

Optimizing the matrix computations in environmental models

1. *Types of the matrix computations that appear in environmental models*
2. *Matrix computations related to the pseudo-spectral method*
3. *Matrix computations related to finite elements and finite differences*
4. *Matrix computations related to the chemical sub-model*
5. *Dense matrix computations*
6. *Implementation of a special sparse matrix technique*
7. *Numerical results*
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Types of matrix computations

- **Using FFTs in the computations**
 - **Computations with different banded matrices**
 - **Computations with general sparse matrices**
 - **Computations with dense matrices**
- 1. A large part of the computations is related to matrices**
 - 2. The implementation of efficient techniques for optimizing the matrix computations is crucial**

Matrix computations related to the pseudo-spectral method

1. FFTs - reduction of the number of computations from N^2 to $N \log(N)$
2. Development for different algorithms for different computers
3. Swarztrauber's algorithm
4. Using **multiple** FFTs (Temperton's algorithm)
5. NAG FFTs
6. Difficulties (sometimes transposed matrices have to be formed and used, special order is very often needed)

Matrix computations related to finite elements and finite differences

- Banded matrices, sometimes symmetric and positive definite, sometimes tri-diagonal
- Typical operations: matrix-vector multiplications, factorizations, back substitutions and inner products
- Standard libraries used: LAPACK, NAG Library, SCALAPACK
- Exploiting the symmetry: leads to reduction of the storage requirements, but degrades the speed of the computations
- Recursive algorithms (Andersen, Wasniewski and Gustavson, 2001)

Matrix computations related to the chemical sub-model

- **QSSA - practically no matrix computations are used with this method**
- **Classical ODE methods: calculation of the Jacobian matrix, formation of the shifted Jacobian matrix, factorization, back substitution**
- **Partitioned methods: the same operations as above, but only for the diagonal blocks (the strong blocks)**
- **Using conjugate gradient type methods (matrix-vector multiplication and inner products). Preconditioning might be needed**

Dense matrix computations

- Direct use of sparse matrix packages for general matrices (Duff, Erisman and Reid, 1986, Zlatev, 1991) is **not** efficient because the matrices are small and the number of non-zero elements is relatively high).
- The use of **dense matrix** techniques is **more efficient** than exploiting the sparsity by applying **general-purpose** sparse matrix software
- Used packages: LAPACK, BLAS, NAG Library
- Recursive algorithms can again be applied
- General properties of the dense matrix computations: (i) high performance can be achieved (regular structure), but (ii) the number of computations is high

Special sparse matrix techniques

Why is a special sparse matrix technique needed?

Disadvantages of the general sparse matrix techniques

- **Implicit addressing**
- **Treatment of fill-ins**
- **Copies of rows and columns to the end of the ordered lists**
- **Garbage collections**
- **Finding a pivotal element**
- **Many short loops are to be carried out**
- **Use of many integer arrays**

Removing the disadvantages

- **No pivoting for numerical stability:** allows us to perform a preliminary reordering (Markowitz type of strategy for reducing the number of fill-ins)
- The **positions** of all fill-ins are determined and locations for the fill-ins are **reserved**
- **A loop-free code** for the calculation of the LU factorization is developed
- **A loop-free code** for the calculation of the back-substitution is developed

The price that has to be paid

- **The pivoting for numerical stability is sacrificed**
- **Loop-free codes tend to be long**

Similar ideas:

Willoughby (1970)

IBM Center (Yorktown Heights)

Sandu et al. (1999)

Iowa University

Swart and Blom (1996)

CWI (Amsterdam)

Numerical results

<u>Method</u>	<u>Computing time</u>	
QSSA-1	12.85	
QSSA-2	11.72	
Euler	15.11	
Trapez.	15.94	
RK-2	28.49	
Part. dense	10.09	Based on Euler
Part. sparse	9.21	Based on Euler

Conclusions and open problems

- **The results can be improved by choosing the right way to handle the matrix computations**
- **Some new techniques (such as recursive computations) have not been tried yet**
- **Iterative methods with preconditioning might improve the performance (the problem of finding an optimal preconditioner is open)**