

# 1 - statistical inference

Estimate the unknown parameter  $\theta$  from a sample  
3.3.3, 3.3.3, 7.7.7 drawn from a discrete distribution with probability mass function

$$P(X) = \begin{cases} \theta & \text{if } x=0 \\ 1-\theta & \text{if } x=1 \\ 0 & \text{otherwise} \end{cases}$$

Compute two estimates of  $\theta$ :  
 ✓ the method of moments estimator:  $\hat{\theta}_{MOM} = \frac{2(1-\bar{x})}{1-\bar{x}+2\bar{x}^2}$   
 ✓ the maximum likelihood estimator:  $\hat{\theta}_{MLE} = \frac{1}{1-\bar{x}}$

Also,  $\hat{\theta}_{MLE} = \frac{1}{1-\bar{x}} = \frac{1}{1-\frac{1}{n}\sum_{i=1}^n x_i} = \frac{n}{n-\sum_{i=1}^n x_i}$

✓ Estimate the standard error of each estimator of  $\theta$ .

9.2. The number of times a computer code is executed until it runs without errors has a Geometric distribution with unknown parameter  $p$ . For 5 independent computer projects, a student records the following numbers of runs:

3 7 5 3 2

Estimate  $p$

(a) by the method of moments:  
 (b) by the method of maximum likelihood:

9.3. Use method of moments and method of maximum likelihood to estimate

(a) parameters  $a$  and  $b$  if a sample from  $Unif(a, b)$  distribution is observed;  
 (b) parameter  $\lambda$  if a sample from  $Exp(\lambda)$  distribution is observed;  
 (c) parameter  $\mu$  if a sample from  $N(\mu, \sigma^2)$  distribution is observed, and we already know  $\sigma$ ;  
 (d) parameter  $\sigma$  if a sample from  $N(\mu, \sigma^2)$  distribution is observed, and we already know  $\mu$ ;  
 (e) parameters  $\mu$  and  $\sigma$  if a sample from  $N(\mu, \sigma^2)$  distribution is observed, and both  $\mu$  and  $\sigma$  are unknown.

✓ A sample of 3 observations  $(X_1 = 0.4, X_2 = 0.7, X_3 = 0.9)$  is collected from a continuous distribution with density

$$f(x) = \begin{cases} \frac{1}{2}(x-0.2)^2 & \text{if } 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

Estimate  $\theta$  by your favorite method.

✓ Verify columns 3-5 in Table 8.1 on p. 273. Section 9.7 will help you with the last row of the table.

✓ In order to ensure efficient usage of a server, it is necessary to estimate the mean number of randomly selected users who log in to the server at the same time. A sample of 100 randomly selected users is  $\{27, 30, 32, \dots, 100\}$ , with a standard deviation  $\sigma = 0.2$ .

✓ Construct a 90% confidence interval for the expectation of the number of concurrent users. One mean  $\text{conf. I.} (73.3, 100, 0.5, 14.115)$

✓ At the 0.05 significance level, do these data provide significant evidence that the mean number of concurrent users is less than 100? Ho:  $H_0: \mu = 100$ , Ha:  $H_a: \mu < 100$

✓ Installation of a new software application took an average of 32 minutes. A technician installs the software on 40 different computers, with an average installation time of 42 minutes. Compute a 95% confidence interval for the population mean installation time.

✓ A computer technician installs this software on 60 different computers, with an average installation time of 42 minutes. A technician installs the software on your PC. What is the probability that the installation time will be within the interval computed in (a)?

✓ Software of entry-level computer engineers have Normal distribution with unknown mean and variance. Three randomly selected computer engineers have salaries ( $\$$  in 1000s):  
 $30, 50, 70$ . Construct a 95% confidence interval for the average salary of an entry-level engineer.  
 ✓ Does this sample provide a sufficient evidence, at the level of significance  $\alpha = 0.05$ , that the mean salary of entry-level engineers is  $\mu > 40000$ ? Ho:  $H_0: \mu = 40000$ , Ha:  $H_a: \mu > 40000$

✓ Looking at this sample, one may think that the starting salary of entry-level engineers is looking for. Construct a 90% confidence interval for the standard deviation of entry-level engineers.

$$\frac{(n-1)s^2}{\chi^2_{\alpha/2}} < \frac{(n-1)s^2}{\chi^2_{1-\alpha/2}} < \frac{(n-1)s^2}{\chi^2_{1-\alpha}}$$

$$\frac{2(50-30)}{2.778} < \frac{2(50-30)}{3.214} < \frac{2(50-30)}{3.070}$$

**standard = 110.9**

✓ We have to accept or reject a large number of items. For quality control purposes, we collect a sample of 200 items and find 24 defective items in it.

(a) Construct a 95% confidence interval for the proportion of defective items in the whole shipment.  $[0.0928, 0.1071]$

(b) The manufacturer claims that at most one in 10 items in the shipment is defective. At the 0.05 level of significance, does the sufficient evidence to reject this claim?  
 Ho:  $H_0: p \leq 0.1$ , Ha:  $H_a: p > 0.1$

✓ Note: In Section 9.10, we discussed the P-value approach. We consider an alternative hypothesis  $H_a$  to be true. A sample of 100 items produced by the manufacturer shows 12 defectives. Is there significant evidence that the quality of items produced by the new supplier is  $\text{worse}$  than the quality of items in Exercise 9.10? What is the P-value?  $H_0: H_0: p \leq 0.1$ ,  $H_a: p > 0.1$ ,  $\alpha = 0.05$ ,  $n = 100$ ,  $x = 12$ .  $P = 0.1578 \rightarrow \text{no, not enough evidence}$

9.12. An electronic parts factory produces resistors. Statistical analysis of the output shows that resistances follow an approximately Normal distribution with a standard deviation of 0.02 ohms. A sample of 12 resistors has the average resistance of 0.6 ohms.

(a) Based on these data, construct a 95% confidence interval for the population mean resistance.  
 (b) If the actual population mean resistance is exactly 0.6 ohms, what is the probability that an average of 12 resistors has the average resistance of 0.62 ohms?

9.13. Compute a P-value for the right-tail test in Example 9.20 on p. 272 and state your conclusion about a significant increase in the number of concurrent users.

9.14. Is there significant difference in speed between the two servers in Example 9.21 on p. 267?

(a) Use the confidence interval from Example 9.21 to conduct a two-sided test at the 5% level of significance.  
 (b) Compute a P-value for the two-sided test in (a).  
 (c) Is server A really faster? How strong is the evidence? Formulate the suitable hypothesis and alternative and compute the corresponding P-value.

State your conclusions in (a), (b), and (c).

9.15. According to Example 9.17 on p. 257, there is no significant difference, at the 5% level, between the mean time spent by men and women on the Internet. However, the  $\alpha = 0.10$  was chosen rather than 0.05, and the researcher still does not know if he can trust the results when planning his campaign. Can we compare the two servers at all reasonable levels of significance? State your conclusions in (a), (b), and (c).

✓ A sample of 200 items from lot A contains 10 defective items, and a sample of 300 items from lot B is found to contain 18 defective items.

(a) Construct a 95% confidence interval for the proportion of defective items in lot A.  
 (b) At a significance level of 0.05, is there a significant difference between the quality of the two lots?  $H_0: p_A = p_B$ ,  $H_a: p_A \neq p_B$

✓ A news agency publishes results of a recent poll. It reports that 40000 people in the US were asked to participate in the poll. Of those, 18000 supported candidate M, whereas only 35% supported M. Compute a 95% confidence interval for each of the total estimates, 10%, 35%, and 65%. Notice that 900 people participated in the poll, and the reported margin of error typically corresponds to 95% confidence intervals.

✓ Consider the data about the number of blocked intrusions in Exercise 8.1, p. 223.

✓ Construct a 95% confidence interval for the difference between the average number of intrusion attempts per day before and after the new firewall settings (assume equal variances).  
 ✓ Can we claim significant reduction in the rate of intrusion attempted? The number of intrusion attempts each day has approximately Normal distribution. Compute P-values and state your conclusions under the assumption of equal variances and without it. Does this assumption make a difference?

✓ equal var  $\rightarrow t = 3.596$   
 $t = 3.596$   
 unequal var  $\rightarrow t = 3.625$   
 $v = 20$

Ho:  $X_1 \leq X_2$   
 H<sub>a</sub>:  $X_1 > X_2$

inverse +

before:  $X_1 = 124$   
 $X_2 = 67.616$   
 $X = 50$

after:  $X_1 = 20$   
 $X_2 = 16.562$   
 $X = 18.535$

$t = 2.82$  [L: 2.149, R: 3.53]  
 $t = 2.82$  [L: 2.149, R: 3.53]  
 t = 2.82  
 $t = 2.82$  [L: 2.149, R: 3.53]

$t = 3.596$   
 $t = 3.596$   
 $t = 3.625$   
 $t = 3.625$   
 $P = 0.005$   
 $P = 0.005$

$\leftarrow P < 0.05 \rightarrow \text{reject } H_0$

Null hypothesis $H_0$	Parameter, estimator $\theta, \hat{\theta}$	If $H_0$ is true:		Test statistic $Z = \frac{\hat{\theta} - \theta_0}{\sqrt{\text{Var}(\hat{\theta})}}$
		$E(\hat{\theta})$	$\text{Var}(\hat{\theta})$	
One-sample Z-tests for means and proportions, based on a sample of size $n$				
$\mu = \mu_0$	$\mu, \bar{X}$	$\mu_0$	$\frac{\sigma^2}{n}$	$\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$
$p = p_0$	$p, \hat{p}$	$p_0$	$\frac{p_0(1 - p_0)}{n}$	$\frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$
Two-sample Z-tests comparing means and proportions of two populations, based on independent samples of size $n$ and $m$				
$\mu_X - \mu_Y = D$	$\mu_X - \mu_Y, \bar{X} - \bar{Y}$	$D$	$\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}$	$\frac{\bar{X} - \bar{Y} - D}{\sqrt{\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}}}$
$p_1 - p_2 = D$	$p_1 - p_2, \hat{p}_1 - \hat{p}_2$	$D$	$\frac{p_1(1 - p_1)}{n} + \frac{p_2(1 - p_2)}{m}$	$\frac{\hat{p}_1 - \hat{p}_2 - D}{\sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n} + \frac{\hat{p}_2(1-\hat{p}_2)}{m}}}$
$p_1 = p_2$	$p_1 - p_2, \hat{p}_1 - \hat{p}_2$	0	$p(1 - p) \left( \frac{1}{n} + \frac{1}{m} \right)$ , where $p = p_1 = p_2$	$\frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1 - \hat{p}) \left( \frac{1}{n} + \frac{1}{m} \right)}}$ where $\hat{p} = \frac{n\hat{p}_1 + m\hat{p}_2}{n + m}$

Hypothesis $H_0$	Conditions	Test statistic $t$	Degrees of freedom
$\mu = \mu_0$	Sample size $n$ ; unknown $\sigma$	$t = \frac{\bar{X} - \mu_0}{s/\sqrt{n}}$	$n - 1$
$\mu_X - \mu_Y = D$	Sample sizes $n, m$ ; unknown but equal standard deviations, $\sigma_X = \sigma_Y$	$t = \frac{\bar{X} - \bar{Y} - D}{s_{\bar{X}-\bar{Y}} \sqrt{\frac{1}{n} + \frac{1}{m}}}$	$n + m - 2$
$\mu_X - \mu_Y = D$	Sample sizes $n, m$ ; unknown, unequal standard deviations, $\sigma_X \neq \sigma_Y$	$t = \frac{\bar{X} - \bar{Y} - D}{\sqrt{\frac{s_X^2}{n} + \frac{s_Y^2}{m}}}$	Satterthwaite approximation, formula (9.12)

TABLE 9.2: Summary of T-tests.

### ① parameter estimation

methods of moment = calculate mean from given data

$$= \sum x \cdot p(x) \text{ (discrete)}$$

$$= \int x f(x) \text{ (continuous)}$$

max likelihood =  $L = \text{product of all given values probabilities}$

- write both side as  $L^n$
- take derivative = 0
- find  $\theta$  : check if it is max point by checking second derivative  $L$  is negative

standard error = find a equation like  $\theta = a\bar{x} + b$  → does not matter  
 $\sigma(\theta)$  = is error  $\downarrow$  mean

to find it  $\sigma(\theta) = a \cdot \sigma(x)$  (sample)  
 $= a \cdot \frac{\sigma(x)}{\sqrt{n}}$  (population)

 $E(x) = \text{mean} = \sum x p(x) = \int x f(x)$   
 $E(x^2) = \sum x^2 p(x) = \int x^2 f(x)$   
 $= a \cdot \sqrt{\frac{E(x^2) - E(x)^2}{n}}$

## CONFIDENCE INTERVALS

one population

- 1- mean ( $\mu$ )
- 2- proportion ( $p$ )
- 3- variance ( $\sigma^2$ )

t-mean

### ⓐ known variance

$$\bar{x} - e < \mu < \bar{x} + e \quad e = (z_{\frac{\alpha}{2}}) \cdot \frac{\sigma}{\sqrt{n}}$$

### ⓑ unknown variance with $n \geq 30$

$$\bar{x} - e < \mu < \bar{x} + e \quad e = (z_{\frac{\alpha}{2}}) \cdot \frac{s}{\sqrt{n}} \quad s = \text{std dev of sample}$$

### ⓒ unknown variance with $n < 30$

$$\bar{x} - e < \mu < \bar{x} + e \quad e = t_{(\frac{\alpha}{2}, n-1)} \cdot \frac{s}{\sqrt{n}} \quad n-1 = \text{degrees of freedom}$$

2- proportion

$$\hat{p} - e < p < \hat{p} + e \quad e = \left( \frac{\alpha}{2} \right) \cdot \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

3-variance

$$\frac{(n-1)s^2}{\chi_{\frac{\alpha}{2}}^2} < \sigma^2 < \frac{(n-1)s^2}{\chi_{(1-\frac{\alpha}{2})}^2}$$

two populations

1-mean ( $\mu_1 - \mu_2$ )

2-proportion ( $p_1 - p_2$ )

3-variance

1-mean

a known populations variances

$$(\bar{x}_1 - \bar{x}_2) - e < \mu_1 - \mu_2 < (\bar{x}_1 - \bar{x}_2) + e \quad e = \left( \frac{\alpha}{2} \right) \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

b unknown populations variances with  $\sigma_{\text{sample}} = \sigma_{\text{population}}$

$$(\bar{x}_1 - \bar{x}_2) - e < \mu_1 - \mu_2 < (\bar{x}_1 - \bar{x}_2) + e \quad e = t\left(\frac{\alpha}{2}, v\right) \cdot s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

$v = n_1 + n_2 - 2$

$$s_p = \sqrt{\frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1 + n_2 - 2}}$$

c unknown populations variances with  $\sigma_{\text{sample}} \neq \sigma_{\text{population}}$

$$(\bar{x}_1 - \bar{x}_2) - e < \mu_1 - \mu_2 < (\bar{x}_1 - \bar{x}_2) + e$$

$$\text{closest int} \approx v = \frac{\left( \frac{s_1^2}{n_1} + \frac{s_2^2}{n_2} \right)^2}{\frac{(s_1^2)^2}{n_1-1} + \frac{(s_2^2)^2}{n_2-1}} \rightarrow \text{Satterthwaite formula}$$

$$\min(n_1, n_2) < v < (n_1 + n_2)$$

$$e = t\left(\frac{\alpha}{2}, v\right) \cdot \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

we know population

2-proportion

$$(\hat{p}_1 - \hat{p}_2) - e < p_1 - p_2 < (\hat{p}_1 - \hat{p}_2) + e \quad e = \left( \frac{\alpha}{2} \right) \cdot \sqrt{\frac{\hat{p}_1 \cdot (1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2 \cdot (1-\hat{p}_2)}{n_2}}$$

## HYPOTHESIS TESTING

one population

- 1- mean ( $\mu$ )
- 2- proportion ( $p$ )
- 3- variance ( $\sigma^2$ )

### 1-mean

- (a) Known variance (pop)

$$\text{test statistic} = Z = \frac{\bar{x} - \mu_0}{\left(\frac{\sigma}{\sqrt{n}}\right)}$$

claimed value in the hypothesis

- (b) unknown variance with  $n \geq 30$

$$\text{test statistic} = Z = \frac{\bar{x} - \mu_0}{\left(\frac{s}{\sqrt{n}}\right)}$$

in hypothesis  
sample std dev

- (c) unknown variance with  $n < 30$

$$\text{test statistic} = t = \frac{\bar{x} - \mu_0}{\left(\frac{s}{\sqrt{n}}\right)}$$

in hypothesis  
sample std dev

- if  $\sigma(x)$  given

$$\sigma(\bar{x}) = \frac{\sigma(x)}{\sqrt{n}}$$

in no form

- if  $\sigma(\bar{x})$  is given

$$z = \frac{\bar{x} - \mu_0}{\sigma(\bar{x})}$$

critical value =  $z_\alpha$  (one sided)  
two sided:  $\frac{\alpha}{2}$

critical value =  $z_\alpha$  (one sided)  
two sided:  $\frac{\alpha}{2}$

from calculator  
(two sided)  
or  
(one sided)  
critical value =  $t(\alpha, n-1)$   
degree of freedom

### 2-proportion

$$\text{test statistic} = Z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

in hypothesis

critical value =  $z_\alpha$  (one sided)  
( $\frac{\alpha}{2} \rightarrow$  two sided)

### two populations

- 1- mean ( $\mu_1, \mu_2$ )
- 2- proportion ( $p_1 - p_2$ )
- 3- variance

### 1-mean

- (a) Known variance (pop)

$$\text{test statistic} = Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

critical value =  $z_\alpha$  (one sided)  
( $\frac{\alpha}{2} \rightarrow$  two sided)

⑥ unknown <sup>population</sup> variance with  $\sigma_1 = \sigma_2$  ( $s_1 = s_2$ )

just hint  
one/two tailed  
degrees of freedom

test statistic =  $t = \frac{\bar{x}_1 - \bar{x}_2}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$

critical value =  $t(\alpha, n+m-2)$

$s_p = \sqrt{\frac{(n-1)s_x^2 + (m-1)s_y^2}{n+m-2}}$

standard dev of population =  $\sigma_1 = \sigma_2$

parameter estimate

⑦ unknown <sup>population</sup> variance with  $\sigma_1 \neq \sigma_2$

test statistic =  $t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$

critical value =  $t(\alpha, v)$

scatterthwaite approx  $v = \frac{\frac{s_x^2}{n} + \frac{s_y^2}{m}}{\frac{s_x^4}{n^2(n-1)} + \frac{s_y^4}{m^2(m-1)}}$

### 2-proportion

test statistic =  $Z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}}$

critical value =  $Z_\alpha$

(pooled better approx)  $Z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p}) \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$      $\hat{p} = \frac{x_1 + x_2}{n_1 + n_2}$

p value

Testing  $H_0$  with a P-value

For $\alpha < P$ ,	accept $H_0$
For $\alpha > P$ ,	reject $H_0$
<i>Practically,</i>	
If $P < 0.01$ ,	reject $H_0$
If $P > 0.1$ ,	accept $H_0$

Hypothesis $H_0$	Alternative $H_A$	P-value	Computation
$\theta = \theta_0$	right-tail $\theta > \theta_0$	$P\{Z \geq Z_{\text{obs}}\}$	$1 - \Phi(Z_{\text{obs}})$
	left-tail $\theta < \theta_0$	$P\{Z \leq Z_{\text{obs}}\}$	$\Phi(Z_{\text{obs}})$
	two-sided $\theta \neq \theta_0$	$P\{ Z  \geq  Z_{\text{obs}} \}$	$2(1 - \Phi( Z_{\text{obs}} ))$

↑ this x will or p>0.05

Hypothesis $H_0$	Alternative $H_A$	P-value	Computation
$\theta = \theta_0$	right-tail $\theta > \theta_0$	$P\{t \geq t_{\text{obs}}\}$	$1 - F_\nu(t_{\text{obs}})$
	left-tail $\theta < \theta_0$	$P\{t \leq t_{\text{obs}}\}$	$F_\nu(t_{\text{obs}})$
	two-sided $\theta \neq \theta_0$	$P\{ t  \geq  t_{\text{obs}} \}$	$2(1 - F_\nu( t_{\text{obs}} ))$

TABLE 9.4:  $P$ -values for T-tests ( $F_\nu$  is the cdf of T-distribution with the suitable number  $\nu$  of degrees of freedom).

```
#include <iostream>
#include <math.h>

double df_satt_formula(double s1,double s2,double n1, double n2);//std1
std2 n1 n2 -- no mean values
//one population mean confidence interval
void one_mean_conf(double mean, double n, double std_dev, double
zt_val);
void one_prop_conf(double portion_in_n, double total_n, double z_val);
//two populations mean confidence interval
void two_mean_conf(double mean1, double mean2, double std_dev1,
double std_dev2, double n1, double n2, double zt_val);//known pop var +
is also for. unknow unqual variances
void two_mean_conf_unk_eq(double mean1, double mean2, double
std_dev1, double std_dev2, double n1, double n2, double t_val);//unknown
pop var, equal
void two_prop_conf(double total_n1, double total_n2, double
portion_in_n1, double portion_in_n2, double z_val);
//one population
double one_mean(double mean, double claimed_mean, double std_dev,
```

```

double n);
double one_prop(double claimed_p , double portion_in_n, double
total_n);
//two populations
double two_mean(double mean1, double mean2, double std_1, double
std_2, double n1, double n2);
double two_prop(double total_n1, double total_n2, double portion_in_n1,
double portion_in_n2);
//pooled is used mostly
double two_prop_pooled(double total_n1, double total_n2, double
portion_in_n1, double portion_in_n2);

double sp(double s1,double s2,double n1, double n2){
    return sqrt(((n1-1)*s1*s1+(n2-1)*s2*s2)/(n1+n2-2));
}

```

```

int main()
{
    one_prop_conf(55, 154, 1.43953);
    std::cout<<"---\n";
    // std::cout << two_mean(6.2, 5.8, 1.5, 1.1, 55, 55);
    std::cout<<"\n---\n";
    // two_prop_conf(400,400, 192, 148, 1.95996);

}

```

```

double df_satt_formula(double s1,double s2,double n1, double n2){
    return (s1*s1/n1 + s2*s2/n2)*(s1*s1/n1 + s2*s2/n2)/ (s1*s1*s1/
(n1*n1*(n1-1)) + s2*s2*s2*s2/(n2*n2*(n2-1)));
}

void one_mean_conf(double mean, double n, double std_dev, double
zt_val){
    double e = zt_val * std_dev/sqrt(n);
    std::cout<<"mean: "<<mean<<"\nmargin:<<e<<"\ninterval: [ "<<mean-
e<<" , "<<mean+e<<" ]\n";
}

void one_prop_conf(double portion_in_n, double total_n, double z_val){

```

```

void one_prop_conf(double portion_in_n, double total_n, double z_val){
```

```

    double p = portion_in_n/total_n;
    double e = z_val * sqrt(p*(1-p)/total_n);
    std::cout<<"propotion: "<<p<<"\nmargin:"<<e<<"\ninterval: [ "<<p-
e<<" , "<<p+e<<" ]\n";
}

```

```

void two_mean_conf(double mean1, double mean2, double std_dev1,
double std_dev2, double n1, double n2, double zt_val){
    double e = zt_val * sqrt(std_dev1*std_dev1/n1 + std_dev2*std_dev2/
n2);
    double dif = mean1-mean2;
    std::cout<<"mean differences: "<<dif<<"\nmargin:"<<e<<"\ninterval:
[ "<<dif-e<<" , "<<dif+e<<" ]\n";
}

```

```

void two_mean_conf_unk_eq(double mean1, double mean2, double
std_dev1, double std_dev2, double n1, double n2, double t_val){
    double sp = sqrt( ( (n1-1)*std_dev1*std_dev1 +
(n2-1)*std_dev2*std_dev2 ) / (n1+n2-2));
    double e = t_val * sp* sqrt(1/n1 + 1/n2);
    double dif = mean1-mean2;
    std::cout<<"mean differences: "<<dif<<"\nmargin:"<<e<<"\ninterval:
[ "<<dif-e<<" , "<<dif+e<<" ]\n";
}

```

```

void two_prop_conf(double total_n1, double total_n2, double
portion_in_n1, double portion_in_n2, double z_val){
    double p_1=portion_in_n1/total_n1;
    double p_2=portion_in_n2/total_n2;
    double dif = p_1-p_2;
    double e = z_val * sqrt(p_1*(1-p_1)/total_n1 + p_2*(1-p_2)/total_n2);
    std::cout<<"proportion differences:
"<<dif<<"\nmargin:"<<e<<"\ninterval: [ "<<dif-e<<" , "<<dif+e<<" ]\n";
}

```

```

//hypothesis
double one_prop(double new_p , double portion_in_n, double total_n){
    double old_p=portion_in_n/total_n;
    return (old_p-new_p)*sqrt(total_n / (new_p*(1-new_p)));
}

```

```

double one_mean(double mean, double claimed_mean, double std_dev,
double n){
    return (mean-claimed_mean)*sqrt(n)/std_dev;
}

double two_mean(double mean1, double mean2, double std_1, double
std_2, double n1, double n2){
    return (mean1-mean2)/sqrt((std_1*std_1)/n1 +(std_2*std_2)/n2);
}

double two_prop(double total_n1, double total_n2, double portion_in_n1,
double portion_in_n2){
    double p_1=portion_in_n1/total_n1;
    double p_2=portion_in_n2/total_n2;
    return (p_1-p_2)/sqrt(p_1*(1-p_1)/total_n1 + p_2*(1-p_2)/total_n2);
}

double two_prop_pooled(double total_n1, double total_n2, double
portion_in_n1, double portion_in_n2){
    double p_1=portion_in_n1/total_n1;
    double p_2=portion_in_n2/total_n2;
    double p=(portion_in_n1+portion_in_n2)/(total_n1 + total_n2);
    return (p_1-p_2)/sqrt(p*(1-p)*(1/total_n1 + 1/total_n2));
}

```