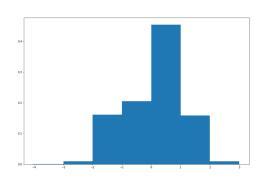
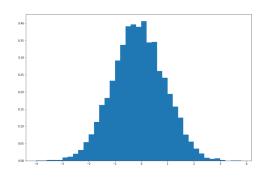
Quick Review

Central Limit Theorem: Let X,,... be i.i.d. RVs with finite expectation μ and variance σ^2 . Then

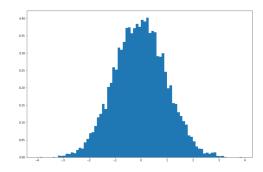
$$\frac{\sum_{i=1}^{n} X_{i} - n_{M}}{\sigma \sqrt{n}} \xrightarrow{\text{converges}} \mathcal{N}(0, 1)$$



Distribution of values for 10,000 trials of n = 10 coin flips



Distribution of values
for 10,000 trials of
n = 100 coin flips



Distribution of values for 10,000 trials of n = 10,000 coin flips

Sometimes we don't know the exact distribution of an RV, but we do have results of experiments

- We can estimate the parameters of the dist, using confidence intervals
- Let f(x) be the actual value (that you don't know) and $\overline{f}(x)$ be the empirical value (e.g. sample mean, sample variance, etc) that you calculate.
- Provide an interval $(\bar{f}(x) \epsilon, \bar{f}(x) + \epsilon)$ such that f(x) is in this interval with probability 1-6.
- 5 small => & big and vice versa.

1 Confidence Interval Introduction

We observe a random variable X which has mean μ and standard deviation $\sigma \in (0, \infty)$. Assume that the mean μ is unknown, but σ is known.

We would like to give a 95% confidence interval for the unknown mean μ . In other words, we want to give a random interval (a,b) (it is random because it depends on the random observation X) such that the probability that μ lies in (a,b) is at least 95%.

We will use a confidence interval of the form $(X - \varepsilon, X + \varepsilon)$, where $\varepsilon > 0$ is the width of the confidence interval. When ε is smaller, it means that the confidence interval is narrower, i.e., we are giving a more *precise* estimate of μ .

(a) Using Chebyshev's Inequality, calculate an upper bound on $\mathbb{P}\{|X - \mu| \ge \varepsilon\}$.

(b) Explain why $\mathbb{P}\{|X-\mu|<\varepsilon\}$ is the same as $\mathbb{P}\{\mu\in(X-\varepsilon,X+\varepsilon)\}$.

(c) Using the previous two parts, choose the width of the confidence interval ε to be large enough so that $\mathbb{P}\{\mu \in (X - \varepsilon, X + \varepsilon)\}$ is guaranteed to exceed 95%. [Note: Your confidence interval is allowed to depend on X, which is observed, and σ , which is known. Your confidence interval is not allowed to depend on μ , which is unknown.]

2 Poisson Confidence Interval

You collect n samples (n is a positive integer) X_1, \ldots, X_n , which are i.i.d. and known to be drawn from a Poisson distribution (with unknown mean). However, you have a bound on the mean: from a confidential source, you know that $\lambda \leq 2$. Find a $1 - \delta$ confidence interval ($\delta \in (0,1)$) for λ using Chebyshev's Inequality. (Hint: a good estimator for λ is the *sample mean* $\bar{X} := n^{-1} \sum_{i=1}^{n} X_i$)

3 Hypothesis testing

We would like to test the hypothesis claiming that a coin is fair, i.e. P(H) = P(T) = 0.5. To do this, we flip the coin n = 100 times. Let Y be the number of heads in n = 100 flips of the coin. We decide to reject the hypothesis if we observe that the number of heads is less than 50 - c or larger than 50 + c. However, we would like to avoid rejecting the hypothesis if it is true; we want to keep the probability of doing so less than 0.05. Please determine c. (Hints: use the central limit theorem to estimate the probability of rejecting the hypothesis given it is actually true. Table is provided in the appendix.)

4 Appendix

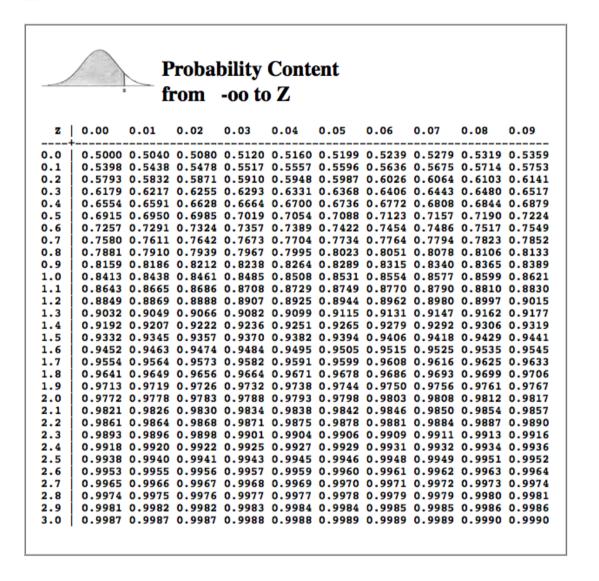


Table 1: Table of the Normal Distribution