A simple C++ library for manipulating scientific data sets as structured data

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Abstract

Representing scientific data sets efficiently on external storage usually involves converting them to a byte string representation using specialized reader/writer routines. The resulting storage files are frequently difficult to interpret without these specialized routines as they do not contain information about the logical structure of the data. Avoiding such problems usually involves heavy-weight data format libraries or database systems. We present a simple C++ library that allows to create and access data files that store structured data. The structure of the data is described by a data type that can be built from elementary data types (integer and floating-point numbers, byte strings) and composite data types (arrays, structures, unions). An abstract data access class presents the data to the application. Different actual data file structures can be implemented under this layer. This method is particularly suited to applications that require complex data structures, e.g. molecular dynamics simulations. Extensions such as late type binding and object persistence are discussed.

1 Introduction

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1.1 Classical input/output mechanisms

Early programming languages had surprisingly advanced features for reading and writing data from external memory. For example, COBOL already had some sort of data definition language, several file formats and data query statements. It was based on the notion that the physical representation of the data on external storage was bit-by-bit identical to the representation in internal memory. As very early computers had primitive operations to copy data from external storage into internal memory, this was efficient, but it also provided a framework for external data representation: records and fields were usually of fixed length, no separator characters were necessary, and numbers could be either represented as ASCII or EBCDIC strings (eventually with an implied decimal point), binary-coded decimal (storing two digits in a byte) or binary numbers. If you could read COBOL’s data definition language, interpreting file contents was merely a matter of counting bytes.

In contrast, the data formats used today in scientific computing are much more flexible, but without human-readable documentation or careful reading of the source code, it is often impossible to decipher the contents of a data file. Designing data exchange between different scientific applications can become a major headache, especially in small- and medium-sized applications where big input/output libraries like NCSA’s HDF [4] or CERN’s RD45 [7] may be inappropriate.

This problem is in part founded in the design of the C language: instead of including input/output statements in the language definition itself, the designers of C decided to implement the whole input/output functionality in the standard C library, using only standard function calls. This implies that C cannot provide a standardized way to store structures - and external data most frequently is organized in structures, i.e. records of data containing dissimilar fields. It is possible to output an arbitrary structure bit-by-bit, but implementation dependencies such as alignment rules easily jeopardize compatibility even between different compiler revisions.

Thus input/output in C programs is usually done manually by writing explicit code that serializes and reassembles data from a byte stream. Worse than the additional work and sources of errors associated with this is the lack of a formal definition of the external data. The structure of the data files must either be given in the human-readable documentation, or—worse—be inferred from the actual source code. As a consequence, there are no general utilities to manipulate binary files, e.g. to produce formatted listings
or extract individual records and fields.

The situation is only slightly remedied in the more popular scientific computing language, FORTRAN. While there are input/output statements in the language, there is no real notion of structured data in FORTRAN 77, so the actual data representation is again encoded in the sequence of READ and WRITE statements in the code.

1.2 Object persistence

In the object-oriented programming paradigm, objects encapsulate data and the operations that act on the data. A persistent object is an object with a life-time that extends beyond the life-time of the program that created it. This means that the object’s data must be stored in external storage, and some operation must be specified to save and recreate the object from external storage. While this can be done by implementing the appropriate read and write methods in the object’s class, it is more desirable to have an automated and standardized way to do so.

A similar problem arises in the design of distributed programming systems. Architectures like CORBA (the Common Object Request Broker Architecture [1, 2, 3]) specify external representations for objects in terms of their methods, but not of their data. They do so by means of an Interface Definition Language (IDL) that is abstractly declares (but not defines) the methods associated with an object class. The IDL is mapped to appropriate method declarations in specific programming languages where the implementations of the methods can be provided. An interoperability protocol defines how methods can be invoked on data residing on different computers in a network.

A logical extension would be to extend the Interface Definition Language by a Data Definition Language (DDL). The relation between the DDL and the actual representation of the data in memory can then be included in the language mapping, and between the DDL and the actual representation in external storage in the interoperability definition.

However, as CORBA is still quite complex and viable implementations are only recently entering the scientific computing community, and a standard for a DDL is still missing, one might look around for a poor man’s solution for object persistence. Such a solution should at least cover the following requirements:
• The structure of external data should be formally specified, similar to the interface definition language of distributed systems.

• The actual representation of the data on external storage should be sufficiently defined by the formal definition of the data structure such that the extraction of the data can be performed automatically. This corresponds to the interoperability specification in distributed programming systems.

• A language mapping or application programmers’ interface that makes external data easily accessible from application programs.

In the following, we present an approach to fulfill these requirements using a set of C++ classes.

• The structure of data is formally defined by a type tree using elementary and composite data types including structures, arrays, and unions (in C parlance). The type tree is built from appropriate C++ classes or specified in textual form, making use of the appropriate methods for reading and printing type trees.

• To provide different storage representations, the interface between the abstract data layer and concrete representations is specified as an abstract C++ class that can be filled in by different data representations. A simple format for arbitrary structured binary data is specified.

• From an application program, data objects are accessed using a C++ class that represents generic structured data and operations performed on such data, such as reading, writing, and extracting members of composite data.

2 Implementation

2.1 Overview

The main class provided by the library is `SomeData`, an universal access layer to structured data. To the programmer, each object of this class represents a structured data item (which may be elementary or composite). To each object of this class a type tree is associated that is described by an object of the class `SomeType`.
Before an object can be used, its data type must be specified. This can be done in three ways:

1. The data type can be specified in the external representation of the data and then be queried by the application program.

2. The data type can be explicitly built using the constructors of the subclasses of SomeType:

   ```
   StructType *t1 = new StructType;
t1->addField("comment",new StringType);
StructType *tAtom = new StructType;
tAtom->addField("name",new StringType);
tAtom->addField("z",new NumType(NumType::i2));
tAtom->addField("partial_charge", new NumType(NumType::f4));
ArrayType *tAtoms = new ArrayType(tAtom);
t1->addField("atoms",tAtoms);
StructType *tBond = new StructType;
tBond->addField("from_atom", new NumType(NumType::i2));
tBond->addField("to_atom", new NumType(NumType::i2));
tBond->addField("type", new NumType(NumType::i2));
ArrayType *tBonds = new ArrayType(tBond);
t1->addField("bonds",tBonds);
   ```

3. Or the data type can be specified in its textual representation as a string, e.g.

   ```
   const char *typetext =
   "struct { "
   "comment : string; "
   "atoms : array of struct { "
   "name : string; "
   "z : integer*2; "
   "partial_charge : real*4; "
   "}; "
   "bonds : array of struct { "
   "from_atom : integer*2; "
   "to_atom : integer*2; "
   "type : integer*2; "
   "}; 
   "}
   ```
SomeType *t1 = SomeType::parse(typetext);

and then parsed by a static member function of SomeType.

After the data type has been specified, a data object must be created. This is done by a data-set class that manufactures an instance SomeData. Data-set classes represent mechanisms where data is stored and retrieved, e.g. file formats or databases. The most basic data-set class is DirectData that stores the data in a linked tree in heap memory. While this is of no use by itself, it can be used by file formats that read their data files in whole and do not wish to provide their own individual access operators.

One such data-set class is DataFile that acts as an interface to text or binary structured files. It provides a method data() that returns the data object associated with the file:

```cpp
DataFile DF1;
DF1.openOut("outfile", t1);
SomeData D1 = DF1.data();

DataFile DF2;
DF2.openIn("outfile");
SomeData D2 = DF.data();
SomeType *t2 = D2.typ();
```

The first group of lines opens a data file for writing using the previously built type t1 and acquires the object D1 to access the data. The second group opens a data file for input and obtains a pointer to the data type in t2. It may use this pointer to ascertain that the data has a certain structure.

The main task of SomeData is to provide data access methods. An example code fragment manipulating an object D would be:

```cpp
D["comment"] = "blubb blubb";
SomeData Datoms = D["atoms"];  
for (int i=0; i<10; i++) {
    SomeData Datom = Datoms[i];
    Datom["z"] = 12;
    Datom["partial_charge"] = 0.0;
    Datom["name"] = "C";
```
}  
SomeData Dbonds = D["bonds"];  
for (int i=0; i<10; i++) {
    SomeData Dbond = Dbonds[i];
    Dbond["from_atom"] = i;
    Dbond["to_atom"] = (i+1)%10;
    Dbond["type"] = 1;
}  
cout << Datom[0]["z"].getInt() << " " <<
    Datom[0]["name"].getString() << endl;

If \( D \) is a structured data object, its members can be accessed by an overloaded indexing operator using either a symbolic name (given as a string, for structure data) or an integer index (for array data). Elementary data types are operated upon either by the assignment operator, which is properly overloaded for the different data types, or using access operators like get\(\text{Int}()\).

2.2 The data type classes

2.2.1 Elementary data types

Three elementary data types are supported

1. Signed or unsigned integer numbers with 1, 2, 4, or 8 bytes, e.g.

   \textbf{integer*4}

   designates a 4-byte signed integer.

2. Floating point numbers in IEEE-format with 4, 8, or 16 bytes, e.g.

   \textbf{real*8}

   is a 8-byte IEEE 754 floating point number.

3. Character strings (that are subject to character-set conversion) and (opaque) byte strings, e.g.

   \textbf{string*10}

   for a 10-byte character string or
opaque*255

for a 255-byte opaque byte string.

The numerical data types can appear as matrices with arbitrary rank (i.e. number of dimensions). Along with the rank, the number of elements in each dimension must be specified, e.g.

real*4[100,100]

for a 100×100 floating-point matrix, or

integer*4[.,2,..]

for a three-dimensional integer matrix whose first and third dimension is specified in the data stream.

2.2.2 Composite data types

The composite data types are arrays and structures. Arrays are repetitions of data elements of the same type accessed by integer indices, while structures are sequences of data elements with different types accessed either by names or integer indices. A variation of the structure is the union, in which only exactly one of many elements of the structure is actually present.

An example of an array is

array[100] of integer*4

for an array of 100 4-byte integers. The number of elements may also be specified in the data stream, e.g.

array[.] of array[3] of real*4

is an array in which each element consists of three real numbers. The number of elements is specified in the data stream.

Structures are specified in the following syntax:

struct {
    atoms : array of
        struct {
            z : integer;
            partial_charge : real;
        }

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bonds : array of struct {
    from_atom : integer;
    to_atom : integer;
    type : integer;
};
positions : array of integer[3];
optional velocities : array of integer[3];

This structure has four fields named atoms, bonds, positions, and velocities. The optional specifier indicates that this field may or may not be present in the data stream.

2.2.3 Implementation

Any data type is represented by a subclass of class SomeType. It provides a method typeP(t) that returns true if the data type is t, where t is one of the constants nilType, numType, stringType, arrayType, structType, or unionType defined in SomeType, similar to a dynamic cast. It also defines a virtual method print() to print the data type and a static method parse() to parse a textual type specification.

The class NumType defines numerical data. It stores the base type of data (one of the enumeration constants i1, u1, i2, u2, i4, u4, i8, u8, f4, f8, f16) and an array of the dimensions of the matrix. A special value dimFree is used for variable-sized dimensions. All this information can be accessed using accessor methods or specified in the constructor.

Similarly, the class StringType defines byte-string data. It stores the number of bytes (or dimFree for variable-sized data) and a flag to indicate whether the data is character or opaque. In the latter case, it will not be subject to any character-set conversion.

Structured data types are represented by the class StructType. It stores an array of fields, each of which is defined by its name, its type and a flag to indicate whether it is optional. Accessor methods allow to access fields by index or by name. Structures a constructed empty, and a method addField(name,typ) is used to add fields. Unions are also represented by StructType, using a special flag that indicates that the structure is to be treated as a union.
Arrays use the ArrayType class which stores the number of elements and the type of the elements. Again, the size can be given as the constant dimFree to indicate variable-sized arrays.

Parsing of textual type specifications is done by the static member function parse() in SomeType. The syntax is chosen such that the first word of the type specifications indicates which subclass the type belongs to, so that the parser can then invoke the static member function parse() in the corresponding subclass.

2.3 The data object class

The class SomeData provides the basic interface for an application to manipulate data. We chose not to duplicate the hierarchy of types classes in corresponding data classes but to include accessor methods for all types of data in a single class. Most accessor routines for data return objects of the class SomeData, so this approach saves the programmer from tedious recasts. When a method is called that is improper for the object's data type, it can either return a null object, or throw an exception.

More important, we wished to include garbage collection by reference counting in the implementation. This is only possible if the application program does not use pointers to access objects of the class SomeData. Instead, the functionality of the access layer is split in two parts: Its actual functionality is provided by the class SomeDataImpl, that can be subclassed by the different data representations, while objects of class SomeData contain a reference-counted pointer to an object of SomeDataImpl. Application programs thus can manipulate objects of class SomeData like pointers (or handles). Most methods in SomeData either pass through directly to the corresponding methods in SomeDataImpl or implement some convenience function that can be expressed in terms of these methods. This is especially advantageous as many operations on SomeData are expressed by overloaded functions, e.g. assign() to assign any type of elementary data. Classes that derive from SomeData and implement just one of these operations, e.g. integer assignment, would have to implement all overloaded versions of assign(). In SomeData, these operations are seperated into functions like assignInt() or assignString() that provide default implementations (namely throwing the appropriate exception) and can be overloaded individually.

SomeData provides a method typeP(t) to test if the data object is of type \( t \), and a method typ() that returns of pointer to its type (represented
by an object of class `SomeType`). It also defines the method `copy()` to copy the contents of one data object into another, assuming that the types are identical. The assignment operator and copy constructor is defined in that way.

For array and structure types, `SomeData` contains a convenience access operator by overloading the indexing operator `operator[]`. If used with a string argument, it accesses a field of a structure, while with an integer argument, it can be used on both array and structures to access by index. Its return value is another object of class `SomeData`. This makes accessing structures and arrays nearly as simple as accessing the corresponding native data types in C++, however, access to fields by name comes with some performance penalty as the character string must first be matched to the names of all fields. This is a general problem with languages that do not provide a symbol data type (as in LISP): The access would be faster if the compiler could convert the string into a more easily manipulated quantity like a 32-bit number that could then be used to perform a hash or binary search in the field table.

For structures, the indexing operator maps to a method `getField()`. A method `nFields()` returns the number of fields in the structure, and `getFieldName()` the name of each field. For optional fields, `unsetField()` removes the field while `fieldPresent()` checks whether the field is present in the actual data. For unions, `getActiveField()` and `setActiveField()` are used to define which field is used.

For arrays, a method `nElements()` returns the number of elements in the array, and `getElem()` is used to access elements. `resize()` can be used to resize the array to a specified number of elements.

Elementary data are read by the methods `getInt()`, `getDouble()`, and `getString()`. Each returns the corresponding C++ data type, or throws an exception if it does not match the actual data type. A method `assign()` with suitable argument types is used to assign data values.

### 2.4 Matrix data

If the elementary data is a matrix, it is represented in C++ by objects of the utility class template `Matrix<T>` where T is the elementary C++ data type. Associated with each matrix is a shape of class `MatrixShape` that stores rank, minimum and maximum indices in each dimension and information about the storage layout. The methods `getShape()` and `setShape()` are used to
manipulate the shape of the data. To access the data itself, the methods `getData()` and `assignData()` read and write the actual representation in memory (as defined by the shape).

2.5 Implementation classes

Actual implementations of data objects are provided by subclassing the class `SomeDataImpl`. Its member functions are similar to the member functions of `SomeData` without convenience functions. Its only member field is `thetype` which is a reference-counted pointer to its data type object.

2.5.1 Direct representation

The class `DirectData` provides an in-memory representation for structured data in a linked tree. Its subclasses `DirectStructData`, `DirectArrayData`, `DirectNumData`, `DirectMatData`, and `DirectStringData` are modelled after the subclasses of `StructType` and provide storage for the respective data types.

These classes are also used to define a simple text file representation of the data. Each implements a static member function `read()` to read tokens from a lexical parser and convert them to an appropriate object. The data format is simple: numbers are represented naturally, strings are quoted, and arrays and structures are surrounded by brackets or braces and their elements separated by (optional) commas. Matrix data are also represented by bracketed lists of numbers, with free dimensions specified in front of the data. Similarly, `DirectData` objects know how to print themselves.

These methods already allow a complete implementation of the structured data format. Their main shortcoming is that the data file must be read as a whole and converted to the `DirectData` format, before it can be accessed by the application.

2.5.2 Binary-file representation

Similarly to the textual representation, a sequential binary stream representation is defined (see sec. 3 for more details). Data fields follow each other without intervening structure information, except for length information of variable-sized arrays and matrices and tag bytes for optional fields and unions.
The class *StreamData* provides read access to such files. Each data item is represented simply by its position in the file. Elementary data are accessed by reading the bytes at the specified position into memory. To get a member of a structured data object, the position of the member is calculated. As the members are in general of variable size, this usually involves reading all the members before the requested item (at least so far, that their size can be determined). This could be avoided by adding size fields to all composite data types, but is not implemented in the basoc data format to keep it as simple as possible.

The big advantage of *StreamData* is that only the requested parts of the data files are held in memory. The binary format for this implementation was designed to be as simple as possible and, in particular, easily writable from FORTRAN programs. However, write access to such files is not as simple, since this kind of data format requires the data to be written in sequence. To provide a convenient representation for writing such data, we once again resort to the *DirectData* implementation and provide a function to write a *DirectData* in binary format to a file. This (as well as reading data) is performed by a class *BinaryDataIO* that encapsulates the parameters of the binary representation and itself used C++ streams.

It is, however, desirable to provide a mechanism to write data in smaller chunks instead of having to store the whole output file in memory. To do so, we need a way of specifying a part of a data structure. This is done by the *SomeDataIterator* class. An object of this class is a reference to a *SomeData* object somewhere in a composite data object. The method *next()* moves the pointer to the next object in the tree on the same level. If the object is composite, *hasSubs()* is true and a method *down()* can be used to access the first member object. After the last object on a level has been retrieved, and end-of-file condition is raised, and the application can use *up()* to go up one level and continue with *next()*.

In this way, all objects in a data tree can be retrieved in exactly the sequence in which they are written in the data format.

The method *writeBinaryRegion()* in *BinaryDataIO* writes the data between to *SomeDataIterators* to disk. To fill in the length information of variable-sized arrays, it keeps a region stack that contains the byte offsets of all composite data objects that enclose the current object.

Using these methods, the class *DataFile* that implements data files provides a method *commit(D)* to commit all data up to but not including the data item *D* to disk. Before they are written to disk, the data are stored in
a DirectData object associated to the data file. After they are written, the corresponding parts of the data tree are deallocated and the memory thus freed.

This procedure may be somewhat unsatisfactory as it does provide full flexibility. However, an implementation that allows filling in data objects in arbitrary order can be achieved only using a more advanced data format.

2.5.3 MallocFile representation

A more flexible data format can be provided by using a MallocFile. This is a flexible block-structured file whose blocks can be manipulated similarly to the blocks in the C heap. Blocks can be allocated to arbitrary size and returned to the free list in any order. Each block is identified by an address and be accessed by acquiring a handle object based on the address. As long as the handle exists, a copy of the block is locked in memory for manipulation and written back when the handle is released.

The simplest mapping of structured data to a MallocFile is to represent each data object by exactly one block. Composite objects then contain a list of addresses that identify blocks that represent their members. When an array is resized, it then suffices to resize the block that contains the array. Unfilled member fields can be represented by null pointers and filled at will.

The format can be improved by not allocating a block to each data objects. Instead, data objects of constant size can be stored directly in their parent blocks, in place of the pointers. Thus, an array whose objects are of constant size can be stored in a single block. A simple rule determines the storage layout: If an object has variable size, a pointer to the object is stored, otherwise, the object itself shows up. An object has variable size, if it is a variable-sized elementary data item (like a matrix with free dimensions), or if it is a structure with variable-sized or optional fields, or if it is an array with a variable-sized element type or with an unspecified number of elements.

3 Data formats

3.1 Binary data format

The primary binary data format has been designed with simplicity in mind. In particular, it can be written from FORTRAN 77 or C simulation programs without using the C++ library. The format is basically a sequential
byte string format in which the data fields are written in the sequence in which they appear when traversing the type tree without intervening meta-information. The only exception are length specifications for variable-sized items (matrices or arrays), tag bytes for optional structure fields, and selector tags for unions. There are no alignment requirements for data items.

The binary data stream is preceded by an ASCII portion of the file that contains an identification line with some metainformation, followed by the textual representation of the data type. Simulation programs can write this part easily as a string constant. The following is an example of such a header portion:

```plaintext
STRUCTURED FILE V0.1 BINARY_BE
#@Date= 18. 3.1998   Time: 15:26
TYPE
struct {
  molecule_description : struct {
    molecule_name: string;
    atom_classes : array of struct {
      atom_class_id : integer*4;
      atom_class_number : integer*4;
      atom_class_name : string;
    };
    atoms : array of struct {
      atom_id : integer*4;
      atom_name : string;
    };
    bonds : array of struct {
      bond_from_id : integer*4;
      bond_to_id : integer*4;
      bond_type : integer*4;
    };
    timesteps : array of struct {
      global_obs : real*4[.];
      coordinates : real*4[3,.];
      optional velocity : real*4[3,.];
      optional potential : struct {
        bb : real*4[3,2];
      };
    };
  };
};
```
The first line starts with the constant "STRUCTURED FILE" to identify the file type, followed by a version specification and optional keyword, here "BINARY_BE" indicating that the data are written in binary format with big-endian byte order. Lines starting with hash signs are comment lines. The keyword TYPE initiates parsing of the type tree textual representation. The binary data stream starts immediately after the end of the line containing the "DATA" keyword.

The following is the specification for the binary data stream:

1. Elementary data objects are written in their natural representation with the byte-order indicated in the identification line. Floating-point numbers are written in IEEE standard representation.

2. Multidimensional data objects are written in FORTRAN order, i.e. the first index varies fastest. If any dimension is unspecified in the data type, the number of elements in this dimension is written as a 4-byte integer in front of the data. This is only done for unspecified dimensions.

3. Character and byte strings are written byte-by-byte. If they are of unspecified size, the actual size precedes them as a 4-byte integer.

4. Structure and array data are written as consecutive data elements. If the array size is unspecified in the data type, it precedes the data as a 4-byte integer.

5. Optional fields are preceded by a single byte. If this byte is zero, the field is not present and there follow no data.

6. Unions are preceded by a 2-byte integer indicating the index (starting from zero) of the active field. It is followed by the binary data for this field only.
4 Extensions

4.1 Named types

Named types are used to build recursive type trees. In order that a recursive type tree does not lead to a data tree with infinite recursion, recursive types usually appear along with unions. An instructive example of their usage is the following data type that is used to externalize type trees:

```c
typedef TypeDescriptor = union {
    num : struct {
        isFloat : integer*1;
        size : integer*1;
        dim : array of integer*4;
    };
    string : struct {
        isOpaque : integer*1;
        size : integer*4;
    };
    struct : struct {
        isUnion : integer*1;
        fields : array of struct {
            name : string;
            typ : type TypeDescriptor;
            isOptional : integer*1;
        );
    };
    array : struct {
        size : integer*4;
        subtype : type TypeDescriptor;
    };
    named : struct {
        name : string;
    };
} 

type TypeDescriptor;
```

The syntax

```c
typedef typename = type
```

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declares the *typename* to stand in for the *type* wherever *type typename* is used. This enables the *struct* and *array* variants of the union to reference other type descriptor trees. The last line of the example is not part of the named-type definition but declares this type to the application.

### 4.2 Late-type binding

It is not always advantageous to specify the data type separately from the data. As far as the methods of the *SomeData* class are concerned, it is sufficient for the data type to specify that an object is a structure or an array, but not what the member fields are. This decision could be deferred to the moment when the fields are actually accessed. The choice of a strong or early type binding that determines the complete type tree when creating the data object was motivated by efficiency considerations. In particular, if a data structure contains arrays of structures, a late-binding data format is forced to repeat the field names in each element of the array.

However, in many cases it is desirable to be able to specify that any data type may appear in a data object, e.g.

```c
struct {
    ...
    userdata: any;
    lots_of_userdata: array of any;
    ...
};
```

specifies a structure with a field *userdata* that can contain an arbitrary data type, and an array *lots_of_userdata* whose elements can contain any and in particular different data types. This is especially desirable if the underlying data format is a late-binding format where the complete type tree can only be retrieved by reading the whole file.

To be able to store arbitrary data types in a file, the data type must be specified in the data. One way to do so is to use the textual representation of the data type and use the *parse()* static member function of *SomeType*. Another is to make use of the data structure from the example in [1.1]. It contains all the fields necessary to externalize a complete type tree. The class *SomeType()* provides methods to convert type trees from and to data objects using this type specification. A data format that wishes to make use of the *any* type will create a data object of this type, have it filled in by these
methods and arrange to have it written along with the actual data object that contains the \texttt{any} data.

To the programmer, the \texttt{any} type is completely transparent when reading data. In particular, no instance of \texttt{SomeData} should ever be of the \texttt{any} type when reading. It will only appear in the type description of members of composite data types. Whenever such a member is accessed by the overloaded indexing operator, it will return the actual data type found in the data file.

When writing data, a data type will be of the \texttt{any} type until data is actually written to it. Before you can assign data to an \texttt{any} type data object, you must specify the actual data type \( t \) by using the \texttt{actualizeType}(\( t \)) method of \texttt{SomeData}. After this method has been called, the data object behaves as if it was of the type \( t \).

Objects of class \texttt{any} are implemented by means of forwarding. Their implementation contains a pointer to the target data object whose type is specified by \texttt{actualizeType()}. Each method of \texttt{SomeDataImpl} forwards to the corresponding method in the target data object when the forwarding pointer is set. Otherwise, the \texttt{any} type object behaves as a null object, illegal to read or write.

\section*{4.3 Object persistence strategies}

An object persistence mechanism, in theory, relieves the programmer of the need to program any input/output. Instead, the compiler and/or the runtime system take over the task of reading and writing objects from or to external storage. This could, in theory, be achieved by a precompiler, but there are several conceptual problems:

\begin{itemize}
  \item A class may contain temporary fields, or fields that have meaning only when the object is in a certain state. The programmer probably does not want these fields to be made persistent, especially if these fields are pointer fields. Handling this requires the introduction of a special keyword to mark persistent fields in an object.
  \item C++ relies heavily on pointers which cannot be externalized. They must be replaced by appropriate object handles, and consequently care must be taken that all objects that are referenced are also made persistent.
\end{itemize}
• Data structures are often implemented by means of template classes, e.g. the Standard Template Library. These template classes must also be made persistent.

Most of these problems stem from the requirement that C++ be efficient and compatible with C. Languages like Java avoid these problems by disallowing pointers or template classes.

Is there a way to achieve something similar without using a precompiler? In Java, a reflection mechanism makes it possible to read out the structure of any data type at runtime. Using such a mechanism, we could provide a method that takes an arbitrary object and constructs an appropriate data type for its externalization. Such a mechanism is currently not available in C++. In particular, there is no way to get a list of the fields in a structure though this could be envisaged as a part of a more general template implementation.

An alternative is the approach taken by Sun’s XDR: Define an abstract data description class whose methods can be used to built data objects. Then, in each persistent class, add a single method that takes a pointer to such an abstract data description object and invoke the methods corresponding to the fields of the class:

```c
struct S : public ADRInterface {
    int a;
    float b;
    struct S2 c;
    struct S3 d[5];

    void adr(ADROperation *op) {
        op->adrInt("a", a);
        op->adrFloat("b", b);
        op->adrStruct("c", c);
        op->adrArray("d", d, 5);
    }
};
```

This can be further enhanced by using overloaded functions for describing the data. The methods `adrStruct()` and `adrArray()` here expect that their second argument, i.e. the data object, is subclassed from `ADRInterface` so they can call the `adr()` method to obtain the structure of the objects.
To use such an interface for the *SomeData* class, three different *ADROperation* objects must be defined: one to obtain the type tree, and one each for reading and writing. Using these operations, any object that implements the *adr()* method can be automatically converted into a *SomeData* object and back.

## 5 Conclusions and outlook

The library presented here presents a uniform interface for handling persistent structured data objects in C++. Together with the simple binary file format defined here, it enables applications to store data on disk in an easily interpreted but highly flexible format. It relieves the programmer from the burden to define binary file formats and moves the data exchange specification to the level of symbolic field names. Practical experience shows that this is the most important feature of the library as it allows to add more fields to a data format without interfering with already existing programs.

For reasons of performance, more sophisticated data representation can be added to the library. One such, based on the MallocFile, has been discussed above, others could include interfaces to established data formats like HDF or to relational or object-oriented data bases using SQL.

As everyone talks about object-oriented programming, extending the library in this direction can be achieved by adding method members to structures, making it possible to invoke operations on *SomeData* objects. Return type and argument lists of such methods can again be specified by abstract type trees and passes as *SomeData* objects. The actual code executed by a method can be hidden away from the application by the data-set classes, thus making it possible to invoke e.g. a Java method from a C++ program. However, this is exactly the feature provided by distributed object systems like CORBA or ILU [5].

This also shows that abstract type trees and a uniform interface to structured data is not restricted to object persistence. Java recently introduced a reflection interface [6] that makes type trees of Java objects accessible from Java code at run-time, and as not everybody can move to Java, especially in scientific computing, a simple kitchen-sink solution like the one presented here might ease some of the everyday problems in C++ programming.

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