Neyman-Pearson Lemma

Lemma 1.1 Let X be a random variable with unknown distribution, either $f_0(x)$ or $f_1(x)$, and consider the test situation

$$H_0: X \sim f_0(x) \text{ versus } H_1: X \sim f_1(x) \tag{1}$$

(simple versus simple). To test H_0 versus H_1 , the most powerful size α test is

$$\varphi(x) = \begin{cases} 1 & x \ni f_0(x)/f_1(x) < k_\alpha \\ \gamma_\alpha & = k_\alpha \\ 0 & > k_\alpha \end{cases}$$
(2)

where k_{α} and γ_{α} are chosen such that $E_{f_0}(\varphi(x)) = \alpha$.

This lemma is the basis for future testing where we move on to composite hypotheses.

Proof 1.1 Consider two tests, φ and φ^* . The difference of the powers of these tests is

$$(1 - \beta_{\varphi}) - (1 - \beta_{\varphi^*}) = \mathcal{E}_{f_1}(\varphi) - \mathcal{E}_{f_1}(\varphi^*)$$
(3)

$$= E_{f_1} \left(\varphi - \varphi^* \right) \tag{4}$$

$$= \int_{\mathfrak{X}} \left(\varphi(x) - \varphi^*(x)\right) f_1(x) dx \tag{5}$$

$$= \int_{x \ge f_0/f_1 < k} (\varphi(x) - \varphi^*(x)) f_1(x) dx + \int_{x \ge f_0/f_1 = k} (\varphi(x) - \varphi^*(x)) f_1(x) dx + \int_{x \ge f_0/f_1 > k} (\varphi(x) - \varphi^*(x)) f_1(x) dx$$
(6)

$$\geq \frac{1}{k} \int_{x \ni f_0/f_1 < k} (\varphi(x) - \varphi^*(x)) f_0(x) dx + \frac{1}{k} \int_{x \ni f_0/f_1 = k} (\varphi(x) - \varphi^*(x)) f_0(x) dx$$

$$+\frac{1}{k} \int_{x \ni f_0/f_1 > k} \left(\varphi(x) - \varphi^*(x)\right) f_0(x) dx$$
(7)

$$\geq \frac{1}{k_{\alpha}} \mathcal{E}_{f_0} \left(\varphi(X) - \varphi^*(X) \right) \tag{8}$$

$$\geq 0$$
 (9)

Thus, the power of φ is no less than the power of φ^* .

The transition between alternative and null hypotheses in equations 6 and 7 is due to the fact that in the first integral we are in the rejection region so $\varphi - \varphi^* = 1 - \varphi^* \ge 0$ and $f_0/f_1 < k$ means that $f_1 \ge f_0/k$. Similarly, in the middle integral we are in the randomization region so $f_0/f_1 = k$ means that $f_1 = f_0/k$. Finally, for the last integral we are in the "acceptance" region so $\varphi - \varphi^* = 0 - \varphi^* \le 0$ and $f_0/f_1 > k$ means that $f_1 \ge f_0/k$.

Equation 9 follows from 8 follows since $k_{\alpha} \geq 0$ and under H_0 we have that $E_0(\varphi(X)) \geq E_0(\varphi^*(X))$ or $\alpha \geq \alpha^*$ by size.

Example 1.1 Let $X_i \stackrel{iid}{\sim} B(1,p)$. To test

$$H_0: p = 0.5 \ versus \ H_1: p = 0.75$$

we look at

$$\frac{f_0(x)}{f_1(x)} = \frac{1^{\sum x_i} 1^{5 - \sum x_i} / 2^5}{3^{\sum x_i} 1^{5 - \sum x_i} / 4^5}$$
(10)

$$= \frac{2^5}{3\Sigma^{x_i}} \tag{11}$$

To make $\alpha = 0.5$ we note that

$$\frac{f_0(x)}{f_1(x)} = \frac{2^5}{3\Sigma^{x_i}} \tag{12}$$

is decreasing in $\sum x_i$. So we will reject H_0 if $T = \sum X_i$ is too large. Now note that $T = \sum X_i \sim B(5, p)$ so that under the null hypothesis

$$f_0(t) = {\binom{5}{t}} \left(\frac{1}{2}\right)^t \left(\frac{1}{2}\right)^{5-t}$$
(13)

Thus

If we reject for $\sum x_i = 5$ $(f_0/f_1 = 0.1317)$ then we reject with probability 0.0313 under H_0 (not enough). Using $\sum x_i \ge 4$ $(f_0/f_1 = .3951)$ we reject with probability 0.1563 + 0.0313 = 0.1876 (too much). So, we should reject if $\sum x_i = 5$ and randomize for $\sum x_i = 4$. To get exact size α we need γ_{α} .

Note that

$$E_{1/2}(\varphi(X)) = P_{1/2}[reject H_0]$$
(14)

$$= \gamma P_{1/2} \left(\sum X_i = 4 \right) + P_{1/2} \left(\sum X_i = 5 \right)$$
(15)

$$= \gamma 0.1563 + 0.0313 \tag{16}$$

$$= 0.05$$
 (17)

Solving for γ we obtain $\gamma = 0.1196$ and our level $\alpha = 0.05$ test is

$$\varphi\left(\mathbf{X}\right) = \begin{cases} 1 & \sum_{i=5}^{N_{i}} X_{i} = 5\\ 0.1196 & \sum_{i=1}^{N_{i}} X_{i} = 4\\ 0 & else \end{cases}$$
(18)

Finally, note that if $p = \frac{3}{4}$, the power of this test is

$$1 - \beta = \left(\frac{3}{4}\right)^5 + \gamma {\binom{5}{4}} \left(\frac{3}{4}\right)^4 \left(\frac{1}{4}\right)$$
(19)

$$= 0.2373 + \gamma 0.3955 \tag{20}$$

$$= 0.2846$$
 (21)

which is pretty low. We might want to use a few more observations.