Locality-Based Optimizations in the Chapel Compiler

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ABSTRACT

In the recent releases, we have added two locality-based optimizations to the Chapel compiler. These optimizations enable the compiler to statically determine the locality of array accesses and aggregate fine-grained copy operations. In this talk, we summarize how they are implemented, their impact on various programming idioms, associated performance improvements and pertinent future directions.

KEYWORDS

parallel programming, compiler optimizations, productivity

1 INTRODUCTION

² Chapel is a parallel programming language that supports parti-

- tioned global address space (PGAS) memory model. The PGAS
- ⁴ model allows programmers to use a single address space, which
- ⁵ improves productivity by making all available system memory ac-
- 6 cessible by every locale without explicit communication. Chapel
- $_{7}$ $\,$ combines the PGAS memory model with other high-level concepts $\,$
- ⁸ such as distributed arrays and data parallel distributed loops to cre-
- ⁹ ate an expressive programming language. However, this paradigm
- is prone to writing code with poor performance and scalability
 because of implicit communication.
- because of implicit communication.
 On the other hand, common programming idioms represented
- by Chapel's first-class, high-level language concepts also enable the
 compiler to make automatic optimizations that would be impossible
- ¹⁵ in low-level paradigms such as message passing. This talk focuses
- on two such optimizations that significantly mitigate common per-
- formance overheads with no programmer effort.

Both optimizations that this talk covers are in the Chapel 1.24 release, and can be readily used by the programmers.

20 2 AUTOMATIC LOCAL ACCESS

- 21 Accesses to Chapel arrays are implemented with a method named
- $_{\rm 22}$ $\,$ this on the array type that is automatically called by the compiler.
- $_{\rm 23}~$ A simplified implementation of this for a distributed array type is

```
<sup>24</sup> shown in Listing 1.
```

```
25 1 proc this(idx) {
26 2 if isLocalIndex(idx) then
```

27 3 return localAccess(idx);

```
28 4 else then
```

```
29 5 return nonLocalAccess(idx);
```

```
30 6 }
```

Listing 1: A Simplified Implementation of Distributed Array Access

- Note that, in line 2, the implementation checks whether idx is
- $_{\rm 32}$ $\,$ local, because if that is the case, the local data can be accessed in a
- ³³ much faster manner. However, this check itself has some small but

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Figure 1: STREAM Bandwidth

noticeable overhead. Consider a STREAM-Triad implementation in Chapel that uses indexed access into distributed arrays, as shown in Listing 2.

```
use BlockDist;
37 1
38
          = newBlockDom(1..n);
   var D
39 3
       A: [D] int,
   var
40 4
        B: [D] int,
41 5
        C: [D] int;
42 6
43
   forall i in A.domain do
44 8
      A[i] = B[i] + alpha * C[i];
45 9
```

Listing 2: STREAM kernel with indexed access

In this snippet, the three distributed arrays are accessed by index in the forall loop body, and they would normally incur the locality checks as discussed above. However, these checks can be avoided because:

- The forall loop will distribute the work in the same way the loop domain (A. domain) is distributed
- All arrays are distributed the same way as the loop domain is distributed
- All three distributed arrays are accessed at the ith index, which is the loop index

Starting in Chapel 1.23, the compiler is able to make this analysis in cases like the above. Moreover, it can also transform the code to do a once-per-loop dynamic check to use localAccess automatically if only a subset of the requirements can be proven at compile time.

Figure 1 shows how this optimization improves STREAM performance, where the kernel is implemented similarly to the one in Listing 2, With this optimization, indexed STREAM performs about ⁶⁴ twice as fast, reaching the limits of the system. This performance

is virtually identical to other idioms that do not use indexed access

66 into distributed arrays.

67 3 AUTOMATIC COPY AGGREGATION

Another common overhead in languages with a PGAS memory 68 model occurs due to fine-grained communication. In some cases 69 where the fine-grained access is predictable, caching and/or prefetch-70 ing the remote data can help mitigate some of these overheads. 71 However, especially in cases where the remote data is accessed 72 randomly, such approaches are generally not very impactful. A 73 solution for these scenarios is aggregating the communication and 74 transferring data in bulk with fewer messages. 75 Listing 3 shows a simplified version of the index_gather kernel 76 from the bale effort [1]. 77 var cycArr = newCyclicArr(...); 78 1 var blockArr = newBlockArr(...); 79 80

s1 4
fillRandom(blockArr);
s2 5
s3 6
var tmp: [blockArr.domain] int;
s4 7
s5 s
forall i in blockArr.domain do

s6 9 tmp[i] = cycArr[blockArr[i]];

Listing 3: Simplified Sketch of the index_gather Kernel

The forall loop iterates over a block-distributed domain, while copying data from a cyclic-distributed array into a block-distributed one. This element-wise, random-access copy operation causes finegrained communication. However, this operation can be done in an aggregated fashion because:

temp[i] (and blockArr[i]) are local accesses because the
 forall is over the same domain as theirs. Furthermore, this
 will be recognized as such by the automatic local access
 optimization that was discussed above,

 Individual copy operations that will execute at each iteration of the, loop can be reordered without impacting the application behavior.

Starting in Chapel 1.24, the compiler is able to perform this analysis and use aggregation in these scenarios. This optimization 100 relies heavily on other compiler capabilities and module-level opti-101 mizations developed before. The aggregation is facilitated through 102 module-level aggregation objects that were designed for this study 103 and have been used in Arkouda [2] in similar scenarios. Therefore, 104 the required AST transformation for this optimization is relatively 105 small. On the other hand, the Chapel compiler already had some 106 analysis to check for safety of unordered execution in similar cases 107 that enabled unordered forall optimization. Automatic aggregation 108 relies on that analysis to make aggregation decisions. 109

Figure 2 shows that without any optimization, this benchmark does not scale (light blue). Unordered forall optimization, firing automatically with no user effort, improves performance by enabling out-of-order communication (medium blue). Finally, manual aggregation (dark blue) and automatic aggregation (solid green)



Figure 2: Bale index_gather

perform very similarly and much better than the other versions, where the latter does not require any user effort, at all.

117 4 CONCLUSION

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This work covers automatic local access and automatic aggregation optimizations that were implemented in the Chapel compiler in the recent releases. The talk gives a brief overview on how they are implemented and how Chapel's high-level features enable them. It demonstrates different idioms where such optimizations do and do not fire. Finally, it concludes by discussing potential future improvements.

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REFERENCES

[1] https://github.com/jdevinney/bale

[2] https://github.com/mhmerrill/arkouda