

## Solution to review problem 5

5) Random samples of size 10 from two independent normal populations, each having variance 5, yield  $\bar{x}_1 = 4$  and  $\bar{x}_2 = 4.9$ .

(a) Test the null hypothesis  $\mu_1 \geq \mu_2$  at the 5% level.

(b) If in fact  $\mu_1 = \mu_2 - 1$ , what is the probability of a type II error with the test you used in (a)?

More generally, consider any two independent normal populations having the same (known) variance  $\sigma$ , and having unknown means  $\mu_1$  and  $\mu_2$ . Let  $x_1, \dots, x_n$  be a random sample from the first population, and let  $y_1, \dots, y_n$  be a random sample from the second population (I am assuming that these samples have the same size  $n$ ). We are testing the null hypothesis  $H_0 : \mu_1 \geq \mu_2$  against the alternative hypothesis  $H_1 : \mu_1 < \mu_2$ . Since this null hypothesis is composite, the results from chapters 8 and 11 don't give a way to test it. The only method we have for testing composite null hypotheses is the likelihood ratio test. The critical regions for this test have the form  $\lambda < k$ , where  $0 < k < 1$  and where  $\lambda$  is the likelihood ratio statistic (which is a function of the  $x_i$ 's and  $y_j$ 's). So we must compute  $\lambda$ .

To do this, first write out the joint density function for  $x_1, \dots, x_n, y_1, \dots, y_n$ : since as usual the  $x_i$ 's and  $y_j$ 's are independent, this is

$$\begin{aligned} f(x_1, \dots, x_n, y_1, \dots, y_n; \mu_1, \mu_2) &= \prod_{i=1}^n \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x_i - \mu_1}{\sigma}\right)^2} \prod_{j=1}^n \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{y_j - \mu_2}{\sigma}\right)^2} \\ &= \left(\frac{1}{\sigma\sqrt{2\pi}}\right)^{2n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n ((x_i - \mu_1)^2 + (y_i - \mu_2)^2)}. \end{aligned}$$

Recall that

$$\lambda = \frac{\max_{(\mu_1, \mu_2) \in H_0} f(x_1, \dots, x_n, y_1, \dots, y_n; \mu_1, \mu_2)}{\max_{(\mu_1, \mu_2) \in H_0 \cup H_1} f(x_1, \dots, x_n, y_1, \dots, y_n; \mu_1, \mu_2)}.$$

We're viewing all the  $x_i$ 's and  $y_j$ 's as being fixed, so for simplicity write

$$L(\mu_1, \mu_2) = f(x_1, \dots, x_n, y_1, \dots, y_n; \mu_1, \mu_2).$$

First let's compute the maximum that occurs in the denominator of  $\lambda$ , so we're maximizing  $L(\mu_1, \mu_2)$  over all real numbers  $\mu_1, \mu_2$ . The maximum

occurs for the same pair  $(\mu_1, \mu_2)$  which maximizes

$$\log L(\mu_1, \mu_2) = -2n \log(\sigma\sqrt{2\pi}) - \frac{1}{2\sigma^2} \sum_{i=1}^n ((x_i - \mu_1)^2 + (y_i - \mu_2)^2).$$

Taking the partial derivative with respect to  $\mu_1$ , we find that  $\log L(\mu_1, \mu_2)$  is maximized (for fixed  $\mu_2$ ) when  $\mu_1 = \bar{x}$ ; likewise, for fixed  $\mu_1$  the maximum occurs when  $\mu_2 = \bar{y}$ . Thus the denominator of  $\lambda$  is just  $L(\bar{x}, \bar{y})$ .

Now consider the numerator of  $\lambda$ , so we're maximizing  $L(\mu_1, \mu_2)$  over all numbers  $\mu_1, \mu_2$  such that  $\mu_1 \geq \mu_2$ . First let's check that the maximum over *all*  $\mu_1, \mu_2$  doesn't occur in this region, i.e., that the values of  $\mu_1$  and  $\mu_2$  which give rise to this maximum should not satisfy  $\mu_1 \geq \mu_2$  (if they satisfied this, then the numerator and denominator of  $\lambda$  would be the same, so  $\lambda = 1$  and this test won't be of any use). Well, the maximum over *all*  $\mu_1, \mu_2$  occurs when  $\mu_1 = \bar{x}$  and  $\mu_2 = \bar{y}$ , and in our specific example indeed  $\bar{x} = 4 < 4.9 = \bar{y}$ . So that's good.

In order to maximize  $L(\mu_1, \mu_2)$  over the region  $\mu_1 \geq \mu_2$ , let's introduce a new variable  $\nu = \mu_1 - \mu_2$ , so  $\mu_1 = \nu + \mu_2$ ; then our region becomes just  $\nu \geq 0$ , which is easier to visualize. Rewrite  $\log L(\mu_1, \mu_2)$  in terms of  $\nu$  and  $\mu_2$ :

$$\log L(\mu_1, \mu_2) = -2n \log(\sigma\sqrt{2\pi}) - \frac{1}{2\sigma^2} \sum_{i=1}^n ((x_i - \nu - \mu_2)^2 + (y_i - \mu_2)^2).$$

We want to maximize this number (call it  $g$ ) over all  $\mu_2$  and all  $\nu \geq 0$ . First set  $\frac{\partial g}{\partial \mu_2}$  to zero, and find that  $g$  is maximized (for fixed  $\nu$ ) when  $\mu_2 = (\bar{x} + \bar{y} - \nu)/2$ . Now substitute this value of  $\mu_2$  into the expression for  $g$ , to get

$$g = -2n \log(\sigma\sqrt{2\pi}) - \frac{1}{2\sigma^2} \sum_{i=1}^n \left( \left( x_i - \frac{\bar{x} + \bar{y} + \nu}{2} \right)^2 + \left( y_i - \frac{\bar{x} + \bar{y} - \nu}{2} \right)^2 \right).$$

Now we need to maximize this last expression over all  $\nu \geq 0$ . Compute

$$\begin{aligned} \frac{\partial g}{\partial \nu} &= -\frac{1}{2\sigma^2} \sum_{i=1}^n \left( -\left( x_i - \frac{\bar{x} + \bar{y} + \nu}{2} \right) + \left( y_i - \frac{\bar{x} + \bar{y} - \nu}{2} \right) \right) \\ &= -\frac{1}{2\sigma^2} \left( -\left( n\bar{x} - \frac{n}{2}(\bar{x} + \bar{y} + \nu) \right) + \left( n\bar{y} - \frac{n}{2}(\bar{x} + \bar{y} - \nu) \right) \right) \\ &= \frac{n}{2\sigma^2} (\bar{x} - \bar{y} - \nu), \end{aligned}$$

which is negative because  $\bar{x} - \bar{y} = 4 - 4.9 < 0$  and  $-\nu \leq 0$ . Thus,  $g$  is a decreasing function of  $\nu$  when  $\nu \geq 0$ , so  $g$  is maximized (in this range) when  $\nu = 0$ . Recalling that  $\mu_2 = (\bar{x} + \bar{y} - \nu)/2$  and  $\mu_1 = \nu + \mu_2$ , this says the numerator of  $\lambda$  is  $L(\frac{\bar{x}+\bar{y}}{2}, \frac{\bar{x}+\bar{y}}{2})$ .

Now we've computed  $\lambda$ , and we just need to simplify the resulting expression. We have

$$\begin{aligned}\lambda &= \frac{L(\frac{\bar{x}+\bar{y}}{2}, \frac{\bar{x}+\bar{y}}{2})}{L(\bar{x}, \bar{y})} \\ &= e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n ((x_i - \frac{\bar{x}+\bar{y}}{2})^2 + (y_i - \frac{\bar{x}+\bar{y}}{2})^2 - (x_i - \bar{x})^2 - (y_i - \bar{y})^2)},\end{aligned}$$

and by writing  $(x_i - \frac{\bar{x}+\bar{y}}{2})^2 = ((x_i - \bar{x}) + \frac{\bar{x}-\bar{y}}{2})^2$  and expanding, we get

$$\lambda = e^{-\frac{(\bar{x}-\bar{y})^2}{4\sigma^2}}.$$

The critical regions are  $\lambda \leq k$  where  $0 < k < 1$ ; equivalently (taking logs), these are

$$\frac{-(\bar{x} - \bar{y})^2}{4\sigma^2} < \log k,$$

or equivalently

$$(\bar{x} - \bar{y})^2 > -4\sigma^2 \log k.$$

In the computation of  $\lambda$ , we assumed that  $\bar{x} < \bar{y}$ , so the critical region can be rewritten as

$$\bar{y} - \bar{x} > \sigma \sqrt{-\log k},$$

or equivalently  $\bar{y} - \bar{x} > c$  for some constant  $c > 0$ .

Now we must determine how the size  $\alpha$  of the critical region depends on  $c$ . For this we can use the result of exercise 8.3 (which I believe I proved in class), namely that  $\bar{y} - \bar{x}$  has normal distribution with mean  $\mu_2 - \mu_1$  and variance  $2\sigma^2/n$ . Then

$$z := \frac{\bar{y} - \bar{x} - (\mu_2 - \mu_1)}{\sigma \sqrt{2/n}}$$

has standard normal distribution, so by definition

$$P(z > z_\alpha) = \alpha,$$

and thus

$$P(\bar{y} - \bar{x} > \mu_2 - \mu_1 + z_\alpha \sigma \sqrt{2/n}) = \alpha.$$

But this depends on  $\mu_2 - \mu_1$ , and we need a critical region that only depends on the  $x_i$ 's and  $y_j$ 's. Note that, for any specific constant  $\delta \leq 0$ , we have computed the critical region of size  $\alpha$  for testing the same alternative hypothesis against the more specific null hypothesis  $\mu_2 - \mu_1 = \delta$ . To find a critical region that simultaneously works for all values of  $\delta$ , we take the intersection of all corresponding critical regions

$$\bar{y} - \bar{x} > \delta + z_\alpha \sigma \sqrt{2/n},$$

getting

$$\bar{y} - \bar{x} > z_\alpha \sigma \sqrt{2/n}.$$

Since we took the intersection of the different critical regions, it follows that when the null hypothesis holds, the probability that  $\bar{y}$  and  $\bar{x}$  satisfy the above inequality is at most  $\alpha$ ; and moreover, if actually  $\delta = 0$ , or equivalently  $\mu_2 = \mu_1$ , then the probability equals  $\alpha$ .

We have shown that the critical region of size  $\alpha$  produced by the likelihood ratio test is

$$\bar{y} - \bar{x} > z_\alpha \sigma \sqrt{2/n}.$$

Now consider the specific problems that were asked. There  $n = 10$  and  $\sigma^2 = 5$ , so the critical region of size 0.05 is

$$\bar{y} - \bar{x} > z_{.05} \sqrt{5} \sqrt{2/10} = z_{.05} = 1.65.$$

This critical region does not include the specific pair  $(\bar{x}, \bar{y}) = (4, 4.9)$ , so we must accept the null hypothesis.

Next suppose  $\mu_1 = \mu_2 - 1$ . Then the probability of a type II error is

$$P(\bar{y} - \bar{x} < 1.65).$$

As noted above,

$$z := \frac{\bar{y} - \bar{x} - (\mu_2 - \mu_1)}{\sigma \sqrt{2/n}}$$

has standard normal distribution, and here  $\mu_2 - \mu_1 = 1$  and  $\sigma \sqrt{2/n} = 1$ , so  $z = \bar{y} - \bar{x} - 1$ . Thus

$$P(\bar{y} - \bar{x} < 1.65) = P(z < 0.65) = 0.7422.$$