

5.2: Theoretical Models for the Distribution of Data

There are four important types of distributions that we will consider in this chapter: the uniform distribution, the binomial distribution, the Poisson distribution, and the normal, or Gaussian, distribution. In Chapter 3 and Chapter 4 we used the analysis of bags of M&Ms to explore ways to visualize data and to summarize data. Here we will use the same data set to explore the distribution of data.

Uniform Distribution

In a uniform distribution, all outcomes are equally probable. Suppose the population of M&Ms has a uniform distribution. If this is the case, then, with six colors, we expect each color to appear with a probability of $1/6$ or 16.7%. Figure 5.2.1 shows a comparison of the theoretical results if we draw 1699 M&Ms—the total number of M&Ms in our sample of 30 bags—from a population with a uniform distribution (on the left) to the actual distribution of the 1699 M&Ms in our sample (on the right). It seems unlikely that the population of M&Ms has a uniform distribution of colors!

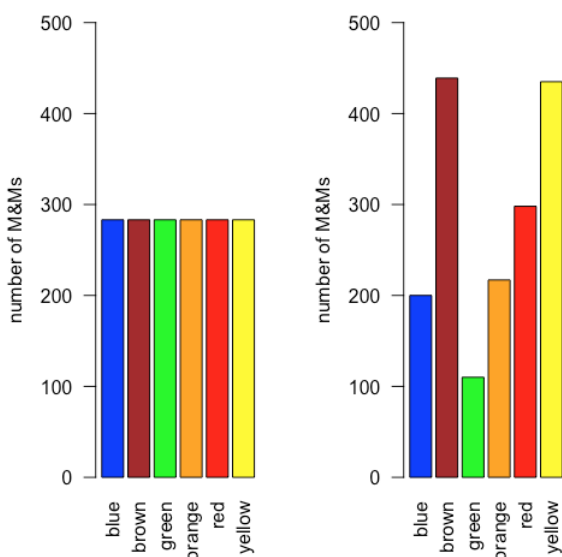


Figure 5.2.1: Comparison of (on the left) a uniform distribution of 1699 M&Ms with (on the right) the actual distribution from the sample in Table 3.1.1.

Binomial Distribution

A binomial distribution shows the probability of obtaining a particular result in a fixed number of trials, where the odds of that result happening in a single trial are known. Mathematically, a binomial distribution is defined by the equation

$$P(X, N) = \frac{N!}{X!(N-X)!} \times p^X \times (1-p)^{N-X}$$

where $P(X, N)$ is the probability that the event happens X times in N trials, and where p is the probability that the event happens in a single trial. The binomial distribution has a theoretical mean, μ , and a theoretical variance, σ^2 , of

$$\mu = Np \quad \sigma^2 = Np(1-p)$$

Figure 5.2.2 compares the expected binomial distribution for drawing 0, 1, 2, 3, 4, or 5 yellow M&Ms in the first five M&Ms—assuming that the probability of drawing a yellow M&M is $435/1699$, the ratio of the number of yellow M&Ms and the total number of M&Ms—to the actual distribution of results. The similarity between the theoretical and the actual results seems evident; in Chapter 6 we will consider ways to test this claim.

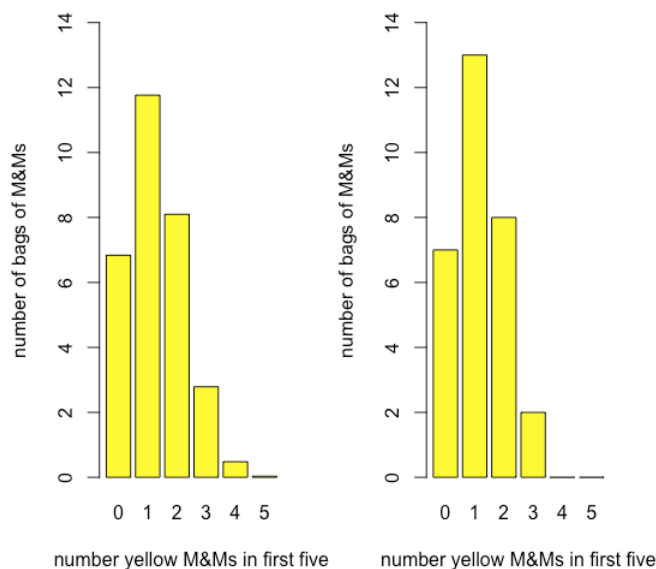


Figure 5.2.2: Comparison of (on the left) the theoretical binomial distribution of yellow M&Ms in the first five selected from a bag of M&Ms and (on the right) the actual distribution of M&Ms.

Poisson Distribution

The binomial distribution is useful if we wish to model the probability of finding a fixed number of yellow M&Ms in a sample of M&Ms of fixed size—such as the first five M&Ms that we draw from a bag—but not the probability of finding a fixed number of yellow M&Ms in a single bag because there is some variability in the total number of M&Ms per bag.

A Poisson distribution gives the probability that a given number of events will occur in a fixed interval in time or space if the event has a known average rate and if each new event is independent of the preceding event. Mathematically a Poisson distribution is defined by the equation

$$P(X, \lambda) = \frac{e^{-\lambda} \lambda^X}{X!}$$

where $P(X, \lambda)$ is the probability that an event happens X times given the event's average rate, λ . The Poisson distribution has a theoretical mean, μ , and a theoretical variance, σ^2 , that are each equal to λ .

The bar plot in Figure 5.2.3 shows the actual distribution of green M&Ms in 35 small bags of M&Ms (as reported by M. A. Xu-Friedman “Illustrating concepts of quantal analysis with an intuitive classroom model,” *Adv. Physiol. Educ.* **2013**, 37, 112–116). Superimposed on the bar plot is the theoretical Poisson distribution based on their reported average rate of 3.4 green M&Ms per bag. The similarity between the theoretical and the actual results seems evident; in Chapter 6 we will consider ways to test this claim.

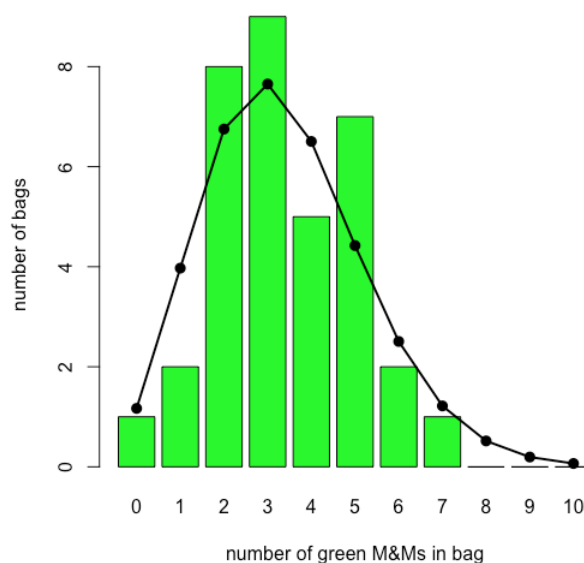


Figure 5.2.3: Comparison of a Poisson distribution for green M&Ms (dots and line) to experimental results (bars). The data are from M. A. Xu-Friedman, “Illustrating concepts of quantal analysis with an intuitive classroom model,” *Adv. Physiol. Educ.* **2013**, 37, 112–116.

Normal Distribution

A uniform distribution, a binomial distribution, and a Poisson distribution predict the probability of a discrete event, such as the probability of finding exactly two green M&Ms in the next bag of M&Ms that we open. Not all of the data we collect is discrete. The net weights of bags of M&Ms is an example of continuous data as the mass of an individual bag is not restricted to a discrete set of allowed values. In many cases we can model continuous data using a normal (or Gaussian) distribution, which gives the probability of obtaining a particular outcome, $P(x)$, from a population with a known mean, μ , and a known variance, σ^2 . Mathematically a normal distribution is defined by the equation

$$P(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/(2\sigma^2)}$$

Figure 5.2.4 shows the expected normal distribution for the net weights of our sample of 30 bags of M&Ms if we assume that their mean, \bar{X} , of 48.98 g and standard deviation, s , of 1.433 g are good predictors of the population’s mean, μ , and standard deviation, σ . Given the small sample of 30 bags, the agreement between the model and the data seems reasonable.

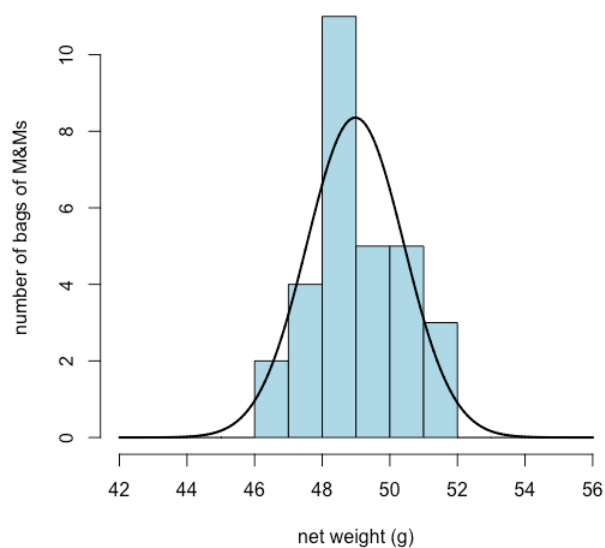


Figure 5.2.4: Comparison of a normal distribution for the net weights of M&Ms (line) to the experimental results (bars).

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