

Math 340: Lec 13: Variance, Weak LLN, Chebyshev/Markov)

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0.1 Variance of sum of iids

Remark. Recall that generally, $\text{Var}(\sum_{i=1}^n X_i) = \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{i < j} \text{Cov}(X_i, X_j)$

Proposition 1. If random variables X_1, \dots, X_n have equal variance,

$$\text{Var}(X_1 + \dots + X_n) = \sum_{i=1}^n \sigma^2 = n\sigma^2$$

and

$$\text{SD}(X_1 + \dots + X_n) = \sigma\sqrt{n}$$

Remark. The standard deviation of a sum of iid random variables grows more slowly than the sum of the variables

Example (Compute the variance of $X \sim \text{Bin}(n, p)$). We can express X as a sum of indicators. Each has the same variance, $p(1-p)$, so $\text{Var}(X) = np(1-p)$

0.2 Inequalities

Theorem 2 (Markov's inequality). Intuition: Extreme values of a random variable are unlikely, because we need to balance probability mass around the mean. For a positive random variable Y ,

$$\mathbb{P}(Y > t) \leq \frac{1}{t} \mathbb{E}[Y]$$

The tail probability of a positive random variable is bounded by its expected value.

Theorem 3 (Chebyshev's inequality). Intuition: If the $\text{Var}(X)$ is small, then X is unlikely to be far from its mean. The amount a random variable X with $\text{Var}(X) < \infty$ can vary from its mean, μ is bounded:

$$\mathbb{P}(|X - \mu| \geq t) \leq \frac{\text{Var}(x)}{t^2}, \forall t > 0$$

Remark. Markov's inequality leads to the weak law of large numbers

Theorem 4 (Weak law of large numbers). Let X be the sum of n iid random variables. When $\text{Var}(X) < \infty$, we can bound the probability that the average of the X_1, \dots, X_n deviates from the mean of the random variables:

$$\mathbb{P}\left(\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| < \varepsilon\right) \leq \frac{\sigma^2}{\varepsilon^2 n}$$

And even if $\text{Var}(X)$ is infinite, if $\mathbb{E}[X] < \infty$,

$$\lim_{n \rightarrow \infty} \mathbb{P}\left(\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| > \varepsilon\right) = 0$$