Descriptive Statistics

• Sample Mean:

$$\overline{y} = \frac{y_1 + y_2 + \ldots + y_n}{n} = \frac{\sum y_i}{n}$$

• Sample Variance

$$s^{2} = \frac{\sum_{i=1}^{n} (y_{i} - \overline{y})^{2}}{n-1} = \frac{\sum_{i=1}^{n} y_{i}^{2} - \frac{\left(\sum_{i=1}^{n} y_{i}\right)^{2}}{n}}{n-1}$$

• Sample Standard Deviation: $s = \sqrt{\text{sample variance}} = \sqrt{s^2}$

- Range = max min
- Interquartile Range (IQR) = $Q_3 Q_1$
- 5 # Summary: min, Q₁, median, Q₃, max
- Outliers:

Lower fence = $Q_1 - 1.5 \times IQR$ Upper fence = $Q_3 + 1.5 \times IQR$

- z-score: $z = \frac{y \mu}{\sigma}$
- Empirical Rule:

Empirical rule: For a relatively bell-shaped data set, approximately 68%, 95%, 99.7% of the measurements are within one, two and three standard deviations of the mean, respectively.

Probability Theory

- Addition Rule: $P(A \text{ or } B) = P(A \cup B) = P(A) + P(B) - P(A \cap B)$
- Complement Rule: $P(A^c) = 1 P(A)$
- Multiplication Rule (general): $P(A \text{ and } B) = P(A \cap B)$ $= P(A) \times P(B|A)$ $= P(B) \times P(A|B)$
- Multiplication Rule for **Independent** Events: If A and B are independent, then $P(A \text{ and } B) = P(A \cap B) = P(A) \times P(B)$
- Conditional Probability of A given B, if P(B) > 0: $P(A/B) = \frac{P(A \cap B)}{P(B)}$
- Probability of A: $P(A) = P(A \cap B) + P(A \cap B^c)$

Population Distributions

- The mean (expected value) of a discrete random variable: $\mu = E(X) = \sum xp(x)$
- The variance of a discrete random variable: $\sigma^2 = Var(X) = \sum_{n=0}^{\infty} (x - \mu)^2 p(x)$
- The standard deviation of a discrete random variable: $\sigma = \sqrt{\sigma^2}$
- For two random variables X and Y, and constants a,
 b, and c:

$$\circ \quad E[aX+b] = aE[X] + b$$

$$\circ V[aX+b] = a^2V[X]$$

$$\circ \quad E[aX + bY + c] = aE[X] + bE[Y] + c$$

o
$$V[aX + bY + c] = a^2V[X] + b^2V[Y]$$
, if X and Y are independent.

Normal Distribution

 $Y \sim Normal(\mu, \sigma), -\infty < y < \infty$

•
$$P(a < Y < b) = P\left(\frac{a - \mu}{\sigma} < Z < \frac{b - \mu}{\sigma}\right)$$

Sampling Distributions

• Sampling Distribution of a Sample Proportion, \hat{p} :

-
$$Mean(\hat{p}) = \mu_{\hat{p}} = p$$

-
$$SD(\hat{p}) = \sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

- For large n ($np \ge 10$ and $n(1-p) \ge 10$),

$$\hat{p} \stackrel{\cdot}{\sim} N \left(\mu_{\hat{p}} = p, \sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}} \right)$$

• Sampling Distribution of a Sample Mean, \bar{y} :

-
$$Mean(\bar{y}) = \mu_{\bar{y}} = \mu$$

$$- SD(\bar{y}) = \sigma_{\bar{y}} = \frac{\sigma}{\sqrt{n}}$$

- If
$$y \sim N(\mu, \sigma)$$
, then

$$\bar{y} \sim N \left(\mu_{\bar{y}} = \mu, \sigma_{\bar{y}} = \frac{\sigma}{\sqrt{n}} \right)$$

- CLT: If
$$y \sim ?(\mu, \sigma)$$
, then for $n \ge 30$

$$\bar{y} \approx N \left(\mu_{\bar{y}} = \mu, \sigma_{\bar{y}} = \frac{\sigma}{\sqrt{n}} \right)$$