

How to Deal with Infinities: Why $x_i/\sqrt{1+r^2}$?

Jeffrey L. Vanarsdall, Dario A. Vazquez, and Vladik Kreinovich
Department of Computer Science, University of Texas at El Paso
jlvanarsdall@miners.utep.edu, davazquez5@miners.utep.edu, vladik@utep.edu

Formulation of the problem. Most effective numerical techniques, in particular, optimization techniques and techniques for solving systems of nonlinear equations – assume that the set of possible alternatives is bounded. However, in practice, we sometimes have situations in which no such bounds are known. A natural idea is to apply some 1-1 transformation T of the whole unbounded n -dimensional space into a bounded domain. This way, we will be able to use bounded-domain methods to solve the correspondingly transformed problems.

There are many possible transformations. Which of them should we choose? In [1], it was shown that good results appear if we use the transformation $x_i \mapsto T_0(x) = x_i/\sqrt{1+r^2}$, where $r^2 \stackrel{\text{def}}{=} \sum_i x_i^2$. A natural question is: why this particular transformation was – empirically – most successful? In this talk, we provide a possible answer to this question.

First requirement: the transformation T should be rotation-invariant. Let us start enumerating reasonable requirements on T . Many practical problems are invariant with respect to rotations R , and this invariance helps to solve them. It is therefore reasonable to require that T preserve all rotations, i.e., that (1) if x is invariant under rotation R , then $T(x)$ should be similarly invariant, and, more generally, (2) if $y = R(x)$, then we should have $T(y) = R(T(x))$. In particular, each vector x is invariant with respect to all rotations around x . Thus, $T(x)$ should be similarly invariant – which means that $T(x)$ should be parallel to x : $T(x) = c \cdot x$ for some scalar c . Every two vectors of the same length can be rotated into each other, so c should only depend on the length r of the vector: $x_i \mapsto x_i \cdot c(r)$.

Second requirement: T should be fast to compute. The fastest are hardware supported operations, and only arithmetic operations are hardware supported. Computing r requires taking a square root, so we better describe c as depending on r^2 – which is mathematically the same, so $x_i \mapsto x_i \cdot c(r^2)$.

Let us show that we cannot use only arithmetic operations to compute $c(r^2)$. Indeed, then, as one can show by induction, we get a rational function, i.e., a ratio $P(r^2)/Q(r^2)$ of two polynomials. If the degree $\deg(P)$ of P is $\geq \deg(Q)$, then for $r \rightarrow \infty$, we have $c(r^2) \sim r^{2a}$ for some $a \geq 0$, so $x_i \cdot c(r^2) \rightarrow \infty$, while we want a bounded domain. If $\deg(P) < \deg(Q)$, then for $r \rightarrow \infty$, we have $c(r^2) \sim r^{-2a}$ for some $a \geq 1$. Then $x_i \cdot c(r^2) \rightarrow 0$ when $r \rightarrow \infty$, and we also have $x_i \cdot c(r^2) \rightarrow 0$ when $r \rightarrow 0$, so T is not 1-1.

Thus, we need to use at least one non-arithmetic operation. Square root – which is part of the definition of r – is a natural choice. So, the next try is $x_i \mapsto x_i \cdot \sqrt{C(r^2)}$ for some rational function $C(r^2)$. Computing $x_i/\sqrt{1+r^2}$ requires – besides computing the square root – one addition and one division. We need at least one division – otherwise, C will be a polynomial and we will have $T(x) \rightarrow \infty$, while we want a bounded domain. If we only have one division and no other operations, then the only choice is $x_i/\sqrt{r^2} = x_i/r$ which is not 1-1: x_i and $2x_i$ transform into the same vector. Thus, we need at least one operation in addition to division, so indeed operation from [1] is the fastest to compute (and one can show that no other operation is as fast).

[1] H. Schichl, A. Neumaier, M. C. Markót, and F. Domes, “On solving mixed-integer constraint satisfaction problems with unbounded variables”, In: C. Gomes and M. Sellmann (eds.), *Integration of AI and OR Techniques in Constraint Programming for Combinatorial Optimization Problems*, Springer, Berlin, Heidelberg, 2013, pp. 216–233.