

REVIEW ARTICLE

Object Oriented Programming and Numerical Methods

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Abstract: The paper examines the ways in which the object oriented approach to programming can be applied to engineering problems. The principles of object oriented programming are described in terms of the finite element method, and an object oriented implementation of an equation solver for an MS-DOS system is described in detail. These two examples are used to illustrate the added flexibility that can be gained by using the object oriented approach when writing numerical method programs.

INTRODUCTION

One of the fields of software development that is currently very active is that of object oriented programming (OOP). The key feature of OOP is that data and procedures are intrinsically linked. This will be described in detail a little later, but the main advantages of this are that it allows the production of more flexible and less bug prone code. OOP can be implemented either via languages specifically designed around the OOP concept (e.g. 'Smalltalk') or by extensions to more traditional languages (e.g. C++ and Pascal). Most of the current applications of OOP appear to be graphics and database programs. This paper will look specifically at OOP in the context of numerical methods, i.e. in the context of a field that is perhaps more familiar to engineers.

The last 10 years have seen an enormous increase in the power of the personal computer (PC), a trend which is continuing. This has meant that computers have changed from being remote devices that are used only by specialists to being something that almost anyone can have ready access to. This has been coupled with great improvements in the quality and user-friendliness of available software. Following some way behind this

trend is a similar development in finite element (FE) packages. Originally FE packages were of necessity implemented solely on mainframe computers; they also tended to be difficult and tedious to use. Now there is an increasing number of them available on PCs, and data entry and results analysis (pre- and post-processors) are becoming much easier to use.

FE programs, and indeed most numerical method software, are written predominantly in Fortran. In the early days of FE software this was the best way of doing it. This paper will describe how an OOP approach can be applied to engineering software based around numerical methods such as the FE method. For the sake of definiteness all programming will be described in terms of Turbo Pascal (version 6·0) on an MS-DOS system, but it is emphasised that the concepts illustrated apply to OOP languages in general.

One particular example that will be studied is the implementation of an equation solver, such as is at the heart of every FE program, on an MS-DOS based PC. Two of the important features of MS-DOS are: (i) MS-DOS works in memory segments of up to 64 kb in size; (ii) MS-DOS cannot directly access more than 640 kb of memory. This means that MS-DOS programs are better at handling arrays of size less than 64 kb than they are at handling larger ones. Some compilers need to be told that a program contains large arrays, while others such as Turbo Pascal do not allow any one data item to occupy more than 64 kb, and two or more 'arrays' would have to be used to hold a single large array. Whatever the case, using large arrays involves an additional overhead in MS-DOS. FE programs often use large amounts of data, breaking the 640 kb limit. Many computers now have significantly more than 640 kb of memory, and MS-DOS can access this memory using expanded memory managers (EMS), but again there will be some loss in speed. These three situations will be

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referred to as 'memory models', and will be called small, large and EMS:

- 1. Small—all arrays less than 64 kb.
- 2. Large—arrays greater than 64 kb.
- 3. EMS—use expanded memory.

A program that used EMS all the time could be written. Such a program could solve both large and small problems, but when solving small problems would be unnecessarily slow. On the other hand, a program that used a small memory model would be severely limited in the size of problem it could deal with. The ideal solution is a program that decides which memory model to use only when it knows the size of problem that it is dealing with at the time. This paper will describe how such a program can be written using OOP. OOP is not the only way of doing it, but it is the most elegant way.

The power of PCs, specifically the development of the 386 and 486 chips, has now outstripped the power of MS-DOS. However, the sheer weight of investment of time and money in MS-DOS software means that it is likely to remain the most common PC operating system for some time yet. Therefore, the work that is presented here is of direct use. However, the main purpose of the paper is to illustrate how an OOP approach can be of benefit to engineering programming, and the examples described should be viewed primarily in this light.

OBJECT ORIENTED PROGRAMMING

Descriptions of the OOP concept can be found in many computer journals^{3,4,6} and language user guides. In the majority of them the descriptions are given in terms of graphics programs or general organisational problems, though there are some that deal with mathematical applications.² Here the concept will be described directly in terms of the FE method. The hope is that this will enable engineering programmers to understand the concept and see its potential more easily.

In traditional programming languages, such as Fortran, data and subroutines are separate entities and are linked together only by the passing of data to a subroutine when it is called. Indeed the data structures in Fortran are also rather limited: integers, reals, logical, characters and arrays being the only types available in standard Fortran 77. Other languages allow more complex data structures such as records in Pascal. So a record such as the following could be defined in Pascal:

```
elem_data = record

material: integer;

no_nodes: integer;

node_list: Anode_list;

end;
```

where

```
Anode\_list = array[1..8] of integer;
```

In Fortran the three sets of data contained within elem_data would need to be stored separately and it would be up to the programmer to ensure that they always appeared together at the right time. With records the three sets of data are intrinsically linked and will always be together; this appears to be more logical as the sets of data are actually linked. However, the data and the procedures (Pascal's equivalent of subroutines) are only linked when the procedure is called. OOP goes one step further and intrinsically links procedures and data—the procedures are commonly called methods in OOP terminology. So an object such as the following could be defined:

```
element = object
  data: elem_data;
  procedure get_stiff_mat(var nsize:integer; var k: Akmat);
end;
```

where

```
Akmat = array[1..126] of single;
```

(The number 126 is the size needed to store a symmetric 16×16 matrix; if necessary the arrays *Anode_list* and *Akmat* could be made larger.)

If elem was a specific instance of an element object, a call to the method get_stiff_mat would look as follows:

```
elem.get_stiff_mat(n,kmat);
```

The part of the program making this call would then obtain the stiffness matrix for *elem*. Note that there was no explicit need to pass the node list, material number, etc., to *get_stiff_mat* as the procedure and the data are automatically linked together.

Linking data and procedures is logical, and this should lead to clearer thinking and hence to better program design and fewer errors. However, the real power of OOP only becomes evident when two further features are considered: inheritance and virtual methods.

Inheritance and virtual methods

OOP allows descendants of objects to be defined, and a descendant automatically inherits all the data and the methods belonging to its ancestor. However, some of the descendant's methods, while being generally similar, might be different in detail. The use of virtual methods allows this to be taken care of. Again, consider a FE. Certain things are true of all elements, for instance, they all consist of a collection of nodes and are made of a certain material. So the data part of the definition of an element given above applies to any sort of element. Any

type of element object should inherit this data. In addition, stiffness matrices can be obtained from any sort of element, but the precise way the stiffness matrix is calculated depends on the type of element, e.g. beam, plane stress or plate. So each type of element object would have a different method to calculate the stiffness matrix.

Now consider the assembly process in a FE package (this may, of course, be an integral part of the equation solver, e.g. if the frontal solution method⁵ is used). At some stage, the assembler needs to obtain the stiffness matrix of each element, but the assembler is not in the least concerned about how the stiffness matrix was calculated, it only wants to know what it is. This is where virtual methods come in. Change the definition of element to the following:

```
element = object
data: elem_data;
procedure get_stiff_mat(var nsize:integer; var k: Akmat);
virtual;
```

This procedure get_stiff_mat could be a dummy method that did not actually do anything; all the definition is intended to do is to declare that stiffness matrices can be obtained from elements, which is all that the assembler requires. In itself this object is not particularly useful as, at present, it does not actually calculate stiffness matrices. This is done by defining descendants of element. For instance a plane stress element could be defined:

element_ps is a descendant of element and so automatically inherits the data. It is also capable of giving a stiffness matrix, and the procedure get_stiff_mat would contain the code needed to calculate the stiffness matrix. Similarly, beam and plate elements could be defined with their own procedures for calculating stiffness matrices.

Suppose that the program contains an array of pointers to element types:

elements = array[1..1000] of element;

Now the following definition is quite legal:

$$elements[i] := elem_ps \tag{1}$$

where elem_ps is a pointer to type element_ps., i.e. it is legal to assign a descendant to its ancestor. Suppose this was done for all the entries in elements, and elements was then given to the assembler. When the assembler came to try and obtain the stiffness matrix of each element the call to get_stiff_mat would automatically use the plane

stress version of the procedure and so give the correct stiffness matrix. Similarly, if elements had all been defined as beam elements the beam version of get_stiff_mat would have been used. This is because get_stiff_mat is a virtual method and the program only decides which particular version to use at runtime. Statements such as eqn (1) let the program know that element[i] is a plane stress element and so when the program encounters the statement

elements[i] .get_stiff_mat(nsize,kmat);

it knows which one to use, and there is no need to use 'if' statements.

In a complete FE package the element object would contain many more methods. Only one has been shown above, so that the object concepts are not obscured by too much detail.

In order to give further insight into the object oriented approach, an object oriented equation solver will be described in some detail. This example is sufficiently complex as to be worthwhile, but simple enough that the main points are not obscured by detail.

AN OBJECT ORIENTED EQUATION SOLVER FOR MS-DOS SYSTEMS

Finite element equations are commonly solved by using either banded solvers, frontal solvers or active column solvers. The key feature of them all is that they seek to take advantage of the sparsity of FE equations (i.e. the high percentage of zeroes that occur). The active column solver will be considered here.

The details of the method can be found in many standard FE textbooks¹ and only the most important features will be mentioned here. The essence of the method is to decompose the stiffness matrix, **K**, as follows:

$$K = L'DL$$

where L is a lower triangular matrix with ones on the diagonal, and D is a diagonal matrix. The sparsity of K is taken advantage of by using a skyline storage scheme. The L^tDL decomposition can be carried out, using the following algorithm:

for
$$j = 2, ..., n$$

$$g_{m_i,j} = k_{m_i,j}$$

$$g_{ij} = k_{ij} - \sum_{r=m_m}^{i-1} l_{ri} g_{rj} i = m_j + 1, ..., j-1$$

where m_j is the row number of the first nonzero element in column j, and

$$m_m = \max\{m_i, m_i\}$$

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$$l_{ij} = \frac{g_{ij}}{d_{ii}}i = m_j, ..., j-1$$

$$d_{jj} = k_{jj} - \sum_{r=m_j}^{j-1} l_{rj} g_{rj}$$

The l_{ij} are the entries in \mathbf{L}^t . The quantities g_{ij} are intermediate quantities only, and the calculated values of l_{ij} and d_{jj} immediately replace the values k_{ij} and k_{jj} , so only the one matrix \mathbf{K} needs to be used. At the start of the process, \mathbf{K} is the structural stiffness matrix, and at the end the diagonal contains \mathbf{D} and the off-diagonal entries contain \mathbf{L}^t .

The elements will typically be stored by columns in a single array, say A, and an integer array will contain indices of where one column begins and another ends. The L'DL decomposition preserves the skyline shape. In the present context the key feature is the need to access the array A. Traditionally, one would simply define an array of sufficient size to handle the largest problem the program is designed to cope with, and the matrix is accessed directly. As mentioned earlier, on an MS-DOS system there are reasons for wanting a more flexible system. Such a system can be designed, using an object oriented approach if a different view of a matrix is taken. Consider the following typical operations that might be needed to be carried out on A:

```
q: = A[i];

A[i]: = q;

A[i]: = A[i] + q;
```

Instead of using an array, an object could be defined as follows:

```
gen_array = object
  function get_val(i:longint):single; virtual;
  procedure set_val(i:longint; q:single); virtual;
  procedure add_to(i:longint; q:single); virtual;
end:
```

The function get_val returns the value at location i, set_val sets the value at location i to q, and add_to adds q to the value at location i. So, if instead of A being a straightforward array, A is defined to be a gen_array , then the three statements above would be replaced by:

```
q: = A.get_val(i);
A.set_val(i,q);
A.add_to(i,q);
```

So anywhere in a solution algorithm the traditional array statements could be replaced by the appropriate object statement. The active column solver could be written in terms of objects; gen_array would then be extended as follows:

```
gen_array = object
no_eqns : integer;
```

```
index : Iarray;
constructor init(n:integer; idx:Iarray);
function get_val(i:longint):single; virtual;
procedure set_val(i:longint; q:single); virtual;
procedure add_to(i:longint; q:single); virtual;
procedure solve_eqn(var x:Rarray); virtual;
end;
```

no_eqns is the number of equations and index the array used to indicate where the various columns are stored; init is used to initialise the object; solve_eqn is a method containing the active column solution algorithm where array operations are written in object terms; and x is an array which on entry contains the load vector, and on exit contains the solution.

The above definition of gen_array says that a value at location i can be accessed, etc., but does not actually say how it is done. What now needs to be done is to define get_val, set_val and add_to for each of the three memory models. So the following objects are defined:

```
gen\_array\_small = object(gen\_array)
  data: big_array
  function get_val(i:longint):single; virtual;
  procedure set_val(i:longint; q:single); virtual;
  procedure add_to(i:longint; q:single); virtual;
end;
gen_array_large = object(gen_array)
  data: array[1..5] of big_array_ptr;
 function get_val(i:longint):single; virtual;
  procedure set_val(i:longint; q:single); virtual;
  procedure add_to(i:longint; q:single); virtual;
end;
gen_array_ems = object(gen_array)
  data: ems_stream;
 function get_val(i:longint):single; virtual;
  procedure set_val(i:longint; q:single); virtual;
  procedure add_to(i:longint; q:single); virtual;
end;
```

big_array is an array of real numbers whose size is less than 64 kb; big_array_ptr is a pointer to a big_array. This allows big_arrays to be dynamically allocated, so gen_array_large allocates as many big_arrays as it needs, providing there is enough memory. ems_stream is a means of accessing EMS memory.

gen_array_small uses a single array of size less than 64 kb, which is the sort that MS-DOS likes best. For this case the methods would be:

```
gen_array.small.get_val;
var
    ii: integer;
begin
ii: = i;
get_val: = data[ii];
end;
gen_array_small.set_val;
var
```

```
ii: integer;
begin
ii: = i;
data[ii]: = q;
end;
gen_array_small.add_to;
var
    ii: integer;
begin
ii: = i;
data[ii]: = data[ii] + q;
end;
```

The line ii:=i is needed in Turbo Pascal because array indices must be of integer type. For gen_array_small the index i will always be small enough to be stored as an integer, but for the other two memory models this is not the case, so i is declared as a longint (4 byte integer) type.

Corresponding routines would be written for gen_array_large and gen_array_ems. With gen_array_large, get_val would decide which of big_arrays to use, and which entry of that array to access. For gen_array_ems the form of the methods would depend on the EMS access method used. For instance, data could be like a direct access file and get_val and set_val would be like read and write statements.

Note that solve_eqn does not have to be rewritten, this is because it has been written in terms of the object methods. So if A is of type gen_array_small, it would use the gen_array_small versions of get_val, etc. If A was of type gen_array_ems the solve_eqn would automatically know to use the EMS versions.

To implement the above equation solver one would make the following declaration:

```
A: Pgen_array;
```

Pgen_array is a pointer to gen_array. Suppose nsize is the problem size. Once nsize has been determined, the following lines could be included:

```
if nsize < = small_limit then
    A: = new(Pgen_array_small,init(n,index))
else if nsize < = large_limit then
    A: = new(Pgen_array_large,init(n,index))
else
    A: = new(Pgen_array_ems,init(n,index));</pre>
```

Pgen_array_small, etc., are pointers to gen_array_small, etc. The above lines tell the program what sort of gen_array A actually is, and hence which versions of get_val, etc., to use. set_val and add_to could then be used to put in the appropriate stiffness matrix terms. Once this has been done, solving the equations would merely require the line:

```
A^*.solve_eqn(R);
```

The important point to note is that the only decision

lines that needed to be put in by the programmer are the ones above which allocate a; the object programming language automatically takes care of the rest. This makes the programming easier and reduces the scope for introducing bugs. The only difference between the three memory models is the means of accessing data. The solution algorithm is exactly the same and, using an OOP approach, only needs to be written once. The programmer needs to define the appropriate means of data access for the various memory models, and does not have to work out where all the places in the program are that this might make a difference, i.e. 'if' statements do not need to be added to take care of the various possibilities.

COMPUTATIONAL COST OF USING VIRTUAL METHODS

One objection that might be raised is that the method of implementing gen_array_small is unnecessarily involved when all the program needs to do is to access a single array, and there will be an associated cost in computation time. It is true that there is some overhead involved in implementing virtual methods, but this has to be balanced against the increase in flexibility. Moreover, steps can be taken to reduce this overhead. For instance, there are stages in the algorithm where a contiguous sequence of numbers need to be accessed. Instead of accessing each one individually, methods could be written to access them all at once. So the object definition would now become:

The methods get_values and set_values access n contiguous values beginning at location i, get_values retrieves these into rec, and set_values writes the values in rec into the array. Different versions of each of these routines would be written for each of the memory models. If it was felt that the cost of using the above access methods was too high for the small memory model, then a special version of $solve_eqn$ could be written for gen_array_small , which used direct array access instead of the methods get_val , etc. The definition of gen_array_small would then be:

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It is important to note that giving gen_array_small its own solve_eqn method is the only change that is needed. There would be no changes elsewhere in the program.

In order to give some idea of the computational times involved, calculations were carried out on a thin plate subjected to uniform plane stress.

Symmetry was used so that a quarter of the plate was analysed. The plate was split up into a uniform mesh of 225 four-node elements, with the nodes numbered by rows. Calculations were carried out on a 16 MHz 80386 machine using various memory models. The results are shown in Fig. 1. Small 2 refers to gen_array_small, using methods to access data, and small 1 refers to gen_array_small, using direct array access. Using the latter method produces a saving of about 2.4 s. Small 2 is about 0.4 s quicker than the large model, and the EMS model is about 6.2 s slower than the large model. For the

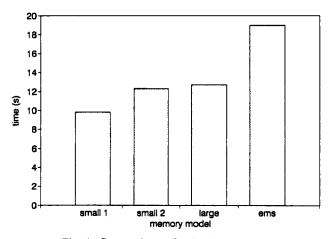


Fig. 1. Comparison of calculation times.

above problem, memory model small 1 clearly gives the best results, but if a larger problem were to be solved this model would not be able to handle it. While the EMS model is significantly slower than the large model, it can handle much larger problems.

SUMMARY AND CONCLUSIONS

The paper has described how the object oriented approach to programming can be used in programs centred around numerical methods. OOP allows flexible programs to be written more easily, and makes for less bug prone programs. The main advantage comes when a part of a program is doing essentially the same thing, but where the details may vary from application to application. One example is finite elements where all elements give stiffness matrices and load vectors, but the way these are calculated varies from case to case. The example of an equation solver was presented in some detail. The same algorithm was used throughout, but different data access methods were used, depending on problem size.

There is some overhead involved in using objects, but against this is the fact that it allows a program to make the best use of the facilities available on the computer. The examples described in the paper are useful in their own right, but it is hoped that they give engineers an appreciation for how an object oriented approach to programming can offer advantages.

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