

**STA 410/2102F, Fall 2007: In-Class Test**  
**SOLUTIONS**

1. [5 points] Write (with full explanation) the number  $-19.25$  in double-precision floating-point format.

**Solution:**  $-19.25 = -(16 + 2 + 1 + \frac{1}{4}) = -(2^4 + 2^1 + 2^0 + 2^{-2}) = -2^4(2^0 + 2^{-3} + 2^{-4} + 2^{-6}) = (-1)^1 2^4 1.001101 = (-1)^1 2^{1027-1023} 1.001101.$

2. [8 points] Determine (with full explanation) what value R will return if given the following expression:  $1 + 2^{(-10)} + 2^{50} - 2^{50}$ .

**Solution:**  $1 + 2^{-10} = (-1)^0 2^0 1.0000000001$  (base 2).

*Then, adding  $2^{50}$  gives  $2^0 1.0000000001 + 2^{50} 1.0 = 2^{50}(2^{-50} 1.0000000001 + 1.0) = 2^{50}(0.[49 \text{ zeroes}]1[9 \text{ zeroes}]1 + 1.0) = 2^{50} 1.[49 \text{ zeroes}]1[9 \text{ zeroes}]1$ , which truncates to  $2^{50} 1.[49 \text{ zeroes}]1$  since R only holds 52 decimal points.*

*Then, subtracting  $2^{50}$  gives  $2^{50} 1.[49 \text{ zeroes}]1 - 2^{50} 1.0 = 2^{50}(1.[49 \text{ zeroes}]1 - 1.0) = 2^{50}(0.[49 \text{ zeroes}]1) = 2^{50}(2^{-50} 1.0) = 2^0 1.0$ , which equals 1.*

*So, the answer is 1.*

3. [7 points] Write a complete R program to estimate the quantity  $\mathbf{E}[Y/(1 + Z^4)]$ , where  $Y \sim \text{Poisson}(5)$  and  $Z \sim \text{Normal}(0, 1)$ , by a Monte Carlo simulation, using 1000 replications. Your program should output both the estimate and its standard error. [Hint: Recall the R commands “rpois(n, lambda)” to generate n independent Poisson(lambda) random variables, and “rnorm(n)” to generate n independent standard normal random variables.]

**Solution:**

```
n = 1000
yvec = rpois(n,5)
zvec = rnorm(n)
estimate = mean(yvec/(1+zvec^4))
se = sd(yvec/(1+zvec^4)) / sqrt(n)
cat("estimate equals", estimate, "\n")
cat("standard error equals", se, "\n")
```

4. Consider the function  $g(x) = x^2 - 5$ . Suppose we begin with the initial interval  $[0, 4]$ . [You may assume that  $g(0) < 0$  and  $g(4) > 0$ . And, in what follows, you do not need to

write a computer program, you just need to explain what intervals will arise.]

(a) [5 points] Present (with full explanation) the next two intervals computed by the Bisection Method.

**Solution:** *First midpoint*  $= (0+4)/2 = 2$ .

*And,*  $g(2) = 2^2 - 5 = 4 - 5 = -1 < 0$ , but  $g(4) > 0$ .

*So, the first new interval is*  $[2, 4]$ .

*Then, second midpoint*  $= (2+4)/2 = 3$ .

*And,*  $g(3) = 3^2 - 5 = 9 - 5 = 4 > 0$ , but  $g(2) > 0$ .

*So, the second new interval is*  $[2, 3]$ .

(b) [5 points] Present (with full explanation) the next one interval computed by the False Position (Safe Bisection) Method.

**Solution:** *The line joining*  $(0, g(0))$  *to*  $(4, g(4))$ , *i.e. joining*  $(0, -5)$  *to*  $(4, 11)$ , *is the line*  $y = 4x - 5$ .

*Its root is the value of*  $x$  *such that*  $4x - 5 = 0$ , *i.e.*  $x = 5/4$ .

$g(5/4) = (5/4)^2 - 5 = (25/16) - 5 < 0$ , but  $g(4) > 0$ .

*So, the new interval is*  $[5/4, 4]$ .

5. [10 points] Suppose it is believed that  $Y = (X + \beta)^2 + \text{error}$ , with  $\beta$  unknown. Assume that  $\mathbf{x}$  and  $\mathbf{y}$  are two vectors in  $\mathbb{R}$ , each of length  $\mathbf{n}$ , corresponding to the observed  $\mathbf{x}$  and  $\mathbf{y}$  values. Write a complete R program which runs the Newton-Raphson Method for 100 iterations, with initial value 2, to compute the least-squares estimate of  $\beta$ . [Hint: before writing the program, you may need to compute some derivatives, etc.; be sure to explain all such computations.]

**Solution:** *The squared-error function is*  $f(\beta) = \sum_{i=1}^n (y_i - (x_i + \beta)^2)^2$ .

*Its derivative is*  $g(\beta) \equiv f'(\beta) = -4 \sum_{i=1}^n (y_i - (x_i + \beta)^2) (x_i + \beta)$ . *Then, the derivative of*  $g$  *is*  $g'(\beta) = -4 \sum_{i=1}^n \left( (y_i - (x_i + \beta)^2) - 2(x_i + \beta)^2 \right) = -4 \sum_{i=1}^n (y_i - 3(x_i + \beta)^2)$ .

*We want to find a critical value of*  $f$ , *i.e. find a root of*  $g$ , *i.e. find*  $\beta$  *such that*  $g(\beta) = 0$ , *using the Newton-Raphson method.*

The R program is as follows:

```
g = function(beta) { - 4 * sum( (y-(x+beta))^2 * (x+beta) ) }
gp = function(beta) { - 4 * sum( y - 3 * (x+beta)^2 ) }
# (Or, could omit "- 4" from both functions, since it cancels.)
betalist = c(2) # the initial value
for (i in 1:100) {
  betaprev = betalist[length(betalist)]
  betanew = betaprev - g(betaprev) / gp(betaprev)
  betalist = c(betalist, betanew)
}
print(betanew)
```

6. [10 points] Suppose we observe three  $(x, y)$  pairs:  $(1, 4)$ ,  $(2, 6)$ , and  $(3, 10)$ . Compute (with full explanation) the cross-validation sum of squares (CVSS) for the unconstrained linear model,  $Y = \beta_1 + \beta_2 X + \text{error}$  with  $\beta_1$  and  $\beta_2$  unknown. [Hint: remember that an unconstrained linear fit of two points passes through both points.]

**Solution:**  $f_{-1}(x)$  is an unconstrained linear fit of the two points  $(2, 6)$  and  $(3, 10)$ , so  $f_{-1}(x)$  passes through  $(2, 6)$  and  $(3, 10)$ . Hence,  $f_{-1}(x) = 4x - 2$ , so  $f_{-1}(1) = 4(1) - 2 = 2$ .

Similarly,  $f_{-2}(x)$  passes through  $(1, 4)$  and  $(3, 10)$ , so  $f_{-2}(x) = 3x + 1$ , so  $f_{-2}(2) = 3(2) + 1 = 7$ .

And,  $f_{-3}(x)$  passes through  $(1, 4)$  and  $(2, 6)$ , so  $f_{-3}(x) = 2x + 2$ , so  $f_{-3}(3) = 2(3) + 2 = 8$ .

Hence,  $CVSS = \sum_{i=1}^3 (y_i - f_{-i}(x_i))^2 = (4 - 2)^2 + (6 - 7)^2 + (10 - 8)^2 = 4 + 1 + 4 = 9$ .

7. [10 points] Consider the estimator  $\hat{\theta}$  which is “the second-largest value”, i.e.  $\hat{\theta}(x_1, \dots, x_n) = \max\{x_i : x_i < \max(x_1, \dots, x_n)\}$ . [For example,  $\hat{\theta}(3, 20, 17, 12, 15) = 17$ .] Suppose we observe, from an unknown distribution, four data values: 5, 7, 8, 10. Compute (with explanation)  $\widehat{\text{Var}}(\hat{\theta})$ , the jackknife estimator of the variance of  $\hat{\theta}$ .

**Solution:**  $\hat{\theta}_{-1} = \hat{\theta}(7, 8, 10) = 8$ . And,  $\hat{\theta}_{-2} = \hat{\theta}(5, 8, 10) = 8$ . And,  $\hat{\theta}_{-3} = \hat{\theta}(5, 7, 10) = 7$ . And,  $\hat{\theta}_{-4} = \hat{\theta}(5, 7, 8) = 7$ . So,  $\hat{\theta}_{\bullet} = \frac{1}{n} \sum_{i=1}^n \hat{\theta}_{-i} = \frac{1}{4}(8+8+7+7) = 7.5$ .

Then,  $\widehat{\text{Var}}(\hat{\theta}) = \frac{n-1}{n} \sum_{i=1}^n (\hat{\theta}_{-i} - \hat{\theta}_{\bullet})^2 = \frac{3}{4}((8-7.5)^2 + (8-7.5)^2 + (7-7.5)^2 + (7-7.5)^2) = \frac{3}{4}(0.25 + 0.25 + 0.25 + 0.25) = 3/4$ .