Speculative Runtime Parallelization of Loop Nests: Towards Greater Scope and Efficiency

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APOLLO

HIPS

What is this talk about ?

- A dynamic and speculative optimizer
- Optimization of loop nests that cannot be handled statically
- Dynamic application of polyhedral model
- Non linear extensions to polyhedral model



- Static approaches are limited due to intractable control and memory instructions
 - Indirect memory accesses (A[B[i]])
 - Pointers (ptr = ptr-> next)
 - While loops (While (ptr!=NULL))
- Dynamic approaches can overcome these limitations by using run-time information



- Dynamic approaches require to be speculative to enlarge their scope (Thread level speculation)
- Traditional TLS system relies on centralized verification which is not scalable due to the high memory traffic
- Apollo uses a prediction model based on the polytope model and on linear approximations introduced in this talk



- Predict an optimization
- Verify that the prediction holds
- A fail-safe system to recover from a mis-prediction







while (cond)				
{				
a[i] =;				
••				
••				
••				
x = a[j];				
}				

Time slot 1			
P0	P1	P2	P3
a[1] =	a[2] =	a[3] =	a[4] =
x = a[100] try_commit()	x = a[101] try_commit()	x = a[102] try_commit()	x = a[103] try_commit()



Sequential execution order

 $\begin{array}{l} a[1] = \dots \\ x = a[100] \\ a[201] = \dots \\ x = a[101] \\ a[100] = \dots \\ x = a[102] \\ a[250] = \dots \\ x = a[103] \end{array}$





- What is missing ?
 - A dependence prediction model



- What is missing ?
 - A dependence prediction model
- How is this affected ?
 - Missed parallelization opportunities
 - Higher mis-prediction rates
 - Huge centralization overhead



Apollo



Figure : Mission control: We have lift off



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APOLLO : Automatic speculative POLyhedral Loop Optimizer

Features
• Dynamic
 Polytope model
 Speculative with weakly centralized verification
• Handle all loop types
• Extends the applicability of the polytope model to non linear memory
accesses



Polytope model

- A mathematical way to model the loop nests in a program
- Each instance of a statement is represented by a point in the lattice of the polyhedron
- Widely used for static program optimization (Pluto)



Original user code

- *i* : source iterator
- i' : target iterator

Dependence constraints

$$\begin{array}{l} i >= 2 \\ i <= 9 \\ i' >= 2 \\ i' <= 9 \end{array} \right\} \text{Domain constraints} (1)$$

$$i' = i + 2$$
 Access constraints (2)

 $i'>=i+1
ight\}$ Order constraints (3)



In this code we can extract the linear functions statically

```
for(i = 2; i < 10; i++){
    a[i] = a[i - 2] + 1;
}</pre>
```

This code requires dynamic analysis to extract linear functions

```
while(ptr1 && ptr2){
    ptr1->val = ptr2->val + 1;
    ptr1 = ptr1->next;
    ptr2 = ptr2->next;
}
```



Speculate the linear functions for

- Dynamic memory accesses
- Dynamic loop bounds
- Scalar variables which carry cross iteration dependencies



- Profile the code by sampling
- Interpolate memory addresses and scalar values
- Compute the data dependencies and build the prediction model
- Compute optimizing valid transformation
- Speculatively execute the optimized code
- Verify the speculation while the optimized code is running



Challenges

- How to instrument?
- How to build the prediction model?
- How to compute the optimizing transformation?
- How to generate the optimized code?
- How to verify the speculation?



Apollo consists of two core components

- Static module
- Runtime module



Static Module

- A set of dedicated LLVM compiler passes
- Statically analyze memory instructions which can be disambiguated at compile time
- Transforms any kind of target loops into for loops
- Generates an instrumented version to track memory accesses
- Creates optimized code skeletons



Code skeleton

- General frameworks representing a class of transformations.
- Skeletons are parametrized. Instantiating different parameters results in different transformations.
- Instrumentation skeleton is used to track memory accesses
- Optimized skeletons are used for parallelization and other code optimizations (data locality ...)





Figure : Optimized skeleton



Runtime Module

- Runs the instrumentation skeleton for a small outermost loop slice
- Builds a linear prediction model for the loop bounds and memory accesses
- Computes the dependencies between the memory accesses
- Computes the transformation
- Selects and instantiates the appropriate code skeleton
- Monitor the execution to verify the correctness of the linear functions and thereby the transformation



Chunking

consider the following simple code

```
for(i = 0; i < 1000; i++){
    a[i] = b[i + 2] + 1;
}</pre>
```



Chunking

consider the following simple code

```
for(i = 0; i < 1000; i++){
    a[i] = b[i + 2] + 1;
}</pre>
```







Apollo global view





Memory backup

- The execution is speculative
- A mis-speculation can trigger rollback
- In order to prevent memory corruption, all the predicted memory write regions are backed up
- Thanks to the linear prediction model; the exact write regions can be identified



Verification Module

- The validity of the polytope model is proven by construction
- The transformation is valid as long as the predicted linear functions are valid
- Uses the linear access functions generated during the instrumentation phase



Verification Module

- Runtime verification system ensures that the memory accesses follows the predicted linear functions
- Polyhedral transformations will also affect the execution order of iterations inside each thread. Hence each iteration must be verified
- If verification fails a rollback is triggered



Apollo {Runtime Verification}



Figure : Optimized skeleton



Non affine memory accesses

- The polyhedral model cannot handle non affine accesses (even with dynamic analysis)
- In the presence of non affine accesses, the computed dependencies may be inaccurate
- The validity of the transformation cannot be guaranteed



Why should we be concerned about non affine accesses

- Most of the dynamic programs exhibit non affine behavior
- Most of the indirect accesses and pointers to dynamic memory are typically non linear
- A dependence prediction model is vital for the efficiency of TLS systems



Challenges

- Polytope model as such is not compatible
- Non linear accesses will require a centralization system
- Live backup is required
- There is no linear function available for validation
- The overhead cost should be acceptable



Solution

- Relax the polytope model by adding some non affine accesses
- Account for the relaxation by adding additional verification



During Instrumentation

- Identify potential non linear accesses
- Compute regression lines modeling each non linear access
- Refine the regression line by removing outliers



During Instrumentation

- Compute the regression correlation coefficient
- Correlation coefficient measures the quality of regression line
- Characterize each memory access as
 - Affine
 - Non affine
 - Nearly affine



Apollo {Access types}



Building the dependence polytope

- Nearly affine : The correlation coefficient is greater than 0.9
- Non affine : The correlation coefficient is lower than 0.9



Building the dependence polytope : Nearly affine

- The memory accesses are well characterized
- \bullet Approximate the regression hyperplane from ${\mathbb R}$ domain to ${\mathbb Z}$ domain
- Compute two bounding hyperplanes *close* to the regression hyperplane (tubes), one *lower* and the other *higher*
- Encode these bounding hyperplanes to the polytope model





Figure : Bounding hyperplanes



Building the dependence polytope : Non affine

- The memory accesses are not well characterized
- Adding them to the dependence polytope will have adverse effects
- Hence do not encode them to the dependence polytope



Pre-execution validation

- Detect any possible violation as early as possible
- For the instrumented memory accesses, verify that non linear memory accesses do not invalidate transformation
- This can be done by checking for intersection of memory access between affine and non affine accesses



Building safe point

- Backing up while running can hurt performance a lot
- For non affine and nearly affine accesses there is no way to exactly
 predict the memory addresses that will be written
- For non affine accesses, compute a range information, and backup
- For nearly affine accesses, compute the area inside the bounding hyperplanes and backup this region



Execution

- Based on the transformation suggested by the scheduler, select and initiate a skeleton
- Pluto is used dynamically for scheduling



Verification

- Affine access: Verify that accesses follow the affine function
- Nearly affine access: If the instance falls inside the tube, the access is valid. If not treat that particular instance as a non affine access
- Non affine access: For the non predicted ranges, perform live backup.





Figure : Optimized skeleton with non linear support





Figure : Speedup : The higher the better



- A dependence model is a must for the TLS system
- Thanks to dynamic and speculative environment, a well designed extension to the polytope model can amend the model to consider non linear accesses
- Can be used in any general dynamic speculative system



Questions?

