This chapter describes the support provided for writing parallel programs on Origin2000. It assumes that you are familiar with basic parallel constructs. For more information on parallel constructs, refer to Chapter 5, “Fortran Enhancements for Multiprocessors.”

Topics covered in this chapter include:

- “Performance Tuning of Parallel Programs on Origin2000” on page 122
- “Data Distribution Directives” on page 127
- “Nested Doacross Directive” on page 129
- “Affinity Scheduling” on page 129
- “Specifying Processor Topology With the ONTO Clause” on page 132
- “Types of Data Distribution” on page 133
- “Optional Environment Variables and Compile-Time Options” on page 141
- “Examples” on page 142

A subset of the mechanisms described in this chapter are supported for C and C++, and are described in the C Language Reference Manual, Chapter 11. You can find additional information on parallel programming in “Models of Parallel Computation” in Topics in IRIX Programming.

Note: The multiprocessing features described in this chapter require support from the MP run-time library (libmp). IRIX operating system versions 6.3 (and above) are automatically shipped with this new library. If you wish to access these features on a machine running IRIX 6.2, then contact your local Silicon Graphics service provider or SGI Customer Support (1-800-800-4744) for libmp.
Performance Tuning of Parallel Programs on Origin2000

Origin2000 provides cache-coherent, shared memory in the hardware. Memory is physically distributed across processors. Consequently, references to locations in the remote memory of another processor take substantially longer (by a factor of two or more) to complete than references to locations in local memory. This can severely affect the performance of programs that suffer from a large number of cache misses. Figure 6-1 shows a simplified version of the Origin2000 memory hierarchy.

![Origin2000 Memory Hierarchy](image)

**Figure 6-1** Origin2000 Memory Hierarchy

Improving Program Performance

To obtain good performance in such programs, it is important to schedule computation and distribute data across the underlying processors and memory modules, so that most cache misses are satisfied from local rather than remote memory. The primary goal of the programming support, therefore, is to enable user control over data placement and computation scheduling.
Cache behavior continues to be the largest single factor affecting performance, and programs with good cache behavior usually have little need for explicit data placement. In programs with high cache misses, if the misses correspond to true data communication between processors, then data placement is unlikely to help. In these cases, it may be necessary to redesign the algorithm to reduce inter-processor communication. Figure 6-2 shows this scenario.

If the misses are to data referenced primarily by a single processor, then data placement may be able to convert remote references to local references, thereby reducing the latency of the miss. The possible options for data placement are automatic page migration or explicit data distribution, either regular or reshaped (both of these are described in "Regular Data Distribution" on page 133 and "Data Distribution With Reshaping" on page 134). The differences between these choices are shown in Figure 6-2.

Automatic page migration requires no user intervention and is based on the run-time cache miss behavior of the program. It can therefore adjust to dynamic changes in the reference patterns. However, the page migration heuristics are deliberately conservative, and may be slow to react to changes in the references patterns. They are also limited to performing page-level allocation of data.

Regular data distribution (performing just page-level placement of the array) is also limited to page-level allocation, but is useful when the page migration heuristics are slow and the desired distribution is known to the programmer.

Finally, reshaped data distribution changes the layout of the array thereby overcoming the page-level allocation constraints; however, it is useful only if a data structure has the same (static) distribution for the duration of the program. Given these differences, it may be necessary to use each of these options for different data structures in the same program.
Chapter 6: Parallel Programming on Origin2000

Figure 6-2 Cache Behavior and Solutions
Choosing Between Multiple Options

For a given data structure in the program, you can choose from the options described above based on the following criteria:

- If the program repeatedly references the data structure and benefits from reuse in the cache, then data placement is not needed.
- If the program incurs a large number of cache misses on the data structure, then you should identify the desired distribution in the array dimensions (such as BLOCK or CYCLIC, described in “Data Distribution Directives” on page 127) based on the desired parallelism in the program.

For example,

```c
#doacross
do i=2,n
do j=2,n
   A(i,j) = 3*i + 4*j + A(i, j-1)
enddo
enddo
```

suggests a distribution A(BLOCK, *), whereas:

```c
#doacross
do i=2,n
do j=2,n
   A(i,j) = 3*i + 4*j + A(i-1, j)
enddo
enddo
```

suggests a distribution of A(*, BLOCK).

- Having identified the desired distribution, you can select either regular and reshaped distribution based on the size of an individual processor’s portion of the distributed array. Regular distribution is useful only if each processor’s portion is substantially larger than the page-size in the underlying system (16KBytes on the Origin2000). Otherwise regular distribution is unlikely to be useful, and you should use `distribute_reshape`, where the compiler changes the layout of the array to overcome page-level constraints.
For example, consider the following code:

```fortran
real*8 A(m, n)
c$distribute A(BLOCK, *)
```

In this example, the size of each processor’s portion is approximately \( m/P \) elements (\( 8 \times (m/P) \) bytes), where \( P \) is the number of processors. If \( m \) is 1000,000 then each processor’s portion is likely to exceed a page and regular distribution is sufficient. If instead \( m \) is 10,000 then distribute_reshape is required to obtain the desired distribution.

In contrast, consider the following distribution:

```fortran
c$distribute A(*, BLOCK)
```

In this example, the size of each processor’s portion is approximately \( (m \times n)/P \) elements (\( 8 \times (m \times n)/P \) bytes). So if \( n \) is 100 (for instance), regular distribution may be sufficient even if \( m \) is only 10,000.

As this example illustrates, distributing the outer dimensions of an array increases the size of an individual processor’s portion (favoring regular distribution), whereas distributing the inner dimensions is more likely to require reshaped distribution.

Finally, the IRIX operating system on Origin2000 follows a default “first-touch” page-allocation policy; that is, each page is allocated from the local memory of the processor that incurs a page-fault on that page. Therefore, in programs where the array is initialized (and consequently first referenced) in parallel, even a regular distribution directive may not be necessary, since the underlying pages are allocated from the desired memory location automatically due to the first-touch policy.

**New Directives for Performance Tuning on Origin2000**

The programming support consists of extensions to the existing Power Fortran/C directives/pragmas. Table 6-1 summarizes the new directives. Like the other Power Fortran/C directives, these new directives are ignored except under multiprocessor compilation.
The data distribution directives and doacross nest directive have an optional ONTO clause (described in “Specifying Processor Topology With the ONTO Clause” on page 132) to control the partitioning of processors across multiple dimensions.

### Data Distribution Directives

The data distribution directives allow you to specify High Performance Fortran-like distributions for array data structures. For irregular data structures, directives are provided to explicitly place data directly on a specific processor.

The **distribute**, **dynamic**, and **distribute_reshape** directives are declarations that must be specified in the declaration part of the program, along with the array declaration. The **redistribute** directive is an executable statement and can appear in any executable portion of the program.

---

**Table 6-1** Summary of New Directives

<table>
<thead>
<tr>
<th>Directive</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>c$distribute A (&lt;dist&gt;, &lt;dist&gt;, •••)</td>
<td>Data distribution</td>
</tr>
<tr>
<td>c$redistribute A(&lt;dist&gt;, &lt;dist&gt;)</td>
<td>Dynamic data redistribution</td>
</tr>
<tr>
<td>c$dynamic A</td>
<td>Redistributable annotation</td>
</tr>
<tr>
<td>c$distribute_reshape B(&lt;dist&gt;)</td>
<td>Data distribution with reshaping</td>
</tr>
<tr>
<td>c$page_place (&lt;addr&gt;, &lt;sz&gt;, &lt;thread&gt;)</td>
<td>Explicit placement of data</td>
</tr>
<tr>
<td>c$doacross affinity (i) = data (A(i))</td>
<td>Data-affinity scheduling</td>
</tr>
<tr>
<td>c$doacross affinity (i) = thread (&lt;expr&gt;)</td>
<td>Thread-affinity scheduling</td>
</tr>
<tr>
<td>c$doacross nest (i,j)</td>
<td>Nested doacross</td>
</tr>
</tbody>
</table>

**Note:** 
1. `<dist>` can be one of BLOCK, CYCLIC, CYCLIC(<expr>), or “*”. CYCLIC by itself implies a chunk size of 1.
2. For performance reasons, CYCLIC(3) and CYCLIC(k) (where k has a run-time value of 3), may be incompatible when passing a reshaped array as a parameter to another routine.

The data distribution directives and doacross nest directive have an optional ONTO clause (described in “Specifying Processor Topology With the ONTO Clause” on page 132) to control the partitioning of processors across multiple dimensions.
You can specify a data distribution directive for any local, global, or common-block array. Each dimension of a multi-dimensional array can be independently distributed. The possible distribution types for an array dimension are BLOCK, CYCLIC (<expr>), and * (asterisk not distributed). (A CYCLIC distribution with a chunk size that is either greater than 1 or is determined at runtime is sometimes also called BLOCK-CYCLIC.)

A BLOCK distribution partitions the elements of the dimension of size $N$ into $P$ blocks (one per processor), with each block of size $B = \text{ceiling}(N/P)$.

\[
\begin{array}{ccccc}
P0 & P1 & P2 & \cdots & P_{p-1} \\
B & B & B & \cdots & B \\
\end{array}
\]

A CYCLIC($k$) distribution partitions the elements of the dimension into pieces of size $k$ each, and distributes them sequentially across the processors.

\[
\begin{array}{cccc}
P0 & P1 & P_{p-1} & P0 \\
k & k & \cdots & k \\
k & k & \cdots & k \\
\end{array}
\]

A distributed array is distributed across all of the processors being used in that particular execution of the program, as determined by the environment variable MP_SET_NUMTHREADS. If a distributed array is distributed in more than one dimension, then by default the processors are apportioned as equally as possible across each distributed dimension. For instance, if an array has two distributed dimensions, then an execution with 16 processors assigns 4 processors to each dimension ($4 \times 4 = 16$), whereas an execution with 8 processors assigns 4 processors to the first dimension and 2 processors to the second dimension. You can override this default and explicitly control the number of processors in each dimension using the ONTO clause along with a data distribution directive.
Nested Doacross Directive

The nested doacross directive allows you to exploit nested concurrency in a limited manner. Although true nested parallelism is not supported, you can exploit parallelism across iterations of a perfectly nested loop-nest. For example:

```plaintext
c$doacross nest(i, j)
do i = 1, n
    do j = 1, m
        A(i,j) = 0
    enddo
enddo
```

This directive specifies that the entire set of iterations across the \((i, j)\) loops can be executed concurrently. The restriction is that the do-i and do-j loops must be perfectly nested, that is, no code is allowed between either the do-i and do-j statements or the enddo-i and enddo-j statements. You can also supply the nest clause with the PCF directive pdo.

The existing clauses such as local and shared behave as before. You can combine a nested doacross with an affinity clause (as shown below), or with a schedtype of simple or interleaved (dynamic and gss are not currently supported). The default is simple scheduling, except when accessing reshaped arrays (see “Affinity Scheduling” on page 129).

```plaintext
c$doacross nest(i, j) affinity(i,j) = data(A(i,j))
do i = 2, n-1
    do j = 2, m-1
        A(i,j) = A(i,j) + i*j
    enddo
enddo
```

Affinity Scheduling

The goal of affinity scheduling is to control the mapping of iterations of a parallel loop for execution onto the underlying threads. Specify affinity scheduling with an additional clause to a doacross directive. An affinity clause, if supplied, overrides the SCHEDTYPE clause. This section describes the following topics:

- “Data Affinity” on page 130
- “Thread Affinity” on page 132
Data Affinity

The following code shows an example of data affinity:

```
c$distribute A(block)
c$doacross affinity(i) = data(A(a*i+b))
do i = 1, n
   A(a*i+b) = 0
enddo
```

The multiplier for the loop index variable \(a\) and the constant term \(b\) must both be literal constants, with \(a\) greater than zero.

The effect of this clause is to distribute the iterations of the parallel loop to match the data distribution specified for the array \(A\), such that iteration \(i\) is executed on the processor that owns element \(A(a*i+b)\) based on the distribution for \(A\). The iterations are scheduled based on the specified distribution, and are not affected by the actual underlying data-distribution (which may differ at page boundaries, for example).

In case of a multi-dimensional array, affinity is provided for the dimension that contains the loop-index variable. The loop-index variable cannot appear in more than one dimension in an affinity directive. For example:

```
c$distribute A (block, cyclic(1))
c$doacross affinity (i) = data (A(i+3, j))
do i = 1,n
   do j = 1,n
      A(i+3, j) = A(i+3,j-1)
   enddo
enddo
```

In this example, the loop is scheduled based on the block-distribution of the first dimension. Information on `doacross` is in “Nested Doacross Directive” on page 129. The affinity clause is also available with the PCF `pdo` directive.

The default schedtype for parallel loops is SIMPLE. However, under `-O3` compilation, loops that reference reshaped arrays default to data-affinity scheduling for the most frequently accessed reshaped array in the loop (chosen heuristically by the compiler). To obtain SIMPLE scheduling even at `-O3`, you can explicitly specify the schedtype on the parallel loop.
Data affinity for loops with non-unit stride can sometimes result in non-linear affinity expressions. In such situations the compiler issues a warning, ignores the affinity clause, and defaults to simple scheduling.

**Data Affinity for Redistributed Arrays**

By default, the compiler assumes that a distributed array is not dynamically redistributed, and directly schedules a parallel loop for the specified data affinity. In contrast, a redistributed array can have multiple possible distributions, and data affinity for a redistributed array must be implemented in the run-time system based on the particular distribution.

However, the compiler does not know whether or not an array is redistributed, since the array may be redistributed in another function (possibly even in another file). Therefore, you must explicitly specify the `c$dynamic` declaration for redistributed arrays. This directive is required only in those functions that contain a `doacross` loop with data affinity for that array (see “Nested Doacross Directive” on page 129 for additional information). This informs the compiler that the array can be dynamically redistributed. Data affinity for such arrays is implemented through a run-time lookup.

Implementing data affinity through a run-time lookup incurs some extra overhead compared to a direct compile-time implementation. You can avoid this overhead when a subroutine contains data affinity for a redistributed array and the distribution of the array for the entire duration of that subroutine is known. In this situation, you can supply the `c$distribute` directive with the particular distribution and omit the `c$dynamic` directive.

By default, the compiler assumes that a distributed array is not redistributed at runtime. As a result, the distribution is known at compile time, and data affinity for the array can be implemented directly by the compiler. In contrast, since a redistributed array can have multiple possible distributions at runtime, data affinity for a redistributed array is implemented in the run-time system based on the distribution at runtime, incurring extra run-time overhead.

If an array is redistributed in the program, then you can explicitly specify a `c$dynamic` directive for that array. The only effect of the `c$dynamic` directive is to implement data affinity for that array at runtime rather than at compile time. If you know an array has a specified distribution throughout the duration of a subroutine, then you do not have to supply the `c$dynamic` directive. The result is more efficient compile time affinity scheduling.
Since reshaped arrays cannot be dynamically redistributed, this is an issue only for regular data distribution.

**Data Affinity for a Formal Parameter**

You can supply a `c$distribute` directive on a formal parameter, thereby specifying the distribution on the incoming actual parameter. If different calls to the subroutine have parameters with different distributions, then you can omit the `c$distribute` directive on the formal parameter; data affinity loops in that subroutine are automatically implemented through a run-time lookup of the distribution. (This is permissible only for regular data distribution. For reshaped array parameters, the distribution must be fully specified on the formal parameter.)

**Thread Affinity**

Similar to data affinity, you can specify thread affinity as an additional clause on a `doacross` directive (refer to “Nested Doacross Directive” on page 129 for details). The syntax for thread affinity is as follows:

`c$doacross affinity (i) = thread(<expr>)`

The effect of this directive is to execute iteration `i` on the thread number given by the user-supplied expression (modulo the number of threads). Since the threads may need to evaluate this expression in each iteration of the loop, the variables used in the expression (other than the loop induction variable) must be declared shared and must not be modified during the execution of the loop. Violating these rules can lead to incorrect results.

If the expression does not depend on the loop induction variable, then all iterations will execute on the same thread and will not benefit from parallel execution.

**Specifying Processor Topology With the ONTO Clause**

This clause allows you to specify the processor topology when two (or more) dimensions of processors are required. For instance, if an array is distributed in two dimensions, then you can use the ONTO clause to specify how to partition the processors across the distributed dimensions. Or, in a nested `doacross` with two or more nested loops, you can use the ONTO clause to specify the partitioning of processors across the multiple parallel loops.
Types of Data Distribution

For example:

C Assign processor in the ratio 1:2 to the two dimension
real*8 A (100, 200)
c$distribute A (block, block) onto (1, 2)

C Use 2 processors in the do-i loop, and the remaining in the do-j loop
C$doacross nest (i, j) onto (2, *)
do i = 1, n
  do j = 1, m
    • • •
  enddo
enddo

Types of Data Distribution

There are two types of data distribution: regular and reshaped. The following sections describe each of these distributions.

Regular Data Distribution

The regular data distribution directives try to achieve the desired distribution solely by influencing the mapping of virtual addresses to physical pages without affecting the layout of the data structure. Since the granularity of data allocation is a physical page (at least 16 Kbytes), the achieved distribution is limited by the underlying page-granularity. However, the advantages are that regular data distribution directives can be added to an existing program without any restrictions, and can be used for affinity scheduling (see ”Data Affinity” on page 130).

Distributed arrays can be dynamically redistributed with the following redistribute statement:

c$distribute A (block, cyclic(k))

The redistribute is an executable statement that changes the distribution ”permanently” (or until another redistribute statement). It also affects subsequent affinity scheduling.

The csdynamic directive specifies that the named array is redistributed in the program, and is useful in controlling affinity scheduling for dynamically redistributed arrays. It is discussed in ”Data Affinity for Redistributed Arrays” on page 131.
Data Distribution With Reshaping

Similar to regular data distribution, the reshape directive specifies the desired distribution of an array. In addition, however, the reshape directive declares that the program makes no assumptions about the storage layout of that array. The compiler performs aggressive optimizations for reshaped arrays that violate standard Fortran-77 layout assumptions but guarantee the desired data distribution for that array.

The reshape directive accepts the same distributions as the regular data distribution directive, but uses a different keyword, as shown below:

c$distribute_reshape (block, cyclic(1))

Restrictions on Reshaped Arrays

Since the distribute_reshape directive specifies that the program does not depend on the storage layout of the reshaped array, restrictions on the arrays that can be reshaped include the following:

- The distribution of a reshaped array cannot be changed dynamically (that is, there is no redistribute_reshape directive).
- Initialized data cannot be reshaped.
- Arrays that are explicitly allocated through alloca/malloc and accessed through pointers cannot be reshaped.
- An array that is equivalenced to another array cannot be reshaped.
- I/O for a reshaped array cannot be mixed with namelist I/O or a function call in the same I/O statement.
- A COMMON block containing a reshaped array cannot be linked -Xlocal.

Caution: This user error is not caught by the compiler/linker.
If a reshaped array is passed as an actual parameter to a subroutine, two possible scenarios exist:

- The array is passed in its entirety (`call func(A)` passes the entire array `A`, whereas `call func(A(i,j))` passes a portion of `A`). The compiler automatically clones a copy of the called subroutine and compiles it for the incoming distribution. The actual and formal parameters must match in the number of dimensions, and the size of each dimension.

You can restrict a subroutine to accept a particular reshaped distribution on a parameter by specifying a `distribute reshape` directive on the formal parameter within the subroutine. All calls to this subroutine with a mismatched distribution will lead to compile- or link-time errors.

- A portion of the array can be passed as a parameter, but the callee must access only a single processor’s portion. If the callee exceeds a single processor’s portion, then the results are undefined. You can use intrinsics to access details about the array distribution (described in “Query Intrinsics for Distributed Arrays” on page 136).

**Error-Detection Support**

Most errors in accessing reshaped arrays are caught either at compile time or at link time. These include:

- Inconsistencies in reshaped arrays across COMMON blocks (including across files)
- Declaring a reshaped array EQUIVALENCED to another array
- Inconsistencies in reshaped distributions on actual and formal parameters
- Other errors such as disallowed I/O statements involving reshaped arrays, reshaping initialized data, or reshaping dynamically allocated data

Errors such as matching the declared size of an array dimension typically are caught only at runtime. The compiler option, `-MP:check reshape=on`, generates code to perform these tests at runtime. These run-time checks are not generated by default, since they incur overhead, but are useful during program development.

The run-time checks include:

- Inconsistencies in array-bound declarations on each actual and formal parameter
- Inconsistencies in declared bounds of a formal parameter that corresponds to a portion of a reshaped actual parameter.
Query Intrinsics for Distributed Arrays

You can use the following set of intrinsics to obtain information about an individual dimension of a distributed array. Fortran array dimensions are numbered starting at 1. All routines work with 64-bit integers as shown below, and return -1 in case of an error (except `dsm_this_startingindex` where -1 may be a legal return value).

```c
extern INT64 dsm_numthreads (void* array, INT64 dim)
    Called with a distributed array and a dimension number. Returns the number of threads in that dimension.

extern INT64 dsm_chunksize (void* array, INT64 dim)
    Returns the chunk size (ignoring partial chunks) in the given dimension for each of block, cyclic(..), and star distributions.

extern INT64 dsm_this_chunksize (void* array, INT64 dim, INT64 index)
    Returns the chunk size for the chunk containing the given index value for each of block, cyclic(..), and star. This value may be different from `dsm_chunksize` due to edge effects that may lead to a partial chunk.

extern INT64 dsm_rem_chunksize (void* array, INT64 dim, INT64 index)
    Returns the remaining chunk size from index to the end of the current chunk, inclusive of each end point. Essentially it is the distance from index to the end of that contiguous block, inclusive.

extern INT64 dsm_this_startingindex (void* array, INT64 dim, INT64 index)
    Returns the starting index value of the chunk containing the supplied index.

extern INT64 dsm_numchunks (void* array, INT64 dim)
    Returns the number of chunks (including partial chunks) in given dim for each of block, cyclic(..), and star distributions.

extern INT64 dsm_this_threadnum(void* array, INT64 dim, INT64 index)
    Returns the thread number for the chunk containing the given index value for each of block, cyclic(..), and star distributions.

extern INT64 dsm_distribution_block (void* array, INT64 dim)
extern INT64 dsm_distribution_cyclic (void* array, INT64 dim)
extern INT64 dsm_distribution_star (void* array, INT64 dim)
    Boolean routines to query the distribution of a given dimension.
```
extern INT64 dsm_isreshaped (void* array)
    Boolean routine to query whether reshaped or not.

extern INT64 dsm_isdistributed (void* array)
    Boolean routine to query whether distributed (regular or reshaped) or not.

**Implementation of Reshaped Arrays**

The compiler transforms a reshaped array into a pointer to a "processor array." The processor array has one element per processor, with the element pointing to the portion of the array local to the corresponding processor.

Figure 6-3 shows the effect of the `distribute reshape` directive with a BLOCK distribution on a 1-dimensional array. $N$ is the size of the array dimension, $P$ is the number of processors, and $B$ is the block-size on each processor, ceiling = $(N/P)$.

![Figure 6-3](image)

*Figure 6-3*  Implementation of BLOCK Distribution

With this implementation, an array reference $A(i)$ is transformed into a two-dimensional reference $A[i/B][i%B]$ (in C syntax with C dimension order), where $B$ is the size of each block, and given by ceiling $(N/P)$. Thus $A[i/B]$ points to a processor's local portion of the array, and $A[i/B][i%B]$ refers to a specific element within the local processor's portion.
A CYCLIC distribution with a chunk size of 1 is implemented as shown in Figure 6-4.

![Figure 6-4 Implementation of CYCLIC(1) Distribution](image)

An array reference, \( A(i) \), is transformed to \( A[i\%P][i/P] \) where \( P \) is the number of threads in that distributed dimension.

Finally, a CYCLIC distribution with a chunk size that is either a constant greater than 1 or a run-time value (also called BLOCK-CYCLIC) is implemented as Figure 6-5 shows.
Types of Data Distribution

An array reference, $A(i)$, is transformed to the three-dimensional reference $A[(i/k)\%P] [i/(Pk)] [i\%k]$, where $P$ is the total number of threads in that dimension, and $k$ is the chunk size.

The compiler tries to optimize these divide/modulo operations out of inner loops through aggressive loop transformations such as blocking and peeling.

Regular vs. Reshaped Data Distribution

In summary, consider the differences between regular and reshaped data distribution. The advantage of regular distributions is that they do not impose any restrictions on the distributed arrays and can be freely applied in existing codes. Furthermore, they work well for distributions where page granularity is not a problem. For example, consider a BLOCK distribution of the columns of a two-dimensional Fortran array of size $A(r, c)$ (column-major layout) and distribution (*, BLOCK). If the size of each processor’s portion, ceiling = $(c/P)*r*element\_size$ is significantly greater than the page size (16KB on Origin2000), then regular data distribution should be effective in placing the data in the desired fashion.
However, regular data distribution is limited by page-granularity considerations. For instance, consider a (BLOCK, BLOCK) distribution of a two-dimensional array where the size of a column is much smaller than a page. Each physical page is likely to contain data belonging to multiple processors, making the data-distribution quite ineffective. (Data distribution may still be desirable from affinity-scheduling considerations, described in “Affinity Scheduling” on page 129.)

Reshaped data distribution addresses the problems of regular distributions by changing the layout of the array in memory so as to guarantee the desired distribution. However, since the array no longer conforms to standard Fortran-77 storage layout, there are restrictions on the usage of reshaped arrays.

Given both types of data distribution, you can choose between the two based on the characteristics of the particular array in an application.

**Explicit Placement of Data**

For irregular data structures, you can explicitly place data in the physical memory of a particular processor using the following directive:

```
c$page_place (<addr>, <size>, <threadnum>)
```

where `<addr>` is the starting address, `<size>` is the size in bytes, and `<threadnum>` is the number of the destination processor. This directive causes all the pages spanned by the virtual address range `<addr>· · ·<addr>+<size>` to be allocated from the local memory of processor number `<threadnum>`. It is an executable statement; therefore, you can use it to place either statically or dynamically allocated data.

An example of this directive is as follows:

```
real*8 a(100)
c$page_place (a, 800, 3)
```
Optional Environment Variables and Compile-Time Options

You can control various run-time features through the following optional environment variables:

_DSMP OFF
Disables non-uniform memory access (NUMA) specific calls (for example, to allocate pages from a particular memory).

_DSMP VERBOSE
Prints messages about parameters being used during execution.

_DSMP PPM
Specifies the number of processors to use per memory module. Must be set to an integer value; to use only one processor per memory module, set this variable to 1.

_DSMP FOP
Enables the use of hardware fetch operations to implement the synchronization for the completion of a parallel region, more efficient than the default software implementation using synchronization through shared-memory locations.

_DSMP MIGRATION
Automatic page migration is off by default. This variable, if set, must be set to one of the following:

OFF disables migration (the default).
ON enables migration except for explicitly placed data (using page_place or a data distribution directive).
ALL_ON enables migration for ALL data.

_DSMP ROUND ROBIN
Requests round-robin data allocation across memories (rather than first-touch) for all of stack, data, and text segments. The default is first-touch.

MP Sugnumthd
If set, this variable enables the use of dynamic threads in the multiprocessor (MP) runtime. Essentially, with dynamic threads the MP runtime automatically adjusts the number of threads used for a parallel loop at runtime based on the overall system load. This feature improves the overall throughput of the system. Furthermore, by avoiding excessive concurrency, this feature can reduce delays at synchronization points within a single application.

_PAGESIZE STACK, _PAGESIZE DATA, _PAGESIZE TEXT
Specifies the desired page size in kilobytes. Must be set to an integer value.
Useful compile-time options include:

- **MP:dsm={on, of f}** (default on)
  All the data-distribution and scheduling features described in this chapter are enabled by default under `-mp` compilation. To disable all the DSM-specific directives (for example, distribution and affinity scheduling), compile with `-MP:dsm=of f`.

- **MP:clone={on, of f}** (default on)
  The compiler automatically clones procedures that are called with reshaped arrays as parameters for the incoming distribution. However, if you have explicitly specified the distribution on all relevant formal parameters, then you can disable auto-cloning with `-MP:clone=of f`. The consistency checking of the distribution between actual and formal parameters is *not* affected by this flag, and is always enabled.

- **MP:check reshape={on, of f}** (default off)
  Enables generation of the run-time consistency checks across procedure boundaries when passing reshaped arrays (or portions thereof) as parameters.

**Examples**

The examples in this section include the following:

- "Distributing Columns of a Matrix” on page143
- "Using Data Distribution and Data Affinity Scheduling” on page 144
- "Parameter Passing” on page145
- "Redistributed Arrays” on page146
- "Irregular Distributions and Thread Affinity” on page 147
**Distributing Columns of a Matrix**

The example below distributes sequentially the columns of a matrix. Such a distribution places data effectively only if the size of an individual column exceeds that of a page.

```fortran
real*8 A(n, n)
C Distribute columns in cyclic fashion
c$distribute A (*, CYCLIC(1))

C Perform Gaussian elimination across columns
C The affinity clause distributes the loop iterations based
C on the column distribution of A
C
C$doacross affinity(j) = data(A(i,j))
do i = 1, n
   do j = i+1, n
      ... reduce column j by column i ...
   enddo
enddo
```

If the columns are smaller than a page, then it may be beneficial to reshape the array. This is easily specified by changing the keyword from `distribute` to `distribute_reshape`.

In addition to overcoming size constraints as shown above, the `distribute_reshape` directive is useful when the desired distribution is contrary to the layout of the array. For instance, suppose you want to distribute the rows of a two-dimensional matrix. In the following example, the `distribute_reshape` directive overcomes the storage layout constraints to provide the desired distribution.

```fortran
real*8 A(n, n)
C Distribute rows in block fashion
c$distribute_reshape A (BLOCK, *)
real*8 sum(n)
c$distribute sum(BLOCK)

C Perform sum-reduction on the elements of each row
C$doacross local(j) affinity(i) = data(A(i,j))
do i = 1, n
   do j = 1, n
      sum(i) = sum(i) + A(i,j)
   enddo
enddo
```

Using Data Distribution and Data Affinity Scheduling

This example demonstrates regular data distribution and data affinity. This example, run on a 4-processor Origin2000 server, uses simple block scheduling. Processor 0 will calculate the results of the first 25,000 elements of A, processor 1 will calculate the second 25,000 elements of A, and so on. Arrays B and C are initialized using one processor; hence all of the memory pages are touched by the master processor (processor 0) and are placed in processor 0’s local memory.

Using data distribution changes the placement of memory pages for arrays A, B, and C to match the data reference pattern. Thus, the code runs 33% faster on a 4-processor Origin2000 (than it would run using SMP directives without data distribution).

Without Data Distribution

```fortran
real*8 a(1000000), b(1000000)
real*8 c(1000000)
integer i

c$par parallel shared(a, b, c) local(i)
c$par pdo
   do i = 1, 100000
      a(i) = b(i) + c(i)
   enddo
c$par end parallel
```

With Data Distribution

```fortran
real*8 a(1000000), b(1000000)
real*8 c(1000000)
integer i
c$distribute a(block), b(block), c(block)

c$par parallel shared(a, b, c) local(i)
c$par pdo affinity( i ) = data( a(i) )
do i = 1, 100000
   a(i) = b(i) + c(i)
enddo
c$par end parallel
```
Parameter Passing

A distributed array can be passed as a parameter to a subroutine that has a matching declaration on the formal parameter:

```fortran
real*8 A (m, n)
c$distribute_reshape A (block, *)
call foo (A, m, n)
end
```

```fortran
subroutine foo (A, p, q)
real*8 A (p, q)
c$distribute_reshape A (block, *)
c$doacross affinity (i) = data (A(i, j))
do i = 1, P
  enddo
end
```

Since the array is reshaped, it is required that the `distribute_reshape` directive in the caller and the callee match exactly. Furthermore, all calls to subroutine `foo()` must pass in an array with the exact same distribution.

If the array was only distributed (that is, not reshaped) in the above example, then the subroutine `foo()` can be called from different places with different incoming distributions. In this case, you can omit the distribution directive on the formal parameter, thereby ensuring that any data affinity within the loop is based on the distribution (at runtime) of the incoming actual parameter.

```fortran
real*8 A(m, n), B (p, q)
real*8 A (block, *)
real*8 B (cyclic(1), *)
call foo (A, m, n)
call foo (B, p, q)
```

```fortran
subroutine foo (X, s, t)
real*8 X (s, t)
c$doacross affinity (i) = data (X(i+2, j))
do i = * * *
  enddo
```
Redistributed Arrays

This example shows how an array is redistributed at runtime:

```fortran
subroutine bar (X, n)
real*8 X(n, n)
...
c$redistribute X (*, cyclic(<expr>))
...
end
```

subroutine foo
real*8 LocalArray (1000, 1000)
c$distribute LocalArray (*, BLOCK)
C the call to bar() may redistribute LocalArray
C dynamic LocalArray
...
call bar (LocalArray, 100)
C The distribution for the following doacross
C is not known statically
C$doacross affinity (i) = data (A(i, j))
end
```

The next example illustrates a situation where the c$dynamic directive can be optimized away. The main routine contains a local array A that is both distributed and dynamically redistributed. This array is passed as a parameter to foo() before being redistributed, and to bar() after being (possibly) redistributed. The incoming distribution for foo() is statically known; you can specify a c$distribute directive on the formal parameter, thereby obtaining more efficient static scheduling for the affinity doacross. The subroutine bar(), however, can be called with multiple distributions, requiring run-time scheduling of the affinity doacross.

```fortran
program main
C$distribute A (block, *)
c$dynamic A
call foo (A)
if (x.ne.17) then
  c$redistribute A (cyclic(x), *)
endif
call bar (A)
end
```
subroutine foo (A)
C Incoming distribution is known to the user
c$distribute A(block, *)
c$doacross affinity (i) = data (A(i, j))
...
end

subroutine bar (A)
C Incoming distribution is not known statically
c$dynamic A
c$doacross affinity (i) = data (A(i, j))
...
end

Irregular Distributions and Thread Affinity

The example below consists of a large array that is conceptually partitioned into unequal portions, one for each processor. This array is indexed through an index array idx, which stores the starting index value and the size of each processor’s portion.

real*8 A(N)
C idx ---> index array containing start index into A (idx(p, 0))
C and size (idx(p, 1)) for each processor
real*4 idx (P, 2)
c$page_place (A(idx(0, 0)), idx(0, 1)*8, 0)
c$page_place (A(idx(1, 0)), idx(1, 1)*8, 1)
c$page_place (A(idx(2, 0)), idx(2, 1)*8, 2)
...
c$doacross affinity (i) = thread(i)
do i = 0, P-1
   ... process elements on processor i • • •
   ... A(idx(i, 0)) to A(idx(i,0)+idx(i,1)) • • •
enddo