

# Probabilistic Tools

William Schultz

September 16, 2022

See [1] for a reference on some probabilistic tools.

## Union Bound

Let  $E_1, \dots, E_n$  be arbitrary events in some probability space. Then,

$$Pr \left[ \bigcup_{i=1}^n E_i \right] \leq \sum_{i=1}^n Pr[E_i]$$

That is, for a set of events  $E_1, \dots, E_n$ , the probability of at least one event occurring is less than or equal to the sum of the probabilities of the individual events.

## Expectation and Variance

For a discrete random variable  $X$  taking values in some set  $\Omega \subseteq \mathbb{R}$ , its *expectation* is defined as

$$\mathbb{E}[X] = \sum_{\omega \in \Omega} \omega \cdot Pr[X = \omega]$$

and its *variance* is defined as

$$\text{Var}(X) = \mathbb{E}[(X - \mathbb{E}[X])^2]$$

That is, expectation is essentially a weighted sum of the values that the random variable can take on, where each value is weighted by the probability of that event occurring, and variance is essentially a measure of the average deviation of the variable from its expectation/mean.

## Markov's Inequality

*Markov's inequality* is an elementary large deviation bound valid for *all* non-negative random variables. Let  $X$  be a non-negative random variable with  $\mathbb{E}[X] > 0$ . Then, for all  $\lambda > 0$

$$Pr(X \geq \lambda \mathbb{E}[X]) \leq \frac{1}{\lambda}$$

That is, this establishes a bound, for some given parameter  $\lambda$ , on how likely a random variable is to be far away from its mean. For example, if  $\lambda = 10$ , then this means that the probability that  $X$  is greater than 10 times its mean is  $\leq \frac{1}{10}$ . Note that this bound doesn't take into account anything about the actual distribution (only about its mean), so it may serve as only a rough estimate.

## Chebyshev's Inequality

Let  $X$  be a random variable with  $\text{Var}(X) > 0$ , and where  $\sigma = \sqrt{\text{Var}(X)}$  is its standard deviation. Then, for all  $\lambda > 0$

$$Pr[|X - \mathbb{E}[X]| \geq \lambda \sigma] \leq \frac{1}{\lambda^2}$$

This bound tells us something about how far a random variable is from its expectation, in terms of the variance of the RV. That is, it puts a bound on the probability of how many ( $\lambda$ ) standard deviations ( $\sigma$ ) away from its mean  $X$  may be.

## Chernoff Bound

TODO.

## References

- [1] Benjamin Doerr. Probabilistic tools for the analysis of randomized optimization heuristics. In *Natural Computing Series*, pages 1–87. Springer International Publishing, nov 2019.