Exam II Outline

1. Empirical probability

    a. The empirical probability of an event E is the long run proportion of times that the event occurs

    $$P(E) = \frac{\# \; of \; times \; event \; occurs}{\# \; of \; trials \; or \; experiments}$$

    b. Can be calculated in R using the *replicate* function to replicate an event (implemented in a function) such as flipping a coin.

2. Classical probability

    a. Formula: classical probability of an event E =

    $$P(E) = \frac{\# \; of \; ways \; in \; which \; E \; occurs}{\# \; of \; possible \; outcomes \; in \; the \; sample \; space}$$

    when all possible outcomes are equally likely

    b. Can be calculated in R using the *permutations* function (when order matters) from the library *gtools* or the *combinations* function (when order does not matter)*.*

3. Probability distributions

    a. The probability that a random variable X is less than the value k can be calculated based on classical probability and is

    $$P(X < k) = \frac{\# \; of \; oservations < k}{total \; \# \; of \; observations \; (sample \; size)}$$

    b. This probability is equivalent to the sum of histogram densities for the bars corresponding to $X < k$.

    c. This probability is equivalent to the area under the curve between $-\infty$ and k for a probability density function.

    d. For any probability density function, the area under the curve between points a and b is equal to $P(a < X < b)$.

4. The normal probability distribution

   a. $X \sim N(\mu, \sigma)$ if it is a unimodal, symmetric, bell-shaped distribution with mean μ and standard deviation σ

   b. Empirical rule: for a normal distribution, approximately 68%, 95% and 99% of observations are within 1, 2, and 3 standard deviations of the mean.

   c. The standard normal distribution is a normal distribution with $\mu = 0$ and $\sigma = 1$.

   d. *R* functions:

      i. *pnorm* –calculates probabilities of the form $P(X < k)$, equivalent to the area under the curve to the *left* of k

      ii. *dnorm* – calculates the probability density (e.g., if plotting the curve)

      iii. *qnorm* – calculates percentiles of the normal distribution

      iv. *rnorm* – random number generation

5. Sampling distribution of the sample mean

   a. The sample mean is a random variable!

   b. Central Limit Theorem: If a distribution X has mean μ and standard deviation σ then the sample distribution of the sample mean $\bar{X}_n$ from a sample of size n has mean μ and standard deviation $\sigma/\sqrt{n}$. Furthermore, the distribution of $\bar{X}_n$ is normal if X is normally distributed, and approximately normal for other distributions if $n > 30$. The larger the sample size, the closer the distribution is to normal.

6. Hypothesis testing based on a population proportion, a population mean, the difference between population means, or difference between two proportions.

   a. State the null and alternative hypotheses

   b. Specify the distribution of the sample statistic (i.e., a sample proportion or sample mean) under the null hypothesis.

c. Calculate/find the test statistic (using the *prop.test* or *t.test* functions), or manually using the appropriate formula

d. Find the *p*-value

    i. From *prop.test* or *t.test*

    ii. Based on a *z*- or *t*- test statistic and using the *pnorm* or *pt* function with appropriate degrees of freedom.

e. State the conclusion regarding the null hypothesis in the context of the problem, and justify your conclusion based on the *p*-value.

f. Interpretation of Type I or Type II errors