Appendix I

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Figure I.8 Data miss rates can vary in nonobvious ways as the processor count is increased from 1 to 16. The miss rates include both coherence and capacity miss rates. The compulsory misses in these benchmarks are all very small and are included in the capacity misses. Most of the misses in these applications are generated by accesses to data that are potentially shared, although in the applications with larger miss rates (FFT and Ocean), it is the capacity misses rather than the coherence misses that comprise the majority of the miss rate. Data are potentially shared if they are allocated in a portion of the address space used for shared data. In all except Ocean, the potentially shared data are heavily shared, while in Ocean only the boundaries of the subgrids are actually shared, although the entire grid is treated as a potentially shared data object. Of course, since the boundaries change as we increase the processor count (for a fixed-size problem), different amounts of the grid become shared. The anomalous increase in capacity miss rate for Ocean in moving from 1 to 2 processors arises because of conflict misses in accessing the subgrids. In all cases except Ocean, the fraction of the cache misses caused by coherence transactions rises when a fixed-size problem is run on an increasing number of processors. In Ocean, the coherence misses initially fall as we add processors due to a large number of misses that are write ownership misses to data that are potentially, but not actually, shared. As the subgrids begin to fit in the aggregate cache (around 16 processors), this effect lessens. The single-processor numbers include write upgrade misses, which occur in this protocol even if the data are not actually shared, since they are in the shared state. For all these runs, the cache size is 64 KB, two-way set associative, with 32-byte blocks. Notice that the scale on the y-axis for each benchmark is different, so that the behavior of the individual benchmarks can be seen clearly.
Figure I.9 The miss rate usually drops as the cache size is increased, although coherence misses dampen the effect. The block size is 32 bytes and the cache is two-way set associative. The processor count is fixed at 16 processors. Observe that the scale for each graph is different.
Figure I.10 The data miss rate drops as the cache block size is increased. All these results are for a 16-processor run with a 64 KB cache and two-way set associativity. Once again we use different scales for each benchmark.
Figure 1.11 Bus traffic for data misses climbs steadily as the block size in the data cache is increased. The factor of 3 increase in traffic for Ocean is the best argument against larger block sizes. Remember that our protocol treats ownership or upgrade misses the same as other misses, slightly increasing the penalty for large cache blocks; in both Ocean and FFT, this simplification accounts for less than 10% of the traffic.
Figure I.12 The data miss rate is often steady as processors are added for these benchmarks. Because of its grid structure, Ocean has an initially decreasing miss rate, which rises when there are 64 processors. For Ocean, the local miss rate drops from 5% at 8 processors to 2% at 32, before rising to 4% at 64. The remote miss rate in Ocean, driven primarily by communication, rises monotonically from 1% to 2.5%. Note that, to show the detailed behavior of each benchmark, different scales are used on the $y$-axis. The cache for all these runs is 128 KB, two-way set associative, with 64-byte blocks. Remote misses include any misses that require communication with another node, whether to fetch the data or to deliver an invalidate. In particular, in this figure and other data in this section, the measurement of remote misses includes write upgrade misses where the data are up to date in the local memory but cached elsewhere and, therefore, require invalidations to be sent. Such invalidations do indeed generate remote traffic, but may or may not delay the write, depending on the consistency model (see Section 5.6).
Figure I.13 **Miss rates decrease as cache sizes grow.** Steady decreases are seen in the local miss rate, while the remote miss rate declines to varying degrees, depending on whether the remote miss rate had a large capacity component or was driven primarily by communication misses. In all cases, the decrease in the local miss rate is larger than the decrease in the remote miss rate. The plateau in the miss rate of FFT, which we mentioned in the last section, ends once the cache exceeds 128 KB. These runs were done with 64 processors and 64-byte cache blocks.
Figure I.14 Data miss rate versus block size assuming a 128 KB cache and 64 processors in total. Although difficult to see, the coherence miss rate in Barnes actually rises for the largest block size, just as in the last section.
Figure I.15 The number of bytes per data reference climbs steadily as block size is increased. These data can be used to determine the bandwidth required per node both internally and globally. The data assume a 128 KB cache for each of 64 processors.
Figure I.17 The effective latency of memory references in a DSM multiprocessor depends both on the relative frequency of cache misses and on the location of the memory where the accesses are served. These plots show the memory access cost (a metric called average memory access time in Chapter 2) for each of the benchmarks for 8, 16, 32, and 64 processors, assuming a 512 KB data cache that is two-way set associative with 64-byte blocks. The average memory access cost is composed of four different types of accesses, with the cost of each type given in Figure I.16. For the Barnes and LU benchmarks, the low miss rates lead to low overall access times. In FFT, the higher access cost is determined by a higher local miss rate (1–4%) and a significant three-hop miss rate (1%). The improvement in FFT comes from the reduction in local miss rate from 4% to 1%, as the aggregate cache increases. Ocean shows the biggest change in the cost of memory accesses, and the highest overall cost at 64 processors. The high cost is driven primarily by a high local miss rate (average 1.6%). The memory access cost drops from 8 to 16 processors as the grids more easily fit in the individual caches. At 64 processors, the dataset size is too small to map properly and both local misses and coherence misses rise, as we saw in Figure I.12.
Figure I.18 The BG/L processing node. The unfilled boxes are the PowerPC processors with added floating-point units. The solid gray boxes are network interfaces, and the shaded lighter gray boxes are part of the memory system, which is supplemented by DDR RAMS.
Figure I.19 The 64K-processor Blue Gene/L system.
Figure I.21 The space of large-scale multiprocessors and the relation of different classes.