

# Function Evaluation

In scientific computing we routinely evaluate functions from spaces over real (or complex) numbers into real (or complex) spaces. We write

$$f : D \rightarrow R$$

to say that  $f$  is a function with domain  $D$  and range  $R$  (each element of  $D$  is associated (through  $f$ ) to exactly one element of  $R$  (as in  $f(d) = r$ )). You can think of  $D$  and  $R$  as subsets of the real numbers, but often they are finite dimensional vector spaces over real (or complex) numbers.

We are using a finite model of the reals (the floating point numbers) to approximate the domain and range spaces, so when we use the term “evaluate  $f$  at  $x$ ”, we really mean “evaluate  $f$  at our approximation of  $x$ ”, which will give us an approximation to  $f$  evaluated at our approximation to  $x$ .

Some notation might help here. Let’s say we’d like to evaluate  $f$  at  $x$ , that is, we’d like to compute  $y = f(x)$ . If  $\bar{x}$  is our approximation of  $x \in D$ , then (like-it-or-not) we are actually trying to compute  $\tilde{y} = f(\bar{x})$ . But because of rounding errors in this function evaluation, we instead compute  $\bar{y}$  as our approximation to  $\tilde{y}$ . Our attempt to compute  $y = f(x)$  returns instead  $\bar{y} = \bar{f}(\bar{x})$ , and we hope that  $\bar{y} \approx \tilde{y} \approx y$ .

In summary:

- $y = f(x)$  is what we want, but we have  $\bar{x}$  instead of  $x$ ,
- $\tilde{y} = f(\bar{x})$ , is what we try to compute, but is subject to rounding errors, and
- $\bar{y} = \bar{f}(\bar{x})$  is what we have actually computed.

Attempting to predict or discover how much  $\bar{y}$  and  $y$  might differ is a rather difficult question in general, and is an important part of numerical analysis. One way to simplify the question is to imagine we *can* compute  $\tilde{y}$ : When we speak of “the magic method” we imagine a method which returns the  $\bar{y}$  that is the closest element of our model of  $R$  to  $\tilde{y}$ . This is the best we might do, and is a useful concept to keep in mind in any analysis of computing with real numbers. We’ll see that how good (or poorly) the magic method performs is a measure of how difficult our problem is (condition numbers...).

Think about just how good an approximation the magic method might give for various functions (e.g.  $y = f(x) = 1/x$ ,  $y = f(x) = \sin(x)$ ,  $y = f(A, b) = A^{-1}b$ , etc.), or how you might evaluate  $f(x)$  using only arithmetic operations, or how you might approximate the error  $y - \bar{y}$ .

You may think that “evaluate  $f$  at  $x$ ” is a trivial task, but if you think about trying to solve *any* problem (which has a unique solution), then you come to see that this is exactly what we are trying to do: the solution is  $f(\text{input data})$ , for some function  $f$  which maps the input data to the solution. At the risk of getting too philosophical, I also remind you that this function that we are trying to evaluate is probably only an approximation of some (more complicated, or unknown, or unknowable) function.