

Chapter 8: Random Variables & Probability Models

- **expected value**- $\mu = EV = E(X) = \sum x \cdot P(x)$
 - o $E(X+Y) = E(X) + E(Y)$
 - o $E(X-Y) = E(X) - E(Y)$
 - o $E(2X) = 2E(X)$
 - o $E(X+C) = E(X) + C$

- **Standard deviation**- $\sigma = SD(x) = \sqrt{var(x)} = \sqrt{\sum p(x) \cdot (x-\mu)^2}$
- **Z score**- $(x-\mu) / \sigma$
- $z = \frac{value - mean}{\sigma}$

Binomial Distribution: $B(n, p)$

- $P(x=k) = [n! / (x!(n-x)!)] [p^x q^{n-x}]$ x=successes; n=trials; p=prob of success; q=prob of failure
- $z = (x-np) / \sqrt{npq}$

- **variance**- $\sigma^2 = Var(X) = \sum (x-\mu)^2 P(x)$
 - o $Var(x) = E[(x-\mu)^2] = E(x^2) - \mu^2$
 - o $Var(X+Y) = Var(X) + Var(Y)$
 - o $Var(X-Y) = Var(X) + Var(Y)$
 - o $Var(X+C) = Var(X)$
 - o $Var(a+bx) = b^2 var(x)$

Assumptions
 $np > 10$ $nq > 10$
 random
 independent
 10% of population

Normal Distribution:

- $\mu = E(X) = np$ $N(\mu, \sigma)$
- $\sigma^2 = Var(x) = npq$
- $\sigma = \sqrt{npq}$

use binomial program
 or
 binompdf (n, p, x): spec number of successes
 binomcdf (n, p, x): values $\leq x$

Chapter 9: Sampling Distributions and Confidence Intervals for Proportions

Sample Proportions:

- $p^* = x/n$ x=successes
- $\sigma = \sqrt{p^*q^*/n}$ *use $p=1/2$ when not given
- $p^* \sim N(p, \sqrt{pq/n})$
- $z = (p^* - p_0) / (\sqrt{pq/n})$

Confidence Intervals for Proportions: $p^* \pm z^* \sqrt{p^*q^*/n}$

Margin of Error pop. prop. is: $z^* \sqrt{p^*q^*/n}$ z^* depends on CI%

Sample Size:

- $p \sim N(p, \sqrt{pq/n})$
- $n = [z^* \sqrt{pq} / m]^2$ *if p is unknown, use $p=1/2$

$SD(\hat{p}) = \sqrt{\frac{pq}{n}}$
 $SE(\hat{p}) = \sqrt{\frac{pq}{n}}$

Z* Critical Values:

CL	A	1-side	2-side
90%	.1	1.28	1.645
95%	.05	1.645	1.96
99%	.01	2.33	2.576
98%			2.492

SE is the estimate of SD

Chapter 10: Testing Hypotheses for Proportions

One Proportion Z-Test:

1. Null, Alternative

$H_0: p = p_0$

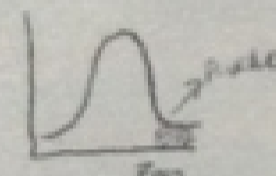
$H_A: p < p_0$ (left)
 $p > p_0$ (right)
 $p \neq p_0$ (two-tail)

2. Test Statistic (z)

$z_{test} = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0 q_0}{n}}}$

reject H_0 if z_{test} falls into a z_{α} region

3. P-value



normalcdf (z, infinity) - P
 (-infinity, z) - P
 (-z, z) - P

4. P vs. Alpha

$P < \alpha$, reject $P > \alpha$, fail to reject

Rejection Regions:

1. $p < p_0$, reject is $z < -z_{\alpha}$
2. $p > p_0$, reject if $z > z_{\alpha}$
3. $p \neq p_0$, reject is z falls in region

Confidence Intervals: reject is value is outside interval; fail to reject if value is inside interval: $CI = p^* \pm z^* \sqrt{p^*q^*/n}$ z^* interval

Chapter 11: Hypotheses Tests for Means

- σ is given- use z
- σ is not given- use t

$z = (x - \bar{x}) / \sigma$
 $t = [\bar{x} - \mu] / [s/\sqrt{n}]$

$x \sim N(\mu, \sigma/\sqrt{n})$
 df: $n-1$

$SE(x) = s/\sqrt{n}$

$CI = \bar{x} \pm z^* [s/\sqrt{n}]$
 $CI = \bar{x} \pm t^* (\frac{s}{\sqrt{n}})$

One Sample T-Test:

1. $H_0: \mu_0 = \mu$

$H_A: \mu < \mu_0$ (left) $\mu > \mu_0$ (right) $\mu \neq \mu_0$ (two-tail)

2. $t = [x - \mu] / [s/\sqrt{n}]$

3. P-value

4. $P < \alpha$, reject $P > \alpha$, fail to reject

Rejection Region:

1. $\mu < \mu_0$, reject is $t < -t_{\alpha}$
2. $\mu > \mu_0$, reject if $t > t_{\alpha}$
3. $\mu \neq \mu_0$, reject is t falls in region

Confidence Interval:
 $\bar{x} \pm t [s/\sqrt{n}]$

Sample Size:

Case I: σ is known $n = [z^* \sigma] / [m]^2$

Case II: σ is unknown

$n_1 = [z^* \sigma] / [m]^2$ use n_1 for df to find t^*
 $n_2 = [t^* \sigma] / [m]^2$ always round up

- Type I Error: reject H_0 when it is true - α
- Type II Error: fail to reject H_0 when H_A is true - β

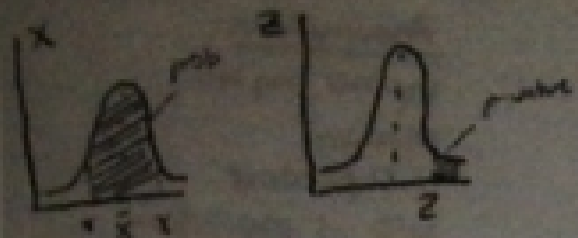
smaller p-value \rightarrow strong evidence against H_0

CI: we are 95% confident that the true proportion is in the interval

Calc. Functions

normalcdf (lower x , upper x , mean, sd) = probability

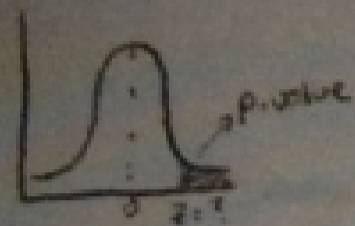
normalcdf (lower z , upper z) = p -value



invNorm(p -value) = z -value [2nd][Vars]

use when given percentile to find x

$$z = \frac{x - \bar{x}}{\sigma}$$



Ztest [Stat] → Tests

- input $\mu_0, \bar{x}, \sigma, n$, claim
- gives you p -val and z -val
- compare p to α

Ttest [Stat] → Tests

- input μ_0, \bar{x}, s, n , claim
- gives you t -val and p -val
- compare p to α

Binomial Probability

1. only 2 outcomes
2. n = trials
 q = prob of failure
 p = prob of success

To find $P(X = \#)$

→ binomial pdf (n, p, x)

To find $P(X < \#)$

→ binomialcdf (n, p, x)

To find $P(X > \#) \Rightarrow 1 - P(X \leq \# - 1)$

→ binomialcdf ($n, p, x - 1$)

Ex. prob at least 7 heads?

$$P(X \geq 7) = 1 - P(X \leq 6)$$

$$\text{binomcdf}(10, .5, 6) = 17\%$$

P = true population proportion

P_0 = hypothesized value

\hat{p} = observed proportion

[2nd][Vars]

binompdf (n, p, x) find # of successes
Ex. flip a coin 10 times, prob of 8 heads?

[2nd]

binomcdf (n, p, x) find values $\leq x$

Ex. Roll dice 10 times, prob less than 3 5's
binompdf ($10, \frac{1}{6}, 3$)

[2nd]

cont

invT (area, df) = t_n , [2nd][Vars]

Ex. CI = 95%, $n = 11$

$$\text{invT}(.975, 10)$$

two-sided

one-sided:

$$\text{invT}(.95, 10)$$

Ex. Prob = 25% $n = 10$

$$\text{invT}(.25, 9)$$