

Performance Debugging: Methods and Tools

David Skinner <u>deskinner@lbl.gov</u>







Lawrence Berkeley National Laboratory

Performance Debugging: Methods and Tools

• Principles

- Topics in performance scalability
- Examples of areas where tools can help

Practice

- Where to find tools
- Specifics to NERSC and Hopper

Scope & Audience

- Budding simulation scientist app dev
- Compiler/middleware dev, YMMV



One Slide about NERSC

 Serving all of DOE Office of Science

- Lots of users
 - ~4K active
 - ~500 logged in



Big Picture of Performance and Scalability





App Performance: Dimensions

A r

chitect

u r

e

- Code
- Input deck
- Computer
- Concurrency
- Workload
- Person



Performance is Relative

To your goals

- Time to solution, $T_q + T_{wall} \dots$
- Your research agenda
- Efficient use of allocation

To the

- application code
- input deck
- machine type/state

Suggestion: Focus on specific use cases as opposed to making *everything* perform well. Bottlenecks can shift.







Performance, more than a single number

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•Plan where to put effort

•Optimization in one area can de-optimize another

 Timings come from timers and also from your calendar, time spent coding

 Sometimes a slower algorithm is simpler to verify correctness

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Different Facets of Performance

Serial

- Leverage ILP on the processor
- Feed the pipelines
- Exploit data locality
- Reuse data in cache
- Parallel
 - Expose concurrency
 - Minimizing latency effects
 - Maximizing work vs. communication





Performance is Hierarchical





...on to specifics about HPC tools

Mostly at NERSC but fairly general





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Tools are Hierarchical



HPC Perf Tool Mechanisms

Sampling

- Regularly interrupt the program and record where it is
- Build up a statistical profile

Tracing / Instrumenting

Insert hooks into program to record and time events, document everything

Use Hardware Event Counters

- Special registers count events on processor
- E.g. floating point instructions
- Many possible events
- Only a few (~4 counters)



Typical Tool Use Requirements

- (Sometimes) Modify your code with macros, API calls, timers
- Compile your code
- Transform your binary for profiling/ tracing with a tool
- Run the transformed binary
 - A data file is produced
- Interpret the results with a tool





Performance Tools @ NERSC

- Vendor Tools:
 - CrayPat

• Community Tools :

- TAU (U. Oregon via ACTS)
- PAPI (Performance Application
- Programming Interface)
- gprof
- IPM: Integrated Performance Monitoring





What HPC tools can tell us?

- CPU and memory usage
 - FLOP rate
 - Memory high water mark
- OpenMP
 - OMP overhead
 - OMP scalability (finding right # threads)
- MPI
 - % wall time in communication
 - Detecting load imbalance
 - Analyzing message sizes



Using the right tool

Tools can add overhead to code execution

What level can you tolerate?

Tools can add overhead to scientists

What level can you tolerate?

Scenarios:

- Debugging a code that is "slow"
- **Detailed performance debugging** \bullet
- **Performance monitoring in production**





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Introduction to CrayPat

- Suite of tools to provide a wide range of performance-related information
- Can be used for both sampling and tracing user codes
 - with or without hardware or network performance counters
 - Built on PAPI
- Supports Fortran, C, C++, UPC, MPI, Coarray Fortran, OpenMP, Pthreads, SHMEM
- Man pages
 - intro_craypat(1), intro_app2(1), intro_papi(1)



Using CrayPat @ Hopper

1. Access the tools

- module load perftools

2. Build your application; keep .o files

- make clean
- make

3. Instrument application

- pat_build ... a.out
- Result is a new file, a.out+pat

4. Run instrumented application to get top time consuming routines

- aprun ... a.out+pat
- Result is a new file XXXXX.xf (or a directory containing .xf files)

5. Run pat_report on that new file; view results

- pat_report XXXXX.xf > my_profile
- vi my_profile
- Result is also a new file: XXXXX.ap2

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Guidelines for Optimization

Derived metric	Optimization needed when*	PAT_RT_HWP C	
Computational intensity	< 0.5 ops/ref	0, 1	
L1 cache hit ratio	< 90%	0, 1, 2	
L1 cache utilization (misses)	< 1 avg hit	0, 1, 2	
L1+L2 cache hit ratio	< 92%	2	
L1+L2 cache utilization (misses)	< 1 avg hit	2	
TLB utilization	< 0.9 avg use	1	
(FP Multiply / FP Ops) or (FP Add / FP Ops)	< 25%	5	
Vectorization	< 1.5 for dp; 3 for sp	12 (13, 14)	

* Suggested by Cray



Perf Debug and Production Tools

- Integrated Performance Monitoring
- MPI profiling, hardware counter metrics, POSIX IO profiling
- IPM requires no code modification & no instrumented binary
 - Only a "module load ipm" before running your program on systems that support dynamic libraries
 - Else link with the IPM library
- IPM uses hooks already in the MPI library to intercept your MPI calls and wrap them with timers and counters



IPM: Let's See

Do "module load ipm", link with \$IPM, then run normally Upon completion you get

##	IPM2v0.xx#	###	* ############################# #####	###########	##	*########
#						
#	command	:	./fish -n 10000			
#	start	:	Tue Feb 10 11:05:21 2012	host	:	nid06027
#	stop	:	Tue Feb 10 11:08:19 2012	wallclock	:	177.71
#	mpi_tasks	:	25 on 2 nodes	%comm	:	1.62
#	mem [GB]	:	0.24	gflop/sec	:	5.06

Maybe that's enough. If so you're done.

Have a nice day 🙂



IPM : IPM_PROFILE=full

#	host : s05601/006035	314C00_AIX	mpi_tasks	: 32 on 2 node	es
#	start : 11/30/04/14:3	5:34	wallclock	: 29.975184 se	ec
#	stop : 11/30/04/14:3	6:00	%comm	: 27.72	
#	gbytes : 6.65863e-01 t	otal	gflop/sec	: 2.33478e+00	total
#		[total]	<avg></avg>	min	max
#	wallclock	953.272	29.7897	29.6092	29.9752
#	user	837.25	26.1641	25.71	26.92
#	system	60.6	1.89375	1.52	2.59
#	mpi	264.267	8.25834	7.73025	8.70985
#	%comm		27.7234	25.8873	29.3705
#	gflop/sec	2.33478	0.0729619	0.072204	0.0745817
#	gbytes	0.665863	0.0208082	0.0195503	0.0237541
#	PM_FPU0_CMPL	2.28827e+10	7.15084e+08	7.07373e+08	7.30171e+08
#	PM_FPU1_CMPL	1.70657e+10	5.33304e+08	5.28487e+08	5.42882e+08
#	PM_FPU_FMA	3.00371e+10	9.3866e+08	9.27762e+08	9.62547e+08
#	PM_INST_CMPL	2.78819e+11	8.71309e+09	8.20981e+09	9.21761e+09
#	PM_LD_CMPL	1.25478e+11	3.92118e+09	3.74541e+09	4.11658e+09
#	PM_ST_CMPL	7.45961e+10	2.33113e+09	2.21164e+09	2.46327e+09
#	PM_TLB_MISS	2.45894e+08	7.68418e+06	6.98733e+06	2.05724e+07
#	PM_CYC	3.0575e+11	9.55467e+09	9.36585e+09	9.62227e+09
#		[time]	[calls]	<%mpi>	<%wall>
#	MPI_Send	188.386	639616	71.29	19.76
#	MPI_Wait	69.5032	639616	26.30	7.29
#	MPI_Irecv	6.34936	639616	2.40	0.67
#	MPI_Barrier	0.0177442	32	0.01	0.00
#	MPI_Reduce	0.00540609	32	0.00	0.00
#	MPI_Comm_rank	0.00465156	32	0.00	0.00
#	MPI_Comm_size	0.000145341	32	0.00	0.00



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Analyzing IPM Data



Tracing: Hard but Thorough



Advice: Develop (some) portable approaches to app optimization

- There is a tradeoff between vendorspecific and vendor neutral tools
 - Each have their roles, vendor tools can often dive deeper
- Portable approaches allow apples-toapples comparisons
 - Events, counters, metrics may be incomparable across vendors
- You can find printf most places
 - Put a few timers in your code?





So you run your code with a perf tool and get some numbers...what do they mean?



Performance: Definitions

How do we measure performance?

Speed up = $T_s/T_p(n)$ Efficiency = $T_s/(n^*T_p(n))$ Be aware there are - multiple definitions for these

terms

n=cores, s=serial, p=parallel

Isoefficiency: contours of constant efficiency amongst all problem sizes and concurrencies





Parallel speedups for x•x, how much?



Systematic Perf Measurement

- Scaling studies involve changing the degree of parallelism. Will we be changing the problem also?
- Strong scaling : Fixed problem size
- Weak scaling: Problem size grows with additional resources
- Optimization: Is the concurrency chosing the problem size or vice versa?



Sharks and Fish: Cartoon

Data:

n_fish is global my_fish is local fish_i = {x, y, ...}

Dynamics:

 $V \approx \sum_{ij} \frac{1}{\gamma_{ij}}$ F = maH = K + V $\& = -\partial H / \partial p$ $\& = -\partial H / \partial q$



MPI_Allgatherv(myfish_buf, len[rank], . for (i = 0; i < my fish; ++i) { for (j = 0; j < n fish; ++j) { // i!=j $a_i += g * mass_i * (fish_i - fish_i) / r_{ii}$ **Move fish**

See a glimpse here: http://www.leinweb.com/snackbar/wator/

Sharks and Fish Results

Running on a NERSC machine

- 100 fish can move 1000 steps in
 - 1 task \rightarrow 5.459s32 tasks \rightarrow 2.756s

• 1000 fish can move 1000 steps in

1 task	\rightarrow	511.14s	v 24.6 speedup
32 tasks	\rightarrow	20.815s∫	x 24.0 Speedup

• What's the "best" way to run?

- How many fish do we really have?
- How large a computer do we have?
- How much "computer time" i.e. allocation do we have?
- How quickly, in real wall time, do we need the answer?



Good 1st Step: Do runtimes make sense?

Running fish_sim for 100-1000 fish on 1-32 CPUs we see



Isoeffciencies



Too much communication

Sharks and Fish (MPI)



Scaling studies are not always so simple



How many perf measurements?



The scalability landscape



Not always so tricky

Main loop in jacobi_omp.f90; ngrid=6144 and maxiter=20



Scaling Studies at Scale



Let's look at scaling performance in more depth. A key impediment is often load imbalance



Load Imbalance : Pitfall 101

Communication Time: 64 tasks show 200s, 960 tasks show 230s



Load imbalance is pernicious





Load Balance : cartoon



Load Balance: Summary

Imbalance often a byproduct of

1) data decomposition or 2) multi-core concurrency quirks
 •Must be addressed **before** further MPI tuning can happen
 •For regular grids consider padding or contracting
 •Good software exists for graph partitioning / remeshing







•Dynamical load balance may be required for adaptive codes



Other performance scenarios



Simple Stuff: What's wrong here?



Communication Event Statistics (100.00% detail)

Buffer Size	Ncalls	Total Time	Min Time	Max Time	%MPI	%Wall
8	3278848	124132.547	0.000	114.920	59.35	16.88
0	35173439489	43439.102	0.000	41.961	20.77	5.91
98304	13221888	15710.953	0.000	3.586	7.51	2.14
196608	13221888	5331.236	0.000	5.716	2.55	0.72
589824	206848	5166.272	0.000	7.265	2.47	0.70
	Buffer Size 8 0 98304 196608 589824	Buffer SizeNcalls83278848035173439489983041322188819660813221888589824206848	Buffer SizeNcallsTotal Time1003278848124132.5471003517343948943439.1021983041322188815710.953196608132218885331.2365898242068485166.272	Buffer SizeNcallsTotal TimeMin Time1003278848124132.5470.0001003517343948943439.1020.0001983041322188815710.9530.000196608132218885331.2360.0005898242068485166.2720.000	Buffer SizeNcallsTotal TimeMin TimeMax Time10003278848124132.5470.000114.92010003517343948943439.1020.00041.9611983041322188815710.9530.0003.586196608132218885331.2360.0005.7165898242068485166.2720.0007.265	Buffer SizeNcallsTotal TimeMin TimeMax Time%MPI13278848124132.5470.000114.92059.3503517343948943439.1020.00041.96120.77983041322188815710.9530.0003.5867.51196608132218885331.2360.0005.7162.555898242068485166.2720.0007.2652.47



Science

Not so simple: Comm. topology



Application Topology



Performance in Batch Queue Space





A few notes on queue optimization

Consider your schedule

- Charge factor
 - regular vs. low
- Scavenger queues
- Xfer queues
 - Downshift concurrency

Consider the queue constraints

- Run limit
- Queue limit
- Wall limit
 - Soft (can you checkpoint?)

Jobs can submit other jobs





Marshalling your own workflow

- Lots of choices in general
 - Hadoop, CondorG, MySGE
- On hopper it's easy

```
#PBS -I mppwidth=4096
aprun –n 512 ./cmd &
aprun –n 512 ./cmd &
```

. . .

```
aprun –n 512 ./cmd &
```

wait

#PBS -I mppwidth=4096
while(work_left) {
 if(nodes_avail) {
 aprun -n X next_job &
 }
 wait
}



