Math 2a

## Math 2a Homework 8 Solutions

**Problem 1.** For x = 0, 1, ..., 10, we compute the likelihood ratio

$$f_0(x)/f_1(x) = \frac{\binom{10}{x} \times 0.6^x \times 0.4^{10-x}}{\binom{10}{x} \times 0.7^x \times 0.3^{10-x}} = \left(\frac{4}{3}\right)^{10} \times \left(\frac{9}{14}\right)^x.$$

Numerically, this gives

The ratio decreases as x increases. So the likelihood ratio test rejects for larger values of x. The rejection region is of the form  $\{x \ge n\}$ . The corresponding significance level is the probability that we reject the null hypothesis when it is true, namely,  $P_{0.6}(X \ge n)$  i.e. the probability that a binomial random variable with 10 trials and a probability of success equal to .6 is greater or equal to n. The values are given below (as a function of the cut-off n).

| n  | significance level |
|----|--------------------|
| 10 | 0.006046618        |
| 9  | 0.046357402        |
| 8  | 0.167289754        |

Problem 2. (a) The mean lifetime of a battery is equal to

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$$\int_0^\infty t f(t|\lambda) dt = \int_0^\infty \frac{t}{\lambda} e^{-t/\lambda} dt = \lambda \int_0^\infty s e^{-s} ds,$$

where we set  $s = t/\lambda$ . Since

$$\int_0^\infty se^{-s}ds = -(s+1)e^{-s}|_0^\infty = 0 - (-1) = 1,$$

we see that the mean lifetime of a battery is  $\lambda$ .

(b) The likelihood ratio is

$$\frac{\prod_{1}^{n} f(T_{i}|1)}{\prod_{1}^{n} f(T_{i}|1.5)} = \frac{\prod_{1}^{n} (e^{-T_{i}/1}/1)}{\prod_{1}^{n} (e^{-T_{i}/1.5}/1.5)}$$
$$= 1.5^{n} \exp(-\sum_{1}^{n} T_{i} + \sum_{1}^{n} T_{i}/1.5) = 1.5^{n} \exp(-\sum_{1}^{n} T_{i}/3).$$

The ratio decreases as  $\sum T_i = n\overline{T}$  increases. So the likelihood ratio test rejects for larger values of  $\overline{T}$ , which means that the rejection is of the form  $\{\overline{T} \ge C\}$ . The corresponding significance level is

$$P_1\left(\sum T_i \ge nC\right) = \int_{\sum t_i \ge nC} \prod f(t_i|1) dt_1 \dots dt_n$$

$$= \int_{t_i \ge 0, \sum t_i \ge nC} e^{-\sum t_i} dt_1 \dots dt_n$$

If we set  $S = \sum t_i$  and  $y_i = t_i/S$ , then  $dt_1 \dots dt_n = S^{n-1} dS dy_1 \dots dy_{n-1}$ . And  $(y_1, \dots, y_{n-1})$  runs over the region  $y_i \ge 0$ ,  $\sum_{1}^{n-1} y_i \le 1$ . So the above formula is equal to

$$\int_{nC}^{\infty} \int_{y_i \ge 0, \sum_{1}^{n-1} y_i \le 1} S^{n-1} e^{-S} dy_1 \dots dy_{n-1} dS = \int_{nC}^{\infty} \frac{S^{n-1}}{(n-1)!} e^{-S} dS$$

By integration by parts, we compute the significance level to be equal to  $e^{-nC} \sum_{k=0}^{n-1} \frac{(nC)^k}{k!}$ .

**Problem 3.** (a) For i = 1, ..., 16, let  $x_i$  denote the number of cans of beer the *i*-th student drank, and  $y_i$  be the corresponding BAC number. We compute  $\overline{x} = \sum_{1}^{16} x_i/16 = 4.8125$  and  $\overline{y} = \sum_{1}^{16} y_i/16 = 0.07375$ . Thus,

$$s_x = \sqrt{\sum_{1}^{16} (x_i - \overline{x})^2 / (16 - 1)} = 2.197536;$$
  

$$s_y = \sqrt{\sum_{1}^{16} (y_i - \overline{y})^2 / (16 - 1)} = 0.04414;$$
  

$$r = \frac{\sum_{1}^{16} (x_i - \overline{x}) \times (y_i - \overline{y})}{(16 - 1) \times s_x \times s_y} = 0.894338.$$

It then follows that the slope of the regression line is

$$b_1 = r \frac{s_y}{s_x} = 0.017964$$

and the intercept is

$$b_0 = \overline{y} - b_1 \overline{x} = -0.0127.$$

Finally,  $r^2 = 0.799841$  and the equation of the regression line is given by

$$y = -0.0127 + 0.017964x.$$

(b) Let  $\rho$  denote the correlation between x and y. We will test

$$H_0: \rho = 0$$
 versus  $H_a: \rho > 0$ .

Compute the *t*-statistic:

$$t = \frac{r\sqrt{16-2}}{\sqrt{1-r^2}} = 7.479592.$$

In terms of a random variable T having t(16-2) distribution, the P-value for the test is

$$P(T \ge t) = 1.48 \times 10^{-6},$$

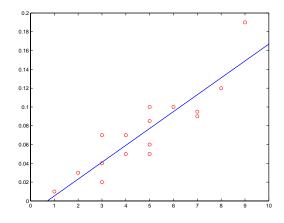


Figure 1: Regression plot

which is very small. The conclusion is that there is very strong evidence that drinking more beers increases blood alcohol.

**Problem 4.** (a) Let y = interval until the next eruption and x = duration of an eruption, both in minutes, for all eruptions of Old Faithful Geyser over the 8-day period. The general linearity of the scatter plot of time-between(y) versus duration (x) suggests use of a simple linear regression model,

$$y = \beta_0 + \beta_1 x + \text{ error}$$

The fitted model is

$$y = b_0 + b_1 x,$$

where

$$b_1 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = 10.3582,$$
  
$$b_0 = \bar{y} - b_1 \bar{x} = 33.9668.$$

For reference,  $\bar{x} = 3.57613$  and  $\bar{y} = 71.0090$ .

(b) We have made n = 222 observations. Suppose we have just observed x = d, we want to predict y. Note that the **mean value** of the interval until the next eruption is  $\mathbf{E}y = \beta_0 + \beta_1 d$ , so we use  $b_0 + b_1 d$  as a (least square) estimator of  $\mathbf{E}y$ . In class, we showed that

$$t = \frac{b_0 + b_1 d - (\beta_0 + \beta_1 d)}{s \sqrt{\frac{1}{n} + \frac{(d - \bar{x})^2}{\sum (x_i - \bar{x})^2}}}$$

is a student's t distribution with n - 2 = 220 degrees of freedom; here,

$$s = \frac{RSS}{n-2} = \frac{\sum (y_i - b_0 - b_1 x_i)^2}{n-2}$$

is our estimate of the std deviation of the errors. Therefore, a 95% confidence interval for Ey is

$$b_0 + b_1 d \pm s \sqrt{\frac{1}{n} + \frac{(d - \bar{x})^2}{\sum (x_i - \bar{x})^2}} \cdot 1.9708,$$

where 1.9708 is such that  $P(|T| \le 1.9708) = 95\%$ , where T has t(220) distribution. As an aside, recall the key intermediate results

$$\mathbf{E}(b_0 + b_1 d) = \beta_0 + \beta_1 d, \quad \text{Var}(b_0 + b_1 d) = \sigma^2 \Big[ \frac{1}{n} + \frac{(d - \bar{x})^2}{\sum (x_i - \bar{x})^2} \Big].$$

(c) Let  $y = \beta_0 + \beta_1 d + \epsilon$  be a new observation. In this case,

$$y - (b_0 + b_1 d) = \epsilon + (\beta_0 + \beta_1 d) - (b_0 + b_1 d)$$

Thus  $y - (b_0 + b_1 d)$  is a normal random variable with mean 0 and variance  $\sigma^2 \left[1 + \frac{1}{n} + \frac{(d-\bar{x})^2}{\sum (x_i - \bar{x})^2}\right]$ . As in part (b), we have

$$\frac{y - \left(b_0 + b_1 d\right)}{s\sqrt{1 + \frac{1}{n} + \frac{(d - \bar{x})^2}{\sum (x_i - \bar{x})^2}}}$$

is a student's t distribution with 220 degrees of freedom. Therefore, a 95% confidence **prediction** interval for y is

$$b_0 + b_1 d \pm s \sqrt{1 + \frac{1}{n} + \frac{(d - \bar{x})^2}{\sum (x_i - \bar{x})^2}} \cdot 1.9708$$

(d) We compute s = 6.15853. For the case d = 4.0,  $b_0 + b_1 d = 75.39958$ , and

$$s\sqrt{\frac{1}{n} + \frac{(d-\bar{x})^2}{\sum(x_i - \bar{x})^2}} \cdot 1.9708 = 0.874932,$$
$$s\sqrt{1 + \frac{1}{n} + \frac{(d-\bar{x})^2}{\sum(x_i - \bar{x})^2}} \cdot 1.9708 = 12.16902.$$

So the numerical answers to (b) and (c) are [74.52465, 76.27451] and [63.23056, 87.56860], respectively.

**Appendix.** There are a total of 222 observations in total in the dataset. Some of them are not in the 8-day period. If we used the whole dataset, we would obtain the following results: The fitted line is

$$y = 33.96676 + 10.35821x.$$

When d = 4, a 95% confidence interval for the mean is

$$75.39958 \pm 0.87493 = [74.52465, 76.27451],$$

and a 95% prediction interval for y is

$$75.39958 \pm 12.16877 = [63.23081, 87.56835].$$

Some intermediate results are:  $\bar{x} = 3.57613$ ,  $\bar{y} = 71.00901$ , s = 6.15853, and the C such that  $P(|T| \le C) = 95\%$ ) for a t distribution with 220 degrees of freedom is C = 1.97081.