

LECTURE - (27)

Agenda:

- (1) ~~p-values~~ p-values
- (2) Examples

Consider a general hypothesis testing problem concerning two hypotheses H_0 and H_A about an unknown parameter θ . Suppose we decide that $\hat{\theta}$ is the appropriate test statistic. Then, given that we want to test at level α , we construct an appropriate rejection region. Depending on whether $\hat{\theta}_{\text{observed}}$ (observed value of $\hat{\theta}$) lies in the rejection region or not, we reject or accept H_0 . NOTE THAT THE REJECTION REGION DEPENDS ON THE CHOICE OF THE LEVEL (α).

Hence, it is possible that if Person A and Person B analyze the same data, but Person A chooses level 0.05, whereas Person B chooses level 0.01, they reach different conclusions. Keeping this fact in mind, the statistical community devised a more informative way of reporting the conclusions of a testing procedure. Instead of reporting ~~"Based on the observed data, H_0 was rejected (or accepted) at level α "~~, we report the

"p-value" of the testing procedure.

DEFINITION: If $\hat{\theta}$ is a test statistic, then

the p-value of the corresponding test is defined as the smallest level α for which ~~we~~ we will reject the null hypothesis based on the observed data.

Hence, if ~~the~~ the level of the test is ^{chosen to be} ~~is~~ ^{would} larger than the p-value, we ^{would} reject the null hypothesis H_0 .

On the other hand, if the level of the test is chosen to be smaller than the p-value, we would accept the null hypothesis H_0 .

THE SMALLER THE p-value, STRONGER IS THE EVIDENCE THAT THE NULL HYPOTHESIS SHOULD BE REJECTED.

Instead of just arbitrarily fixing the level of the test and reporting acceptance or rejection, the p-value allows ~~the~~ the client or the reader to evaluate the extent to which the observed data disagree with the null hypothesis.

In some sense, rather than ^{just} saying who won or who lost, the p-value also gives us the idea about the margin of victory.

RESULT: If the rejection region is of the form $\hat{\theta} \leq k$, then ~~then~~ if $\hat{\theta}_{\text{observed}}$ denotes the observed value of $\hat{\theta}$,

$$p\text{-value} = P(\hat{\theta} \leq \hat{\theta}_{\text{observed}} | H_0 \text{ is true})$$

RESULT: If the rejection region is of the form $\hat{\theta} \geq k$, then if $\hat{\theta}_{\text{observed}}$ denotes the observed value of $\hat{\theta}$,

$$p\text{-value} = P(\hat{\theta} \geq \hat{\theta}_{\text{observed}} | H_0 \text{ is true})$$

Example: Suppose the ^{mean} lifetime μ for a population of bulbs is believed to be 18 time units. A researcher claims that due to improvement in the quality of production, the new mean lifetime is greater than 18. The data $\gamma_1, \gamma_2, \dots, \gamma_{100}$ was gathered for the lifetime of 100 randomly chosen bulbs. If the observed $\bar{\gamma}$ is 20 and S is 14, provide the p-value of the standard testing procedure.

Note that we want to test

$$H_0: \mu = 18 \quad \text{v.s.} \quad H_A: \mu > 18.$$

TEST STATISTIC: \bar{Y}

REJECTION REGION: Reject for large values of \bar{Y} .

Hence,

$$p\text{-value} = P(\bar{Y} > \bar{Y}_{\text{observed}} \mid H_0 \text{ is true})$$

$$= P(\bar{Y} > 20 \mid \mu = 18)$$

$$= P\left(\frac{\bar{Y} - 18}{S} > \frac{20 - 18}{14} \mid \mu = 18\right)$$

$$= P\left(\frac{\sqrt{200}(\bar{Y} - 18)}{S} > \sqrt{100}\left(\frac{20 - 18}{24}\right) \mid \mu = 18\right)$$

$$\approx P(\text{Normal}(0, 1) > \frac{10}{7})$$

$$= 1 - \Phi\left(\frac{10}{7}\right)$$