What it Takes to Assign Blame

Nick Rutar Jeffrey K. Hollingsworth

University of Maryland



Parallel Framework Mapping

- Traditional profiling represented as
 - Functions, Basic Blocks, Statement
- Frameworks have intuitive abstractions
 - Direct ties with mathematical terms
 - PETSc, Cactus, POOMA, GrACE
- Map profiling information to variables
 - Maps to abstractions in case of frameworks
 - Also can be used for standard programs
 - Map Structs, Classes, Arrays, Scalars



| | Example PETSC Program* | | | | |
|---|---|--|--|--|--|
| 50% cache misses — | <pre>int main(int argc,char **args) { Vec x, /* approx solution */ b, /* right hand side */</pre> | | | | |
| 30% MPI operations | u; /* exact solution*/ Mat A; /* linear system matrix */ KSP ksp; /* linear solver context */ | | | | |
| 40% run — time | PC pc; /* preconditioner context */ VecCreate(PETSC_COMM_WORLD,&x); VecDuplicate(x,&b); | | | | |
| | VecDuplicate(x,&u); MatCreate(PETSC_COMM_WORLD,&A); MatAssemblyBegin(A,MAT_FINAL_ASSEMBLY); | | | | |
| MatAssemblyEnd(A,MAT_FINAL_ASSEMBLY); /* Set exact solution */ VecSet(u,one); MatMult(A,u,b); /* Create linear solver context */ KSPCreate(PETSC_COMM_WORLD,&ksp); | | | | | |
| | | | | | |
| 3 | ierr = KSPSolve(ksp,b,x); } | | | | |



Variable "Blame"

- Record writes in a function
- Build association tree of writes from ground up
- Use transfer function to filter information up
 - Up the call stack
 - Aggregate over distributed nodes
- Eventually reach high level abstractions
 - Example: Matrix abstraction
 - Allocated storage for actual data
 - Sparse or Dense
 - Storage for bookkeeping
- Augments traditional profiling approaches





Preliminary Experimental Results

- Chose programs with similar properties to those found in parallel frameworks
- Blame metric is number of cycles
- For each sampling point (instance)
 - Instance gets blamed for set number of cycles
 - Variable that instance maps up to gets blame



FFP_SPARSE

- C++ program that solves Poisson's Equation
 - Approximately 6,700 lines of code & 63 Functions
- Non-parallel program
- Uses Sparse Matrices
 - No specific data structure for representation
 - Composite of primitive pointers declared in 'main'
- Recorded 101 samples from program run



FFP_SPARSE Results

| Name | Туре | Description | Direct | Blame (%) |
|------------------|----------|------------------------------|--------|-------------|
| node_u | double * | Solution vector | 0 | 35 (34.7) |
| a | double * | Coefficient matrix | 0 | 24.5 (24.3) |
| ia | int * | Non-zero row indices of a | 1 | 5 (5.0) |
| ja | int * | Non-zero column indices of a | 1 | 5 (5.0) |
| element_neighbor | int * | Estimate of non-zeroes | 0 | 10 (9.9) |
| node_boundary | bool * | Bool vector for boundary | 0 | 9 (8.9) |
| f | double * | Right hand side of vector | 0 | 3.5 (3.5) |
| | | | | |
| Other | - | | 99 | 9 (8.9) |
| Total | - | | 101 | 101 (100) |

Dyn inst

HPL

• C program that solves a linear system

- Utilizes MPI and BLAS
- Has wrappers for functions from both libraries
- Operations done on dense matrices
- Approximately 18,000 lines of code
- 149 source files

• 32 Red Hat nodes connected via Myrinet

- OpenMPI 1.2.8
- Range of 149-159 samples over the nodes



| | MPL RE | Sulls | Blame over | 32 Nodes | | |
|------------------------------|---------------|--------------|----------------|---------------|--|--|
| | Name | Туре | Mean (Total %) | Node St. Dev. | | |
| | All Instances | - | 154.7(100) | 2.7 | | |
| | ≻ main | | | | | |
| | grid | HPL_T_grid | 2.2(1.4) | 0.4 | | |
| → main→HPL_pdtest | | | | | | |
| | mat | HPL_T_pmat | 139.3(90.0) | 2.8 | | |
| Plama | Anorm1 | double | 1.4(0.9) | 0.8 | | |
| Points | AnormI | double | 1.1(0.7) | 1.0 | | |
| | XnormI | double | 0.5(0.3) | 0.7 | | |
| | Xnorm1 | double | 0.2(0.1) | 0.4 | | |
| → main→HPL_pdtest→HPL_pdgesv | | | | | | |
| | А | HPL_T_pmat * | 136.6(88.3) | 2.9 | | |
| main | | | | | | |
| | PANEL→L2 | HPL_T_pmat | 112.8(72.9) | 8.5 | | |
| | PANEL→A | double | 12.8(8.3) | 3.8 | | |
| 10 | PANEL→U | double | 10.2(6.6) | 5.2 | | |
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Implementation Details

- Mixture of Static and Runtime Tools
- Static Analysis
 - LLVM
 - Boost
- Runtime Analysis
 - Dyninst API
 - Symtab API
 - Stackwalker API
 - PAPI



LLVM (Low Level Virtual Machine)

• What is it?

- Compiler Infrastructure
- Provides Intermediate Representation
 - Each instruction in SSA form

Why we use it?

- Need intermediate representation for static analysis
- SSA form useful for creating dependency relationships
- Intuitive API for accessing
 - Def-use chains
 - Dominator & CFG information
 - Language Independent representation of complex types
- Integration with GCC
- Multiple Language support
 - C, C++, Fortran

• Limitations

- Ilvm-gcc versus gcc



Boost

• What is it? - Widely used portable C++ Libraries • Why we use it? - Implicit/Explicit data flow relationships Can create very large graphs - Boost provides graph libraries Efficient representation of nodes/edges - Descriptors assigned to both DFS, BFS, Uniform Cost Search • Dijkstra's Shortest Path, Kruskal's MST, ... Limitations

- Trade efficiency for requiring one more library

StackWalker API

- What is it?
 - API for runtime traversing of stack
- Why we use it?
 - Instance Generation
 - Used in combination with PAPI
 - · Each sample point we need full path information
 - Use full context given from PAPI
 - Walk up stack until we reach the top
 - Mem-Container Information
 - Used in combination with Dyninst
 - Wrapper functions mean we need full path
 - Every allocation we get full allocation path
- Limitations
 - Frame pointer removal decreases accuracy



DyninstAPI

• What is it?

- Dynamic instrumentation tool
- Why we use it?
 - Need to instrument memory allocation sites
 - Integrated with StackWalkerAPI

Limitations

- Instrumentation overhead



SymtabAPI

• What is it?

- API for accessing symbol information
- Why we use it?
 - General Module/Function Information
 - Line Number Mappings
 - Runtime information mapped back to source
 - Use line number mappings for this

Limitations

- Debugging Information needed



PAPI

• What is it?

- API that provides interface to hardware counters

• Why we use it?

- Instance (Sample Point) Generation
 - PAPI provides sampling interface
 - User chooses metric to trigger sample
 - Metrics can be any measurable event on system
 - PAPI hardware counters
- Limitations
 - Special kernel patch required on certain systems



Advantage of Using Tools

| Application/API | LOC (w/comments) |
|---------------------|------------------|
| Blame | 6K (8K) |
| | |
| Dyninst API 6.0 | 292K (360K) |
| Symtab API 6.0 | 51K (65K) |
| Stackwalker API 6.0 | 52K (66K) |
| LLVM 2.3 | 298K (375K) |
| PAPI 3.6 | 278K (320K) |
| Boost (Graph) 1.36 | 29K (33K) |



Conclusion

- Variable "blame" mapping
 - Switch analysis from delimited regions to variables
 - Alternative to standard profiling techniques

• Lessons Learned

- Standards are a good things
 - PAPI gives ucontext
 - Stackwalker uses information for context
- Best not to reinvent the wheel ... BUT
- Tool interoperability can be a problem
 - Compiler, OS compatibilities
 - Runtime tool interoperability
 - Target application/end-user requirements
- Questions?

