Universal Numerical Algorithms and Their Software Implementation

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Abstract
The concept of a universal algorithm is discussed. Examples of this kind of algorithms are presented. Software implementations of such algorithms in C++-type languages are discussed together with means that provide for computations with an arbitrary accuracy. Particular emphasis is placed on universal algorithms of linear algebra over semirings.

INTRODUCTION

Modern achievements in software development and mathematics make us consider numerical algorithms and their classification from a new point of view. Conventional numerical algorithms are oriented to software (or hardware) implementation with the use of the floating point arithmetic and fixed accuracy. However, it is often desirable to perform computations with variable (and arbitrary) accuracy. For this purpose, algorithms are required that are independent of the accuracy of computations and of a particular computer representation of numbers. In fact, many algorithms are not only independent of the computer representation of numbers, but also of concrete mathematical (algebraic) operations on data. In this case, operations may
be considered as variables. Such algorithms are implemented by *generic programs* based on the abstract data types technique (abstract data types are defined by the user, in addition to predefined types of the language used). The corresponding program tools appeared as early as in Simula-67, but modern object-oriented languages (like C++, see, e.g. [1, 2]) are more convenient for generic programming.

The concept of a generic program was introduced by many authors; for example, in [3], such programs were called program schemes. In this paper, we discuss *universal algorithms* implemented as generic programs and their specific features. This paper is closely related to papers [4, 5], in which the concept of a universal algorithm was defined and software and hardware implementation of such algorithms was discussed in connection with problems of idempotent mathematics [4, 6]. In this paper, the emphasis is placed on software implementation of universal algorithms, computations with arbitrary accuracy, universal algorithms of linear algebra over semirings, and their implementation in C++.

1. UNIVERSAL ALGORITHMS

Computational algorithms are constructed on the basis of certain basic operations. Basic operations manipulate data that describe “numbers”. These “numbers” are elements of a “numerical domain”, i.e., a mathematical object like the field of real numbers, the ring of integers, or an idempotent semiring of numbers (idempotent semirings and their role in idempotent mathematics are discussed in [4, 6] and below in this paper). In every particular computation, elements of the numerical domains are replaced by their computer representations, i.e., by elements of certain finite models of these domains. Examples of models that can be conveniently used for computer representation of real numbers are provided by various modifications of floating point arithmetics, approximate arithmetics of rational numbers [4], and interval arithmetics. The difference between mathematical objects (“ideal” numbers) and their finite models (computer representations) results in computational (e.g., rounding) errors.
An algorithm is called universal if it is independent of a particular numerical domain and (or) of its computer representation. A typical example of a universal algorithm is computation of the scalar product \((x, y)\) of two vectors \(x = (x_1, \ldots, x_n)\) and \(y = (y_1, \ldots, y_n)\) by the formula \((x, y) = x_1y_1 + \ldots + x_ny_n\). This algorithm (formula) is independent of a particular domain and its computer implementation, since the formula is defined for any semiring. It is clear that one algorithm can be more universal than another. For example, the simplest rectangular formula provides the most universal algorithm for numerical integration; indeed, this formula is valid even for idempotent integration (over any idempotent semiring \([4]\)). Other quadrature formulas (e.g., combined trapezoid or Simpson formulas) are independent of computer arithmetics and can be used (e.g., in the iterative form) for computations with arbitrary accuracy. In contrast, algorithms based on Gauss–Jacobi formulas are designed for fixed accuracy computations: they include constants (coefficients and nodes of these formulas) defined with fixed accuracy. Certainly, algorithms of this type can be made more universal by including procedures for computing the constants; however, this results in an unjustified complication of the algorithms.

Computer algebra algorithms used in such systems as Mathematica, Maple, REDUCE, and others are highly universal. Standard algorithms used in linear algebra can be rewritten in such a way that they will be valid over any field and complete idempotent semiring (including semirings of intervals; see \([8, 9]\), where an interval version of the idempotent linear algebra and the corresponding universal algorithms are discussed).

As a rule, iterative algorithms (beginning with the successive approximation method) for solving differential equations (e.g., methods of Euler, Euler–Cauchy, Runge–Kutta, Adams, a number of important versions of the difference approximation method, and the like), methods for calculating elementary and some special functions based on the expansion in Taylor’s series and continuous fractions (Padé approximations) and others are independent of the computer representation of numbers.
2. UNIVERSAL ALGORITHMS AND ACCURACY OF COMPUTATIONS

Calculations on computers usually are based on a floating-point arithmetic with a mantissa of a fixed length; i.e., computations are performed with fixed accuracy. Broadly speaking, with this approach only the relative rounding error is fixed, which can lead to a drastic loss of accuracy and invalid results (e.g., when summing series and subtracting close numbers). On the other hand, this approach provides rather a high speed of computations. Many important numerical algorithms are designed to use a floating-point arithmetic (with fixed accuracy) and ensure the maximum computation speed. However, these algorithms are not universal. The above mentioned Gauss–Jacobi quadrature formulas, computation of elementary and special functions on the basis of the best polynomial or rational approximations or Padé–Chebyshev approximations, and some others belong to this type. Such algorithms use nontrivial constants specified with fixed accuracy.

Recently, problems of accuracy, reliability, and authenticity of computations (including the effect of rounding errors) have come to the fore; in part, this fact is related to the ever-increasing performance of computer hardware. When errors in initial data and rounding errors strongly affect the computation results (ill-posed problems, analysis of stability of solutions, etc.), it is often useful to perform computations with improved and variable accuracy. In particular, the rational arithmetic, in which the rounding error is specified by the user, can be used for this purpose. This arithmetic is a useful complement to the interval analysis. The corresponding computational algorithms must be universal (in the sense that they must be independent of the computer representation of numbers).

4. MATHEMATICS OF SEMIRINGS

A broad class of universal algorithms is related to the concept of a semiring. We reiterate here the definition of a semiring (see, e.g., [11]). Let $S$ be a set on which associative binary operations $\oplus$ and $\odot$, called addition and multiplication, respectively, are defined. We assume that addition
is commutative and that multiplication is distributive over addition; i.e., \( x \odot (y \oplus z) = (x \odot y) \oplus (x \odot z) \) and \( (x \oplus y) \odot z = (x \odot z) \oplus (y \odot z) \) for all \( x, y, z \in S \). In this case, \( S \) is called a semiring. We assume that the semiring \( S \) contains identity 1 and zero 0; i.e., \( 1 \odot x = x \odot 1 = x \) and \( 0 \oplus x = x \), \( 0 \odot x = x \odot 0 = 0 \); in addition, \( 0 \neq 1 \). As is customary, we sometimes omit the multiplication symbol.

A semiring \( S \) is called commutative if multiplication \( \odot \) is commutative. A semiring \( S \) is called idempotent if \( x \oplus x = x \) for all \( x \in S \). If a semiring is a group under addition, it is called a ring (in this case, it cannot be idempotent). If every nonzero element of a commutative ring (semiring) is invertible under multiplication, this ring (semiring) is called a field (semifield).

The best known and most important examples of semirings are “numerical” semirings consisting of real numbers. For example, the set \( \mathbb{R} \) of all real numbers is a field under ordinary arithmetic operations; i.e., \( \oplus = + \), \( \odot = \cdot \), \( 0 = 0 \), \( 1 = 1 \). The set \( \mathbb{R}_{\text{max}} = \mathbb{R} \cup \{-\infty\} \) equipped with operations \( \oplus = \max \) and \( \odot = + \) provides an example of an idempotent semiring (and semifield). Here \( 0 = -\infty \) and \( 1 = 0 \). This semifield is often called the Max-Plus algebra. The semiring \( \mathbb{R}_{\text{min}} = \mathbb{R} \cup \{+\infty\} \) equipped with operations \( \oplus = \min \) and \( \odot = + \) is isomorphic to the Max-Plus algebra. Here \( 0 = +\infty \) and \( 1 = 0 \). Another example is the set \( S_{\text{max,min}}^{[a,b]} \) consisting of the elements of an interval \( [a, b] \), where \( -\infty \leq a < b \leq +\infty \), equipped with operations \( \oplus = \max \) and \( \odot = \min \); here \( 0 = a \) and \( 1 = b \). This commutative semiring is not a semifield.

An important example of a noncommutative semiring is the set \( \text{Mat}_n(S) \) of all matrices of order \( n \times n \) with elements from a commutative semiring \( S \) with ordinary standard operations. The sum of matrices \( A = (a_{ij}) \) and \( B = (b_{ij}) \) is the matrix \( A \oplus B = (a_{ij} \oplus b_{ij}) \), and the product of these matrices is the matrix \( AB = (\bigoplus_{k=1}^{n} a_{ik} \odot b_{kj}) \), where \( i, j = 1, \ldots, n \). Operations on rectangular matrices can be defined similarly. Zero \( O \) and identity \( I \) in \( \text{Mat}_n(S) \) are defined in the conventional way. If the semiring \( S \) is idempotent, then \( \text{Mat}_n(S) \) is also idempotent. Many other important examples can be found in [3] – [6], [8], [9], [11].

On any idempotent semiring, a canonical partial order \( \preceq \) is defined by the
following rule: \( x \lessdot y \) is equivalent to \( x \oplus y = y \). Moreover, \( x \oplus y = \sup \{x, y\} \) with respect to the canonical order. The canonical order is compatible with the semiring addition and multiplication in the common way. For the semirings \( \mathbb{R}_{\text{max}} \) and \( [a, b]_{\text{max, min}} \), the canonical order coincides with the standard order \( \leq \) defined on the set of real numbers; for the semiring \( \mathbb{R}_{\text{min}} \), it is inverse to the standard order.

There exists a (heuristic) correspondence between important, useful, and interesting constructs and results of traditional mathematics over fields and similar constructs and results of idempotent mathematics (i.e., mathematics over idempotent semirings). This idempotent correspondence principle is closely related to the Bohr correspondence principle in quantum mechanics. Traditional mathematics can be considered as a “quantum” theory and idempotent mathematics as its “classical” analogue (see \[4\]). Consistent application of the idempotent correspondence principle leads to various and surprising results, including a methodology for constructing universal algorithms and patenting computer devices \[4, 5\].

The fundamental equations in quantum theory are linear (superposition principle). There is an idempotent version of the superposition principle \[6\]: the Hamilton–Jacobi equation, i.e., the basic (nonlinear) equation of classical mechanics, can be considered as linear over the semiring \( \mathbb{R}_{\text{min}} \); various modifications of the Bellman equation, i.e., the basic equation of optimization theory, are also linear over appropriate idempotent semirings. For example, the finite-dimensional time-independent Bellman equation can be written as

\[
X = A \odot X \oplus B,
\]

where \( A \) is a square matrix with elements from an idempotent semiring \( S \) and \( X \) and \( B \) are vectors (or matrices) with elements from \( S \). The solution \( X \) is found from (1) when \( A \) and \( B \) are given.

In particular, standard problems in dynamic programming correspond to the case \( \mathbb{R}_{\text{max}} \), and the well-known shortest path problem corresponds to \( S = \mathbb{R}_{\text{max}} \). It is shown in \[12\] that the principal optimization algorithms for finite graphs correspond to standard methods for solving systems of linear equations of form (1) over semirings. The Bellman algorithm for the shortest
The most important linear algebra problem is solving the system of linear equations
\[ AX = B, \]  
where \( A \) is a matrix with elements from the basic field and \( X \) and \( B \) are vectors (or matrices) with elements from the same field. It is required to find \( X \) if \( A \) and \( B \) are given. If \( A \) in (2) is not the identity matrix \( I \), then system (2) can be written in form (1), i.e.,
\[ X = AX + B. \]  
(1')

It is well known that form (1) or (1') is convenient for using the successive approximation method. Applying this method with the initial approximation \( X_0 = 0 \), we obtain the solution
\[ X = A^r B, \]  
(3)

where
\[ A^r = I + A + A^2 + \ldots + A^r + \ldots \]  
(4)
On the other hand, it is clear that

\[ A^* = (I - A)^{-1}, \quad (5) \]

if the matrix \( I - A \) is invertible. The inverse matrix \((I - A)^{-1}\) can be considered as a regularized sum of the formal series (4).

The above considerations can be extended to a broad class of semirings. The unary operation \( A \mapsto A^* \) in \( \text{Mat}_n(S) \) is defined (partially) if a unary (partial) operation \( x \mapsto x^* \), called closure, is defined on the semiring \( S \) such that the identity

\[ x^* = 1 \oplus (x^* \odot x) = 1 \oplus (x \odot x^*) \quad \text{(6)} \]

holds true if \( x^* \) is defined. It follows from (6) that

\[ x^* = 1 \oplus x \oplus x^2 \oplus \ldots \oplus x^n \oplus x^*x^{n+1} \]

for any positive integer \( n \); thus, \( x^* \) can be considered as a regularized sum of the formal series

\[ x^* = 1 \oplus x \oplus x^2 \oplus \ldots \oplus x^n \oplus \ldots \]

If \( S \) is a field, then, by definition, \( x^* = (1 - x)^{-1} \) for any \( x \neq 1 \). If \( S \) is an idempotent semiring, then, by definition

\[ x^* = 1 \oplus x \oplus x^2 \oplus \ldots \oplus x^n \oplus \ldots = \text{sup}\{1, x, x^2, \ldots, x^n, \ldots\}, \quad (7) \]

if this supremum (with respect to the canonical order \( \preceq \)) exists. In this case, \( x^* = 1 \) if \( x \preceq 1 \). Therefore, \( x^* = 1 \) in the semiring \( S^{[a,b]} \) for all \( x \). For the semifield \( R_{\text{max}} \) the closure operator \( x \mapsto x^* \) is not defined for \( 1 \prec x \) (however, \( R_{\text{max}} \) can be supplemented by \( +\infty \), which turns this semifield into a semiring; in this case, \( x^* = +\infty \) for \( 1 \prec x \)). It is clear that, for \( x \preceq 1 \), \( x^* = 1 \) in \( R_{\text{max}} \), as well as in other idempotent semirings. These examples show that the closure \( x^* \) of \( x \) is often calculated very simply for idempotent semirings.

The closure operation for matrix semirings \( \text{Mat}_n(S) \) can be defined and computed in terms of the closure operation for \( S \); some methods are described in [3, 4, 11, 12]. One such method is described below (\( LDM \)-factorization). The closure operation \( A \mapsto A^* \) in \( \text{Mat}_n(S) \) satisfies identity (6), which implies
that if $A^*$ is defined, then $X = A^*B = A^* \odot B$ is the solution to the matrix equation (1).

Consider a nontrivial universal algorithm applicable to matrices over semirings with the closure operation defined.

Example: Semiring LDM-Factorization

Factorization of a matrix into the product $A = LDM$, where $L$ and $M$ are lower and upper triangular matrices with a unit diagonal, respectively, and $D$ is a diagonal matrix, is used for solving matrix equations $AX = B$ [13]. We construct a similar decomposition for the Bellman equation $X = AX \oplus B$.

For the case $AX = B$, the decomposition $A = LDM$ induces the following decomposition of the initial equation:

$$LZ = B, \quad DY = Z, \quad MX = Y.$$ (8)

Hence, we have

$$A^{-1} = M^{-1}D^{-1}L^{-1},$$ (9)

if $A$ is invertible. In essence, it is sufficient to find the matrices $L, D$ and $M$, since the linear system (8) is easily solved by a combination of the forward substitution for $Z$, the trivial inversion of a diagonal matrix for $Y$, and the back substitution for $X$.

Using (8) as a pattern, we can write

$$Z = LZ \oplus B, \quad Y = DY \oplus Z, \quad X = MX \oplus Y.$$ (10)

Then

$$A^* = M^*D^*L^*.$$ (11)

A triple $(L, D, M)$ consisting of a lower triangular, diagonal, and upper triangular matrices is called an $LDM$-factorization of a matrix $A$ if relations (10) and (11) are satisfied. We note that in this case, the principal diagonals of $L$ and $M$ are zero.

The modification of the notion of $LDM$-factorization used in matrix analysis for the equation $AX = B$ is constructed by analogy with the construct suggested by Carré in [12] for $LU$-factorization.
We stress that the algorithm described below can be applied to matrix computations over any semiring under the condition that the unary operation \(a \mapsto a^\ast\) is applicable every time it is encountered in the computational process. Indeed, when constructing the algorithm, we use only the basic semiring operations of addition \(\oplus\) and multiplication \(\odot\) and the properties of associativity, commutativity of addition, and distributivity of multiplication over addition.

If \(A\) is a symmetric matrix over a semiring with a commutative multiplication, the amount of computations can be halved, since \(M\) and \(L\) go into each other under transposition.

We begin with the case of a triangular matrix \(A = L\) (or \(A = M\)). Then, finding \(X\) is reduced to the forward (or back) substitution.

**Forward substitution**

We are given:
- \(L = \|l_{ij}\|_{i,j=1}^n\) where \(l_{ij} = 0\) for \(i \leq j\) (a lower triangular matrix with a zero diagonal);
- \(B = \|b^i\|_{i=1}^n\).

It is required to find the solution \(X = \|x^i\|_{i=1}^n\) to the equation \(X = LX \oplus B\). The program fragment solving this problem is as follows.

```plaintext
for i = 1 to n do
{  x^i := b^i;
    for j = 1 to i - 1 do
        x^i := x^i \oplus (l_{ij} \odot x^j); }
```

**Back substitution**

We are given
- \(M = \|m_{ij}\|_{i,j=1}^n\) where \(m_{ij} = 0\) for \(i \geq j\) (an upper triangular matrix with a zero diagonal);
• $B = \|b^i\|^n_{i=1}$.

It is required to find the solution $X = \|x^i\|^n_{i=1}$ to the equation $X = MX \oplus B$. The program fragment solving this problem is as follows.

for $i = n$ to 1 step $-1$ do
\{
  $x^i := b^i$;
  for $j = n$ to $i + 1$ step $-1$ do
  $x^i := x^i \oplus (m_j^i \odot x^i)$;
\}

Both algorithms require $(n^2 - n)/2$ operations $\oplus$ and $\odot$.

Closure of a diagonal matrix

We are given

• $D = \text{diag}(d_1, \ldots, d_n)$;

• $B = \|b^i\|^n_{i=1}$.

It is required to find the solution $X = \|x^i\|^n_{i=1}$ to the equation $X = DX \oplus B$. The program fragment solving this problem is as follows.

for $i = 1$ to $n$ do
\{ $x^i := (d_i)^* \odot b^i$; \}

This algorithm requires $n$ operations $*$ and $n$ multiplications $\odot$.

General case

We are given

• $L = \|l^i_j\|^n_{i,j=1}$, where $l^i_j = \mathbf{0}$ if $i \leq j$;

• $D = \text{diag}(d_1, \ldots, d_n)$;
• $M = \|m^i_j\|_{i,j=1}^n$, where $m^i_j = 0$ if $i \geq j$;

• $B = \|b^i\|_{i=1}^n$.

It is required to find the solution $X = \|x^i\|_{i=1}^n$ to the equation $X = AX \oplus B$, where $L$, $D$, and $M$ form the $LDM$-factorization of $A$. The program fragment solving this problem is as follows.

**FORWARD SUBSTITUTION**
for $i = 1$ to $n$ do
  \{ $x^i := b^i$; 
  for $j = 1$ to $i - 1$ do
    $x^i := x^i \oplus (l^i_j \odot x^j)$; \}

**CLOSURE OF A DIAGONAL MATRIX**
for $i = 1$ to $n$ do
  $x^i := (d_i)^* \odot b^i$;

**BACK SUBSTITUTION**
for $i = n$ to 1 step $-1$ do
  \{ for $j = n$ to $i + 1$ step $-1$ do
    $x^i := x^i \oplus (m^i_j \odot x^j)$; \}

Note that $x^i$ is not initialized in the course of the back substitution. The algorithm requires $n^2 - n$ operations $\oplus$, $n^2$ operations $\odot$, and $n$ operations $\ast$.

**LDM-factorization**

We are given
• $A = \|a^i_j\|_{i,j=1}^n$.

It is required to find the $LDM$-factorization of $A$: $L = \|l^i_j\|_{i,j=1}^n$, $D = \text{diag}(d_1, \ldots, d_n)$, and $M = \|m^i_j\|_{i,j=1}^n$, where $l^i_j = 0$ if $i \leq j$, and $m^i_j = 0$ if $i \geq j$.

The program uses the following internal variables:
• $C = \|c^i_j\|_{i,j=1}^n$;
\[ V = \|v^i\|_{i=1}^n; \]

\[ d. \]

**INITIALISATION**
for \( i = 1 \) to \( n \) do
  for \( j = 1 \) to \( n \) do
    \[ c^i_j = a^i_j; \]

**MAIN LOOP**
for \( j = 1 \) to \( n \) do
  \{ for \( i = 1 \) to \( j \) do
    \[ v^i := a^i_j; \]
    for \( k = 1 \) to \( j - 1 \) do
      for \( i = k + 1 \) to \( j \) do
        \[ v^i := v^i \oplus (a^i_k \odot v^k); \]
    for \( i = 1 \) to \( j - 1 \) do
      \[ a^i_j := (a^i_j)^* \odot v^i; \]
      \[ a^j_j := v^j; \]
    for \( k = 1 \) to \( j - 1 \) do
      for \( i = j + 1 \) to \( n \) do
        \[ a^i_j := a^i_j \oplus (a^i_k \odot v^k); \]
    \[ d = (v^j)^*; \]
  for \( i = j + 1 \) to \( n \) do
    \[ a^i_j := a^i_j \odot d; \]
  \}

This algorithm requires \((2n^3 - 3n^2 + n)/6\) operations \(\oplus\), \((2n^3 + 3n^2 - 5n)/6\) operations \(\odot\), and \(n(n + 1)/2\) operations \(*\). After its completion, the matrices \(L, D,\) and \(M\) are contained, respectively, in the lower triangle, on the diagonal, and in the upper triangle of the matrix \(C\). In the case when \(A\) is symmetric about the principal diagonal and the semiring over which the matrix is defined is commutative, the algorithm can be modified in such a way that the number of operations is reduced approximately by a factor of two. For details see [13].

Other examples can be found in [3], [11] – [15].
6. SOFTWARE IMPLEMENTATION OF UNIVERSAL ALGORITHMS

Object-oriented languages (e.g., C++ and Java) and programming systems that allow abstract data types to be defined provide convenient means for the software implementation of universal algorithms. In this case, program units can operate with abstract (and variable) operations and data types. Specific values of operations are determined by the input data types; these operations (and data types) are implemented by additional program units. Recently, this type of programming technique has been dubbed generic programming (see, e.g., [1, 2]). To help automate the generic programming, the so-called Standard Template Library (STL) was developed in the framework of C++ [2, 3]. However, high-level tools, such as STL, possess both obvious advantages and some disadvantages and must be used with caution.

Using the generic programming technique, a program package was developed in C++ for solving problems in linear algebra over fields and semirings (for various computer implementation of the corresponding numeric domains) and optimization problems on graphs. A hierarchy of abstract data types for basic numeric fields, rings, semifields, and semirings was developed for various computer representations. In particular, various versions of the rational arithmetic [4] can be used and computations can be performed with any given accuracy. Solving systems of linear Bellman equations over idempotent semirings (by various methods), standard optimization problems on graphs can be solved (the dynamic programming problem, shortest path problem, widest path problem, etc.), including interval versions of those problems [5], [6]. The system provides a basis for a more powerful program package based on universal algorithms [3]. This system will be described in detail in subsequent publications.

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