

Assignment 1, due September 20

Corrections: [none yet]

1. We want to design a special floating point format that uses the fewest bits so that any x with $|x| \leq 400$ and $|x| > .1$ is represented to .05% relative accuracy (the relative error is not more than $5 \cdot 10^{-4}$). We need one sign bit, q exponent bits, p fraction bits and an exponent offset e_0 . (These are car velocities in km/h. No car (hopefully) goes faster than 400 km/hr \approx 240 mi/hr, and .1 km/hr \approx .06 mi/hr is a “rolling stop”.) Note that the range of speed exponents is not symmetric, so you may be able to save a bit by choosing the offset carefully.
2. We want to evaluate $f(x) = e^x - 1$ accurately in double precision floating point when x is close to zero. For this problem, assume that exponentiation and multiplication and addition are done with relative accuracy at least ϵ_{mach} , the machine precision of double precision arithmetic.
 - (a) Estimate the relative accuracy of the direct code `f = np.exp(x)-1`. Assume that `np.exp(x)` evaluates e^x to full double precision floating point accuracy.
 - (b) Estimate the relative accuracy of the Taylor series approximation `f = x + .5*x*x`. Assume that the arithmetic is done with double precision floating point accuracy and that the error in the Taylor approximation is $f(x) - (x + .5x^2) = (1/6)x^3$. (This is an accurate estimate of the error for the very small x we have in mind.)
 - (c) Consider the hybrid approximation

```
if ( np.abs(x) < epsilon ):
    f = x + .5*x*x
else:
    f = np.exp(x)-1.
```

Within the range of normalized numbers, the error for this approximation is largest around $x = \pm\epsilon$. For $|x| < \epsilon$ the Taylor expansion becomes more accurate as $|x|$ decreases. For $|x| > \epsilon$, the direct formula becomes more accurate as $|x|$ increases. Find the ϵ that minimizes this largest error. What is the worst case accuracy, given in the number of correct digits of the answer?

3. The *logistic function* with positive *length parameter* r is

$$s(x, r) = \frac{e^{x/r}}{1 + e^{x/r}}.$$

As a function of x , this function switches from $s \approx 0$ when x is a large negative number to $s \approx 1$ when x is a large positive number. It is used in *logistic regression*, which is one of the tools used in data science for classification. The parameter r controls the range of x values where the transition from 0 to 1 happens. The logistic function is used to create *soft thresholds*, which are an alternative to *hard thresholds*. Hard thresholds are equal to zero (on one side of some criterion) or one (on the other side). Evaluate the condition number of the problem of evaluating $s(x, 1)$ (set $r = 1$ for simplicity). Figure out whether it is well conditioned for large negative and positive x . The inverse logistic function problem is find x so that $s(x) = y$, for a given y between 0 and 1. Figure out whether the inverse logistic problem is well conditioned for y near 0 and y near 1.

4. The Fibonacci recurrence relation is

$$f_{n+1} = f_n + f_{n-1} .$$

The *initial conditions* are the values f_0 and f_1 . Once the initial conditions are specified, the rest of the numbers f_n are determined. The *target values* are f_L and f_{L+1} , where L is some positive integer. We are interested in Fibonacci sequences of length L . The Fibonacci numbers are produced using initial conditions $f_0 = 1$ and $f_1 = 1$.

- (a) Show that any Fibonacci sequence has the form

$$f_n = a_1 z_1^n + a_2 z_2^n .$$

The coefficients a_1 and a_2 depend on the initial conditions but the *roots* z_1 and z_2 do not. Hint: The sequence $f_n = z^n$ satisfies the Fibonacci recurrence if and only if z is one of the roots of a given quadratic *characteristic polynomial*. Show it is always possible to find a_1 and a_2 so that $f_0 = a_1 + a_2$ and $f_1 = a_1 z_1 + a_2 z_2$. The roots z_1 and z_2 are positive. Suppose $z_1 > z_2$. The larger root is the *golden mean*.

- (b) Show that $f_{n+1}/f_n \rightarrow z_1$ as $n \rightarrow \infty$ unless $a_1 = 0$. In particular, show that *the* Fibonacci numbers have this property.
- (c) Show that the condition number of computing the target values from the initial conditions is bounded as $L \rightarrow \infty$ unless $a_1 = 0$, when the condition number grows exponentially with L .
- (d) Consider initial conditions with $a_1 \neq 0$ and consider the problem of computing the initial conditions from the target values. Show that the condition number of this problem grows exponentially with L .
5. Write a code in Python 3 related to Problem 4. The code should start with initial conditions (given in the code), compute target values, then reverse the recurrence to re-compute the initial conditions from the target values

$$f_0, f_1 \xrightarrow{\text{floating_point}} \hat{f}_L, \hat{f}_{L+1} \xrightarrow{\text{floating_point}} \hat{f}_0, \hat{f}_1$$

Print the error $f_0 = \widehat{f}_0$ as a function of L .

- (a) Start with $f_0 = \sqrt{2}$ and $f_1 = e$. Show that the errors are roughly explained by the condition number analysis of Problem 4.
- (b) Repeat the exercise with *the* Fibonacci numbers. Why is there no error until L reaches a given point? What determines the L value where error starts?