## Math 128a - Homework 1 - Due Feb 7 at the beginning of class Corrections made on Feb 1 to Questions 1.6 and 4.3

1) In class we saw an example showing that in decimal floating point arithmetic, the computed value of xmid=(xlower+xupper)/2 is not necessarily between xlower and xupper, which would be a problem for the logic in bisection (in 3 decimal digit arithmetic, try xlower = .997 and xupper = .999). We will show that this is impossible in IEEE arithmetic, which is binary. In other words, we will show that in IEEE arithmetic xmid = fl(fl(xlower + xupper)/2) is in the interval [xlower,xupper], assuming overflow does not occur when adding xlower and xupper. (Here fl(a op b) means the floating point result of the operation a op b.

**Part 1.** Using the fact that IEEE arithmetic is correctly rounded, show that it is monotonic, that is if a, b, c, d and x are IEEE floating point numbers then

 $a \le b$  and  $c \le d$  implies  $fl(a+c) \le fl(b+d)$  $a \le b$  and 0 < x implies  $fl(a/x) \le fl(b/x)$ 

(Similar facts hold for subtraction and multiplication, but we will not need these here.)

**Part 2.** Show that fl(2 \* x) = 2 \* x exactly, assuming overflow does not occur.

**Part 3.** Show that  $2 * \text{xlower} \le fl(\text{xlower} + \text{xupper}) \le 2 * \text{xupper}$ .

**Part 4.** Conclude that xlower  $\leq fl(fl(\text{xlower} + \text{xupper})/2) \leq \text{xupper}$ .

Part 5. Where does this argument fail for correctly rounded decimal arithmetic?

**Part 6.** What happens if xlower and xupper are adjacent IEEE floating point numbers?

2) Suppose x is the exact answer to a problem, and  $\hat{x}$  is our approximate answer. In class we defined the absolute error in  $\hat{x}$  as  $|x - \hat{x}|$  and the relative error in  $\hat{x}$  as  $|x - \hat{x}|/|x|$ . In this problem we will explore some simple properties of these error measures.

Write the base  $\beta$  expansion of x > 0 as  $x = .x_1 x_2 \cdots x_n \cdot \beta^{e_x}$ , and the base  $\beta$  expansion of y > 0 as  $y = .y_1 y_2 \cdots y_n \cdot \beta^{e_y}$ . We will say that x and y agree to their leading d base  $\beta$  digits if  $|x - y| < \frac{1}{2}\beta^{\max(e_x, e_y) - d}$ . For example, .1230 and .1226 agree to 3 decimal digits, as do 1.00 and .996, or .1233 and .1237.

- **Part 1.** Suppose you print out  $\hat{x}$  as a base  $\beta$  number. Show that if the relative error  $|x \hat{x}|/|x| < 1$ , then the leading  $\lfloor \log_{\beta} \frac{|x|}{|x \hat{x}|} \rfloor 1$  nonzero base  $\beta$  digits of  $\hat{x}$  are correct, i.e. x and  $\hat{x}$  agree to that many digits. (|x| is the *floor of x*, the largest integer less than or equal to x.)
- **Part 2.** Suppose you have solved your problem and gotten  $\hat{x}$ , and also a bound  $e_{abs} \ge |x \hat{x}|$  on the absolute error (perhaps using rounding error analysis as described in class). You would like a bound  $e_{rel} \ge |x \hat{x}|/|x|$  on the relative e rror. One obvious candidate is  $e_{rel} = e_{abs}/|x|$ , but of course you can't compute this because you don't know x (otherwise we wouldn't need an error bound!). So instead you try  $e_{rel} = e_{abs}/|\hat{x}|$ . Show that it is ok to use  $e_{abs}/|\hat{x}|$  instead of  $e_{abs}/|x|$  by showing that

$$\frac{\frac{|x-\widehat{x}|}{|\widehat{x}|}}{1+\frac{|x-\widehat{x}|}{|\widehat{x}|}} \le \frac{|x-\widehat{x}|}{|x|} \le \frac{\frac{|x-\widehat{x}|}{|\widehat{x}|}}{1-\frac{|x-\widehat{x}|}{|\widehat{x}|}}$$

Conclude that if  $e_{rel} \leq .1$ , then the actual relative error satisfies  $.8e_{rel} \leq |x - \hat{x}|/|x| \leq 1.2e_{rel}$ .

- 3) Let  $1 + r = \prod_{i=1}^{n} (1 + \delta_i)$ , where  $|\delta_i| \le \epsilon < 1$ .
- **Part 1.** Show that if  $n\epsilon < 1$ , then  $|r| \le n\epsilon/(1 n\epsilon)$ .
- **Part 2.** Show that if  $n\epsilon \leq .1$ , then  $r \leq 1.2n\epsilon$ .
- **Part 3.** In IEEE double precision, how big can n be and satisfy  $n \epsilon \leq .1$ ?
- **Part 4.** If you compute  $p = \prod_{i=1}^{n} x_i$  in floating point arithmetic, and no over/underflow occurs, and  $n\epsilon \leq .1$ , about how many leading decial digits of the computed value of p are correct when using IEEE double precision arithmetic with n = 10? n = 1000? n = 10000?

4) Suppose x > 0. Here are two Matlab algorithms for computing  $e^{-x}$ : Algorithm 1: Compute  $e^{-x}$  using a Taylor expansion

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\begin{split} s &= 1; \, t = 1; \, i = 1; \\ \text{while } (abs(t) > eps^*abs(s)) \\ & \dots \text{ stop iterating when adding t to s does not change s} \\ & t = -t^*x/i; \\ & s = s + t; \\ & i = i + 1; \\ end \\ result1 = s; \end{split}
```

Algorithm 2: Compute  $e^{-x}$  as  $1/e^x$ , using a Taylor expansion for  $e^x$ 

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\begin{split} s &= 1; \ t = 1; \ i = 1; \\ \text{while } (abs(t) > eps^*abs(s)) \\ & \dots \text{ stop iterating when adding t to s does not change s} \\ t &= t^*x/i; \\ s &= s + t; \\ i &= i + 1; \\ \text{end} \\ \text{result2} &= 1/s; \end{split}
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- **Part 1.** Run these two algorithms for x = 1:20, tabulating the relative errors and number of iterations to converge for each.
- **Part 2.** Prove that the relative error of result2 is, as you observe, bounded by  $(3i 2)\epsilon$ , i.e. very accurate. You may assume the error from terminating the Taylor expansion is smaller than round off error, and you may ignore terms proportional to  $\epsilon^2$ . Confirm that  $(3i 2)\epsilon$  bounds the relative errors in your table above.
- **Part 3.** Prove that the relative error of result1 is bounded by  $3(i-1)\epsilon e^{2x}$ , i.e. it grows quickly with x, so that Algorithm 1 is much less accurate than Algorithm 2. You may make the same assumptions as before. Confirm that  $3(i-1)\epsilon e^{2x}$  bounds the relative errors in your table above.
- **Part 4.** The computer implementation for  $e^x$  takes the same time for large and small arguments; i.e. it does not use a simple Taylor expansion, which would require more terms for larger arguments. Sketch an algorithm for  $e^x$  that does not take longer for large x. Use the fact that  $e^x = 2^y$  where  $y = x \cdot \log_2 e$ , write  $y = y_{int} + y_{frac}$  as a sum of an integer and a fraction less than 1, and use the fact that  $2^y = 2^{y_{int}} \cdot 2^{y_{frac}}$  is to be rounded to a floating point number. How many term of a Taylor expansion of  $2^{y_{frac}}$  are needed so that the remaining terms contribute less than  $\epsilon$  to the relative error?