)| \geq k - \frac{4}{3}\right) \leq \frac{\frac{4}{9}}{\left(k - \frac{4}{3}\right)^2} \\ &= \frac{4}{(3k-4)^2} \in \Theta\left(\frac{1}{k^2}\right) \end{aligned}$$

So, the probability is in: $\Theta\left(\frac{1}{k^2}\right)$