

4/28/09 Math 152 (on 4/30/09  
the lecture was all  
computer examples).

Review: Confidence intervals and  
Hypothesis testing  
(Confidence intervals for the Mann-Whitney test)

e.g.  $X_1, \dots, X_n$  i.i.d.  $N(\mu, \sigma^2)$

$\mu$ ,  ~~$\sigma$~~  unknown.  
 $\sigma^2$  known

$$H_0: \mu = \mu_0$$

$$H_A: \mu \neq \mu_0.$$

A test at level  $\alpha$  rejects for

$$|\bar{X} - \mu_0| > x_0$$

with  $x_0$  chosen so that:  
 $P(|\bar{X} - \mu_0| > x_0) = \alpha.$

if  $H_0$  is true.

$$\text{i.e. } x_0 = \sigma_{\bar{X}} z(\alpha/2)$$

$$\text{where } z(\alpha/2) = \Phi^{-1}(1 - \alpha/2)$$

$$\text{and } \sigma_{\bar{X}} = \sigma/\sqrt{n}$$

The test accepts when

$$|\bar{X} - \mu_0| \leq \sigma_{\bar{X}} z(\alpha/2)$$

$$\Leftrightarrow -\sigma_{\bar{X}} z(\alpha/2) \leq \bar{X} - \mu_0 \leq \sigma_{\bar{X}} z(\alpha/2)$$

$$\Leftrightarrow \bar{X} - \sigma_{\bar{X}} z(\alpha/2) \leq \mu_0 \leq \bar{X} + \sigma_{\bar{X}} z(\alpha/2)$$

and we see that  $\mu_0$  lies  
in the  $100(1-\alpha)\%$  confidence interval  
for  $\mu \Leftrightarrow$  the hypothesis  
test accepts.

i.e.

The confidence interval consists  
exactly of the values of  $\mu_0$  for which  
the Null hypothesis  $H_0 \mu = \mu_0$  is  
accepted.

More generally

if  $\theta$  is a parameter of a family of probability distributions and

the possible values of  $\theta$  are  $\Theta$

let  $\vec{X}$  denote a vector of data

from one such distribution.

Suppose that for each  $\theta_0 \in \Theta$

there is a test at level  $\alpha$  of

$$H_0: \theta = \theta_0.$$

Let  $A(\theta_0)$  be the acceptance region of the test (of  $H_0: \theta = \theta_0$ ).

Then

$$C(\vec{X}) = \{ \theta : \vec{X} \in A(\theta) \}$$

is  $100(1-\alpha)\%$  confidence region for  $\theta$ .

then

$$A(\theta_0) = \{ \vec{X} \mid \theta_0 \in C(\vec{X}) \}$$

is an acceptance region for  
a test at level  $\alpha$  of

$$H_0: \theta = \theta_0.$$

because

$$\begin{aligned} P(X \in A(\theta_0) \mid \theta = \theta_0) &= P(\theta_0 \in C(\vec{X}) \mid \theta = \theta_0) \\ &= 1 - \alpha. \end{aligned}$$

i.e.

The hypothesis that  $\theta = \theta_0$   
is accepted if  $\theta_0$  lies in  
the confidence region.

In other words

A  $100(1-\alpha)\%$  confidence region for  $\theta$  consists of all values of  $\theta_0$  for which the hypothesis  $\theta = \theta_0$  is not rejected at level  $\alpha$ .

The reason is that

$$P(X \in A(\theta_0) | \theta = \theta_0) = 1 - \alpha.$$

and

$$P[\theta_0 \in C(X) | \theta = \theta_0] = P(X \in A(\theta_0) | \theta = \theta_0) \\ = 1 - \alpha.$$

On the other hand,

if  $(\bar{X})$  is  $100(1-\alpha)\%$  confidence region for  $\theta$

i.e. for each  $\theta_0$ .

$$P[\theta_0 \in C(\bar{X}) | \theta = \theta_0] = 1 - \alpha$$

e.g. Confidence intervals for the Mann  
Whitney Test

$$X's \sim F$$

$$Y's \sim G$$

Suppose  $G(x) = F(x - \Delta)$

i.e. the treated group (Y's)  
is different from the control group  
by an additive effect  $\Delta$ .

We will find a confidence interval  
(100(1- $\alpha$ )%) for  $\Delta$  by testing

$$H_0: \Delta = \Delta_0 \quad \text{at level } \alpha$$

and noting which values of  $\Delta_0$  allow  
 $H_1$  to be accepted.

Recall that

$$U_T = \#\{(i, j) : (X_i - Y_j) < 0\}$$

was used to test the Null

$$H_0: F = G.$$

Under the Null.

$$H_0: \Delta = \Delta_0,$$

$$U_Y(\Delta_0) = \# \{ (i, j) : (X_i - (Y_j - \Delta_0)) < 0 \}$$

(will have the same distribution as  $U_Y$  under  $H_0: \Delta = 0$ .)

$$\rightarrow = \# \{ (i, j) : Y_j - X_i > \Delta_0 \}$$

In particular

$$P\left(U_Y(\Delta_0) = \frac{mn}{2} + k\right) = P\left(U_Y(\Delta_0) = \frac{mn}{2} - k\right)$$

for all integers  $k$ .

If  $k = k(\alpha)$  is chosen so that

$$P(k \leq U_Y(\Delta_0) \leq mn - k) = 1 - \alpha$$

then the test at level  $\alpha$  accepts  $H_0: \Delta = \Delta_0$  for data such that

$$k \leq U_Y(\Delta_0) \leq mn - k.$$

So a  $100(1-\alpha)\%$  confidence interval for  $\Delta$  is.

$$C = \{ \Delta : k \leq U_Y(\Delta) \leq mn - k \}$$

and this is given by:

$$C = [D_{(k)}, D_{(mn-k+1)}]$$

where

the  $D_{(1)}, \dots, D_{(m,n)}$

are the ordered differences.

because if  $\Delta = D_{(k)}$

$$\begin{aligned} U_Y(\Delta) &= \#(X_i - Y_j + \Delta < 0) \\ &= \#(Y_j - X_i > \Delta) \\ &= mn - k. \end{aligned}$$

and if  $\Delta = D_{(mn-k+1)}$  then

$$U_Y(\Delta) = \#(X_j - Y_i \geq \Delta) = k.$$



Note that a "shift" model for a treatment effect is often overly simplistic!

---

examples on the computer.