### Tips and tricks for optimization of Fortran codes

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In this short report, we have shown the simple optimization tips and tricks which can be used in general scientific programming with focus on Fortran.

I.

### Compiler Options;

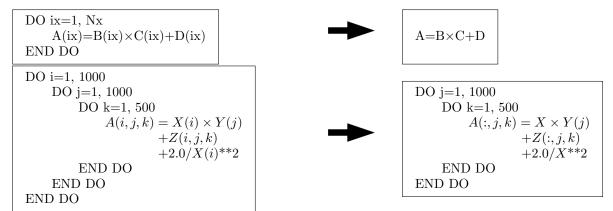
- Substantial gain can be easily obtained by playing with compiler options
- Optimization options are a must. The first and second level of optimization will rarely give no benefits!
- Optimization options can range from -O1 to -O5 with some compilers. -O3 to -O5 might lead to slower code, so try them independently on each subroutine.
- Always check your results when trying optimization options.
- Compiler options might include hardware specifics such as accessing vector units for example.

Intel Fortran and C compiler Options:

ifort, ifc and icc

-O0 -O1 -O2 -O3 -ip -xW -tpp7(for P4) -ip ...

Vectorizing of DO loop; a DO loop can be vectorized when each array calculation is independent of another one



2000 Millions of operations

2 Millions of operations

SUM function; summation by using SUM function instead of DO Loops

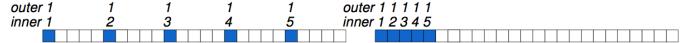
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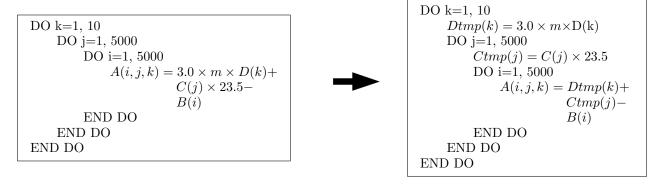
**Array Considerations**; try to minimize the memory jumps, they could be very costly because of cache and TLB misses



## Corresponding memory representation



Minimizing the number of operations; one of the first thing for optimization is reducing the number of unnecessary operations performed by the CPU!



# 1250 Millions of operations 500 Millions of operations

**Complex Numbers**; look for operations on complex numbers that have Imaginary or Real part equal to zero. This is again a question of minimizing the number of operations.

```
! Real part of A elements = 0  
COMPLEX*16 A(1000,1000), B, C(1000,1000)

DO j=1, 1000  
DO i=1, 1000  
C(i,j) = A(i,j) \times B  
END DO  
END DO
```

```
REAL*8 AI(1000,1000)

COMPLEX*16 B, C(1000,1000)

DO j=1, 1000

DO i=1, 1000

C(i,j) = \left(-IMAG(B) \times AI(i,j), AI(i,j) \times REAL(B)\right)
END DO

END DO
```

## 6 Millions of operations

## 2 Millions of operations

Loop Overhead and Objects declarations and instanciations; in Object-Oriented Languages AVOID objects declarations and instanciations within the most inner loops

```
DO j=1, 1000000

DO i=1, 1000000

DO i=1, 1000000

DO i=1, 1000000

DO i=1, 1000000

A(i, j, 1) = B(i, j) + C(1)

A(i, j, 2) = B(i, j) + C(2)

END DO

END DO

END DO

END DO
```

#### Function Call Overhead;

```
DO k=1, 10000
   DO j=1, 10000
       DO i=1, 5000
           A(i,j,k) = F1\Big(C(i),B(j),k\Big)
                                                         DO k=1, 10000
        END DO
                                                             DO j=1, 10000
                                                                 DO i=1,5000
    END DO
                                                                     A(i, j, k) = C(i) * k - B(j)
END DO
                                                                 END DO
                                                             END DO
FUNCTION F1(x,y,m)
  REAL*8 x,y,tmp
                                                         END DO
 INTEGER m
 tmp=x*m - y
  RETURN tmp
END FUNCTION
```

**Blocking**; Blocking is used to reduce cache and TLB misses in nested Matrix operations. The idea is to process as much as possible the data that is brought in the cache.

```
DO i=1, N

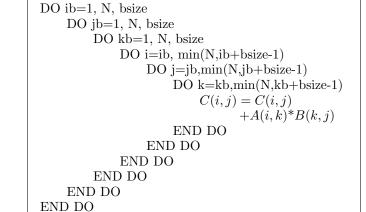
DO j=1, N

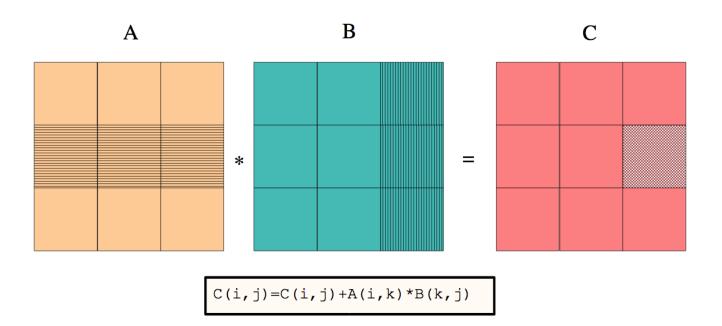
DO k=1, N

C(i,j) = C(i,j) + A(i,k)*B(k,j)
END DO

END DO

END DO
```





**Loop Fusion**; The main advantage of Loop Fusion is the reduction of cache misses when the same array is used in both loops. It also reduces loop overhead and allow a better control of multiple instructions in a single cycle, when hardware allows it (2 FMA or 2 vector operations for example).

DO i=1, 10000  

$$A = A + X(i) + 2.0*Z(i)$$
  
END DO

DO i=1, 10000  
 $B = 3.0 \times X(i) - 5.0$   
END DO

DO i=1, 10000  
 $A = A + X(i) + 2.0*Z(i)$   
 $B = 3.0 \times X(i) - 5.0$   
END DO

**Loop Unrolling**; the main advantage of Loop Unrolling is to reduce or eliminate data dependencies in loops. This is particularly useful when using an architecture with 2 FMA Units (IBM Power3-4) or a Vector unit (SSE2 extensions)

DO i=1, 1000  

$$A = A + X(i) \times Y(i)$$
  
END DO



DO i=1, 1000, 4  

$$A = A + X(i) \times Y(i)$$
  
 $+X(i+1) \times Y(i+1)$   
 $+X(i+2) \times Y(i+2)$   
 $+X(i+3) \times Y(i+3)$   
END DO

**Sum Reductions**; sum reductions is another way of reducing or eliminating data dependencies in loops. It is more explicit than the Loop Unrolling method.

DO i=1, 1000  

$$A = A + X(i) \times Y(i)$$
  
END DO



DO i=1, 1000, 4  

$$A1 = A1 + X(i) \times Y(i)$$
  
 $A2 = A2 + X(i+1) \times Y(i+1)$   
 $A3 = A3 + X(i+2) \times Y(i+2)$   
 $A4 = A4 + X(i+3) \times Y(i+3)$   
END DO  
 $A = A1 + A2 + A3 + A4$ 

Replace divisions by multiplications; Contrary to Floating Point multiplications or additions or subtractions, divisions are very costly in terms of clock cycles.

1 multiplication = 1 cycle 1 division = 14-20 cycles

DO j=1, 10000  
DO i=1, 10000  

$$A(i,j) = \left(B(i) - C(j)\right)/D$$
END DO  
END DO



$$D = 1.0/D$$
 DO j=1, 10000 DO i=1, 10000 
$$A(i,j) = \left(B(i) - C(j)\right) \times D$$
 END DO END DO

Repeated multiplications for exponentials; exponentiation with a small exponent should be done manually. Like divisions exponential operations use many cycles.

$$A = B * *3.0$$



$$tmpc = B * B$$
$$A = tmpc * B$$

**Breaking Interpolations**; the multi-dimensional interpolations should be considered as few one dimensional interpolations

Branching (proper use of IFs); try to minimize as much as possible the use of IFs within the inner loops. The CPU will first assume a YES when it encounters a IF statement while filling up the instruction pipline.