

Distributions



LECTURE 6

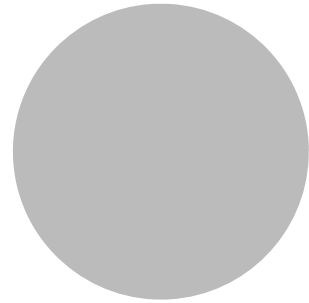
Objectives



- ▶ Define terms.
- ▶ Describe and distinguish between Poisson, binomial, and normal distributions.
- ▶ Describe in qualitative and quantitative terms a normal distribution.
- ▶ Describe the different types of deviations from normality (skew and kurtosis).
- ▶ Calculate proportions of a normal curve using Z-scores.

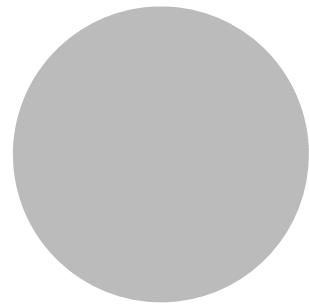
Objectives

- ▶ Explain the central limit theorem.
- ▶ Interpret the standard error of a sample and calculate it.



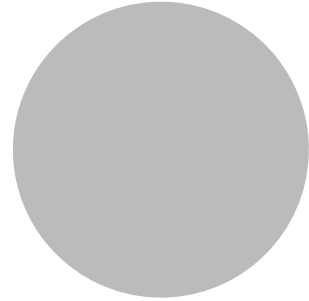
Frequency distributions

- ▶ Empirical distributions
- ▶ Conceptual distributions



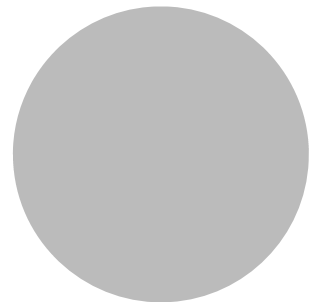
Binomial distribution

- ▶ Discrete probability distribution – counts
- ▶ There a number (n) of observations
 - ▶ a priori
- ▶ The observations are independent.
- ▶ An observation is classified as a “success” or a “failure.”
- ▶ The probability of a success, p , is the same for each observation.



Binomial distribution

- ▶ A binomial distribution has two parameters:
 - ▶ n – number of observations
 - ▶ p – probability of success



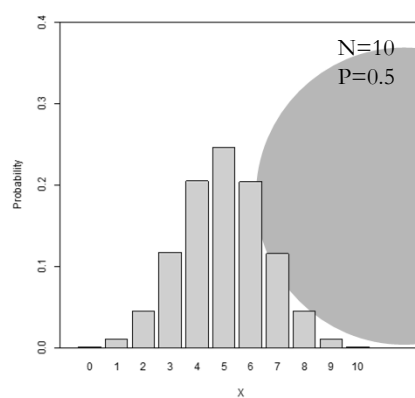
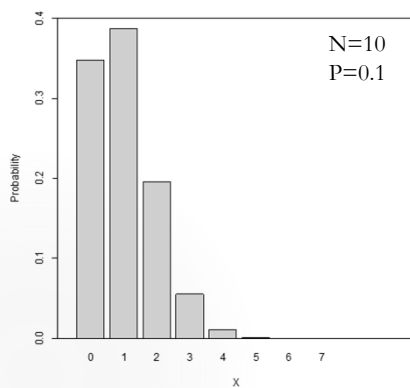
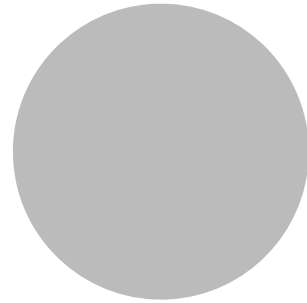
Binomial distribution

- ▶ Binomial coefficient

$$\binom{n}{X} = \frac{n!}{X!(n-X)!}$$

- ▶ Binomial probability

$$P(X) = \binom{n}{X} p^X (1-p)^{n-X}$$



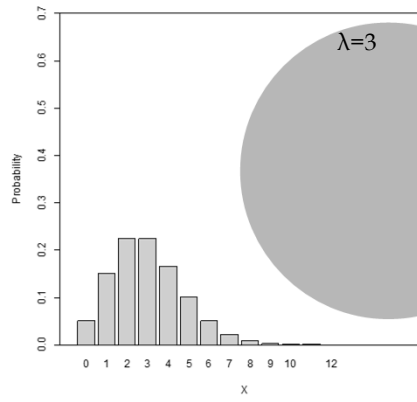
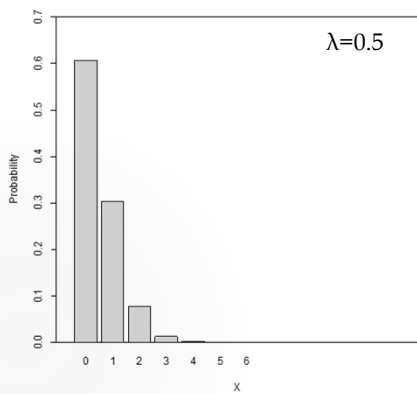
Poisson distribution

- ▶ Discrete probability distribution – counts
- ▶ Describes the count of X occurrences of a defined event in fixed, finite intervals of time or space when:
 - ▶ Occurrences are independent.
 - ▶ The probability of an occurrence is the same over all intervals.

Poisson distribution

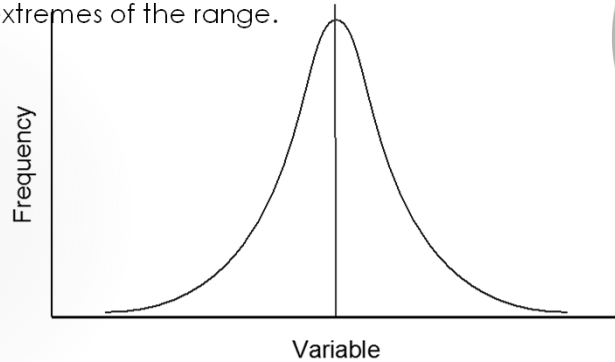
- ▶ Poisson Probability
 - ▶ μ -- Mean number of occurrences per interval
 - ▶ X – Number of occurrences in an interval

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



Bell-shaped curves

- A distribution of interval or ratio scale data observed to have most of its values near the mean with progressively fewer observations near the extremes of the range.



Normal distribution

- ▶ Not all bell-shaped curves are normal distributions
- ▶ Continuous probability distribution.
- ▶ A normal distribution is one in which the curve at X_i is expressed by the relation :

$$Y_i = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(X_i-\mu)^2}{2\sigma^2}}$$

Normal distribution

$$Y_i = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(X_i-\mu)^2}{2\sigma^2}}$$

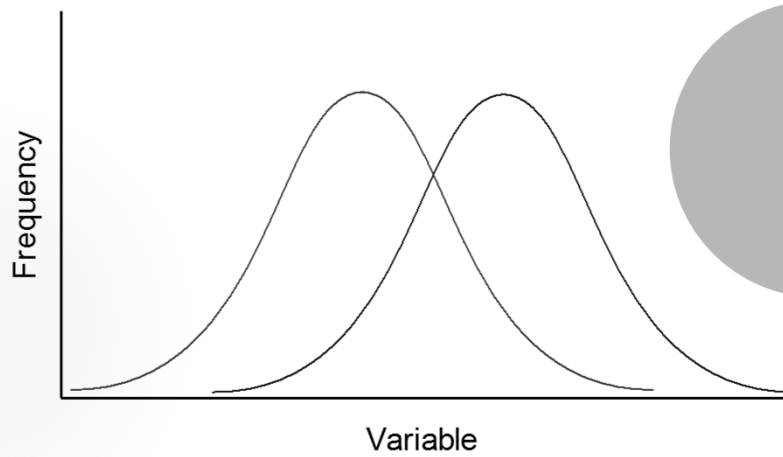
- ▶ Notice that the normal curve is completely described by two parameters: the mean and the standard deviation.

Normal distribution

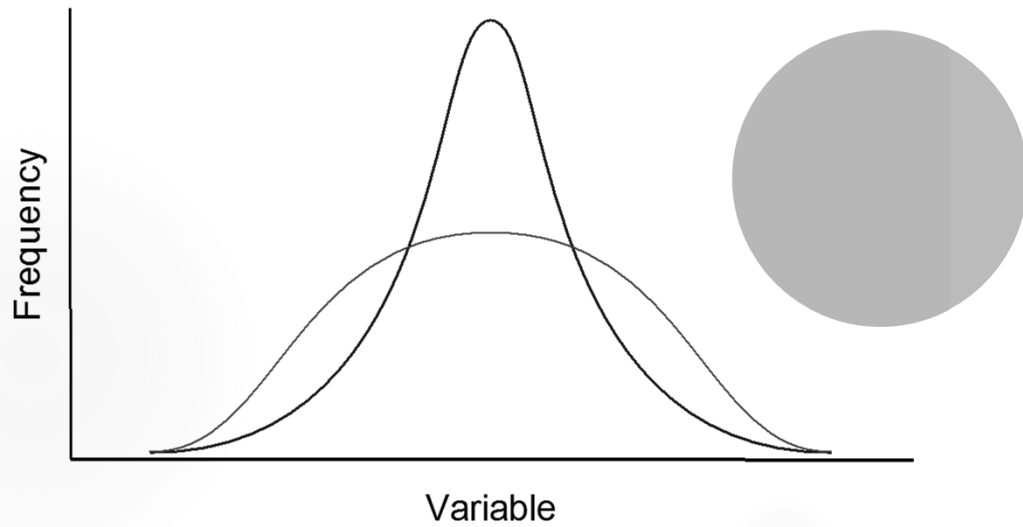
- ▶ The position of the curve is determined by the mean.
- ▶ The shape, to a large degree, is determined by the standard deviation.



Normal distribution



Normal distribution



Normal distribution

- ▶ The “standard normal curve”
 - ▶ $\mu = 0$
 - ▶ $\sigma = 1$
- ▶ Equation for normal distributions simplifies to:

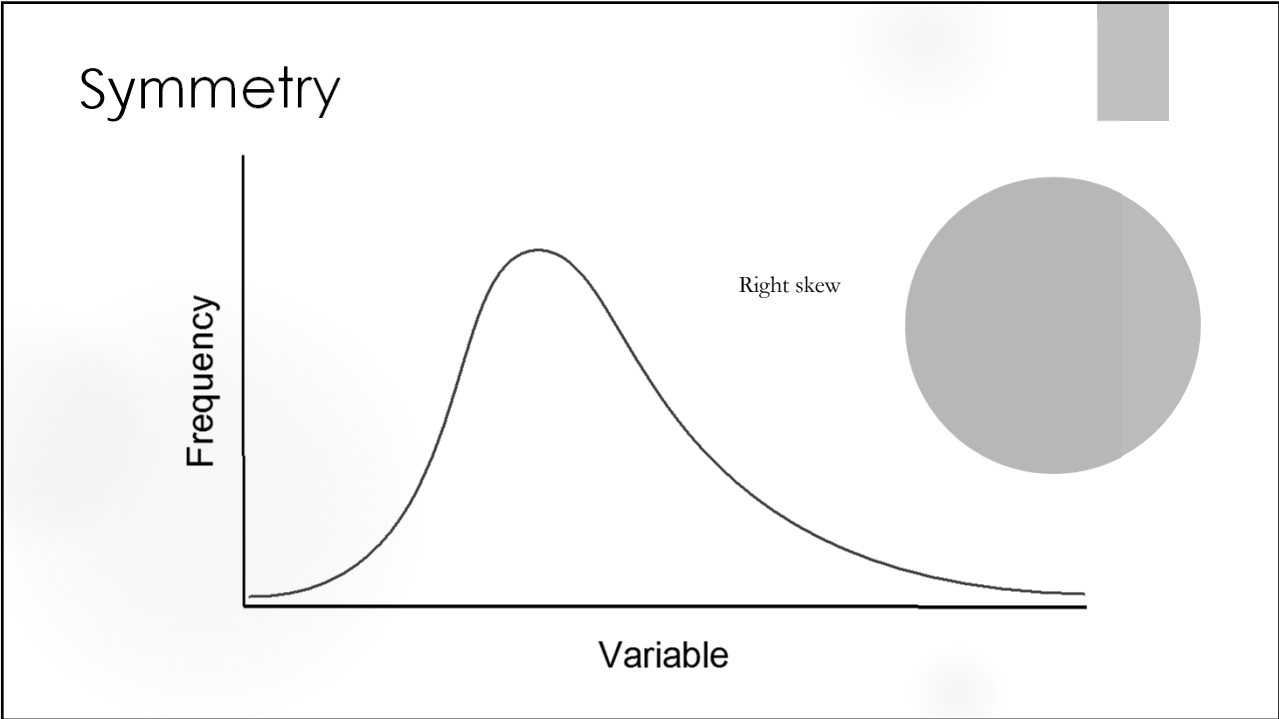
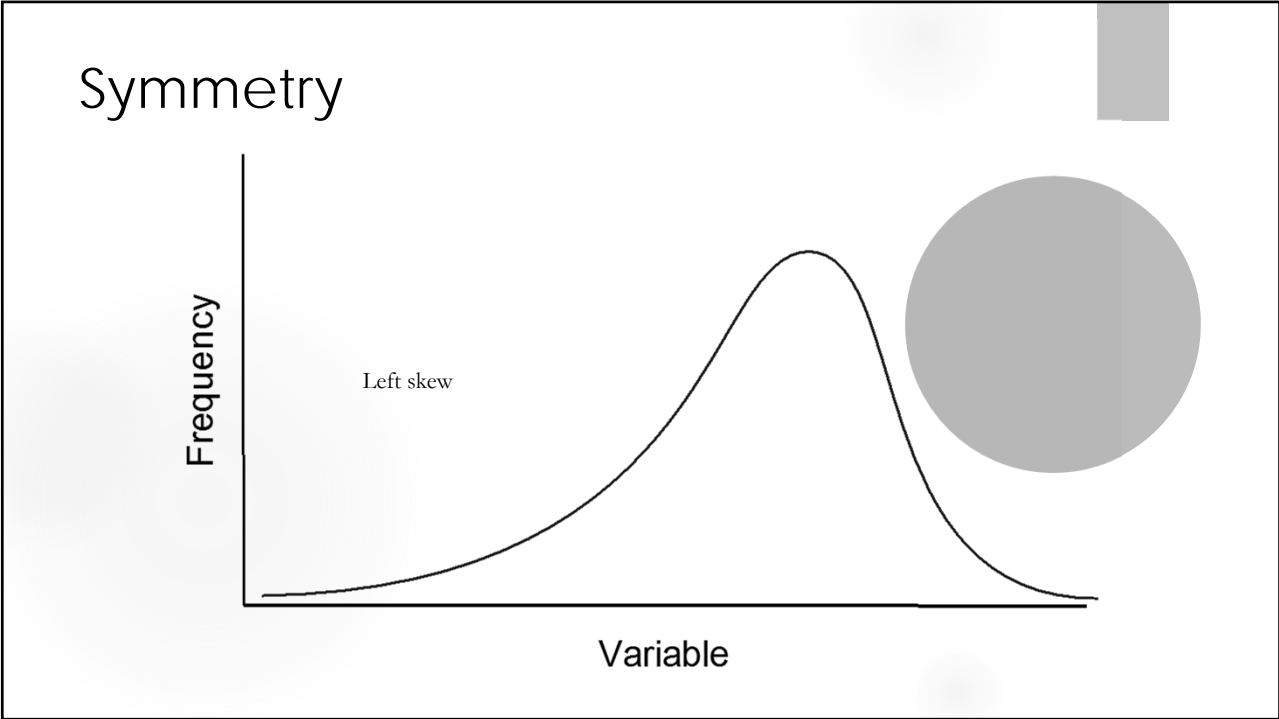
$$Y_i = \frac{1}{\sqrt{2\pi}} e^{\frac{-X_i^2}{2}}$$

Normal distribution

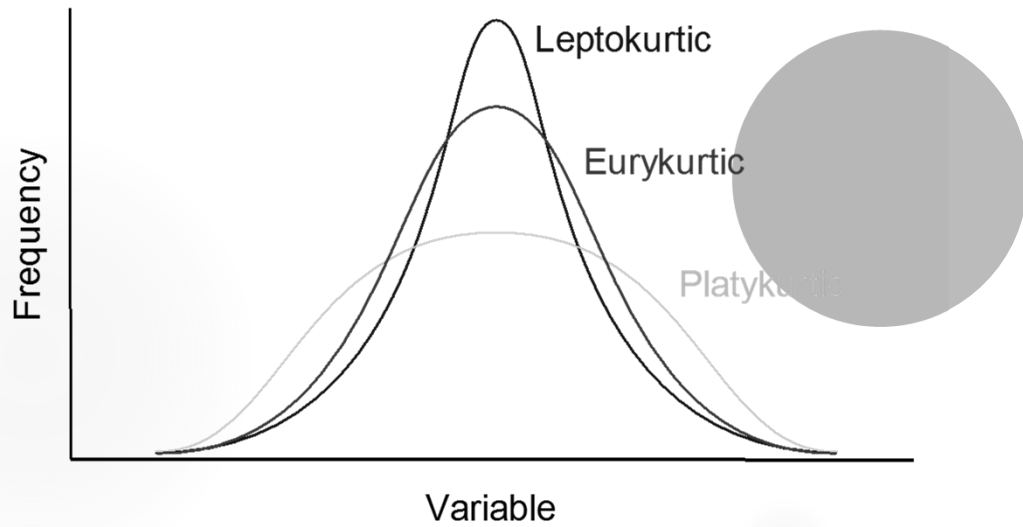
- ▶ The normal curve is “THE standard” in statistics because many large sets of data rather closely approximate a normal curve.
 - ▶ Not all data is normally distributed.
 - ▶ Much biological data is not normally distributed.
 - ▶ Despite what many people say.

Symmetry and kurtosis

- ▶ Distributions can depart from normal in two important ways.
 - ▶ Symmetry – the distribution may not be symmetrical around the mean we call this type of distribution “skewed.”
 - ▶ Kurtosis – “peakedness” – a distribution may be too “pointy” (leptokurtic) or may be too flat (platykurtic) to be a normal curve.



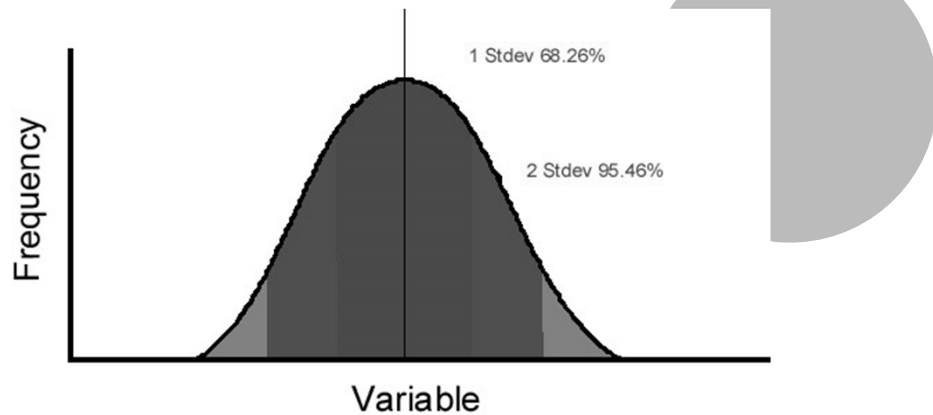
Kurtosis



Proportions of a normal distribution

- ▶ Since standard deviations are in the same units as the mean, the standard deviation can be represented as a distance from the mean ($\mu \pm \sigma$).
- ▶ Frequently normal curves are divided into different areas according to the standard deviations.

Proportions of a normal distribution



Z-score

- The number of standard deviations a particular measurement is above or below the mean (standard deviation units).

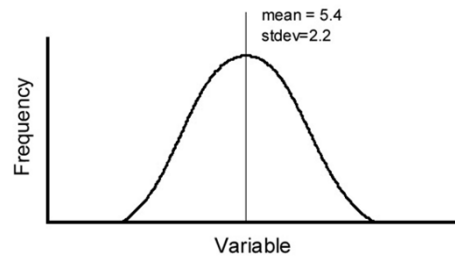
$$Z = \frac{X_i - \mu}{\sigma}$$

Example

- ▶ Data: Normal Distribution

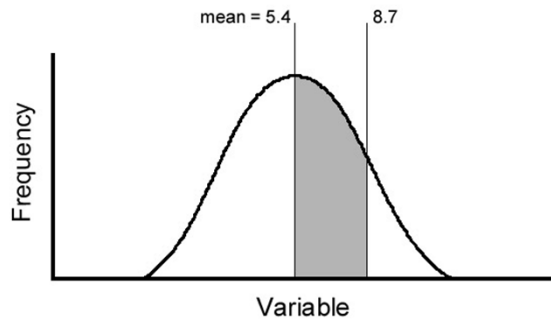
- ▶ $\mu = 5.4$

- ▶ $\sigma = 2.2$



Example

- ▶ What proportion of the distribution is between the mean and 8.7?



Example

$$Z = \frac{X_i - \mu}{\sigma}$$

$$Z = \frac{X_i - \mu}{\sigma} = \frac{8.7 - 5.4}{2.2} = 1.50$$

- ▶ Go to Table A

Example

- ▶ What proportion of the distribution is between the mean and 8.7?
- ▶ 43.32% of the distribution lies between the mean and 8.7.

Distribution of means

- ▶ If you take random samples from a normal population, then calculate the mean for each sample the means of the samples will be normally distributed.

Distribution of means

- ▶ If you take random samples from a non-normal population, then calculate the mean for each sample – the means of the samples will not be normally distributed.

Central Limit Theorem

- ▶ However, as the size of samples increases the distribution of means tends toward normality.
- ▶ Furthermore as the size of samples increases the variance of the distribution of the means will tend to decrease.

Distribution of means

- ▶ The variance of the population of all possible means of samples of size n from a population with a variance σ^2 is

$$\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$$

Distribution of means

- ▶ The standard deviation derived from this variance is called the standard error.

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

Distribution of means

- ▶ Samples work similarly.

$$S_{\bar{x}} = \frac{S}{\sqrt{n}}$$