

# Tools for Performance Debugging HPC Applications

**David Skinner** 

deskinner@lbl.gov









#### **Tools for Performance Debugging**

#### Practice

- Where to find tools
- Specifics to NERSC and Hopper

# Principles

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

# Scope & Audience

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





#### One Slide about NERSC

- Serving all of DOE Office of Science
  - domain breadth
  - range of scales
- Science driven
  - sustained performance

- Lots of users
  - ~4K active
  - ~500 logged in
  - ~300 projects
- Architecture aware
  - procurements driven by workload needs



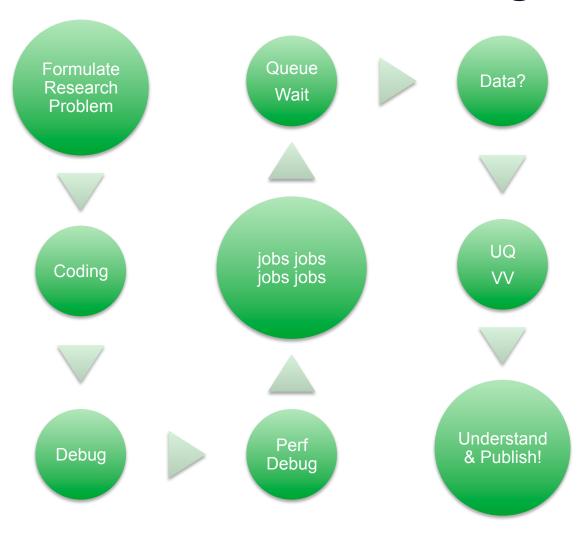


# Big Picture of Performance and Scalability





# Performance is more than a single number



- •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



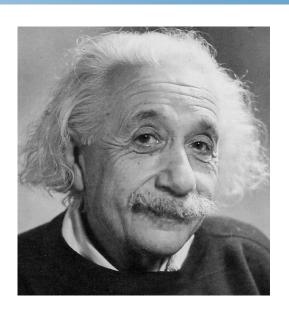




#### Performance is Relative

# To your goals

- Time to solution, T<sub>q</sub>+T<sub>wall</sub> ...
- 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.







# **Specific 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

Registers

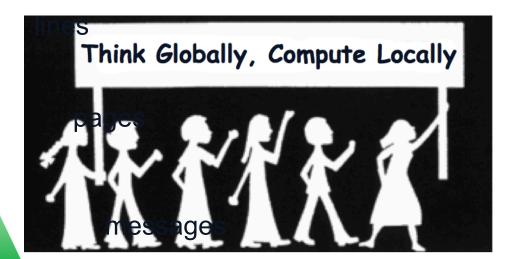
instructions & operands

Caches

**Local Memory** 

**Remote Memory** 

Disk / Filesystem



blocks, files







# ...on to specifics about HPC tools

Mostly at NERSC but fairly general

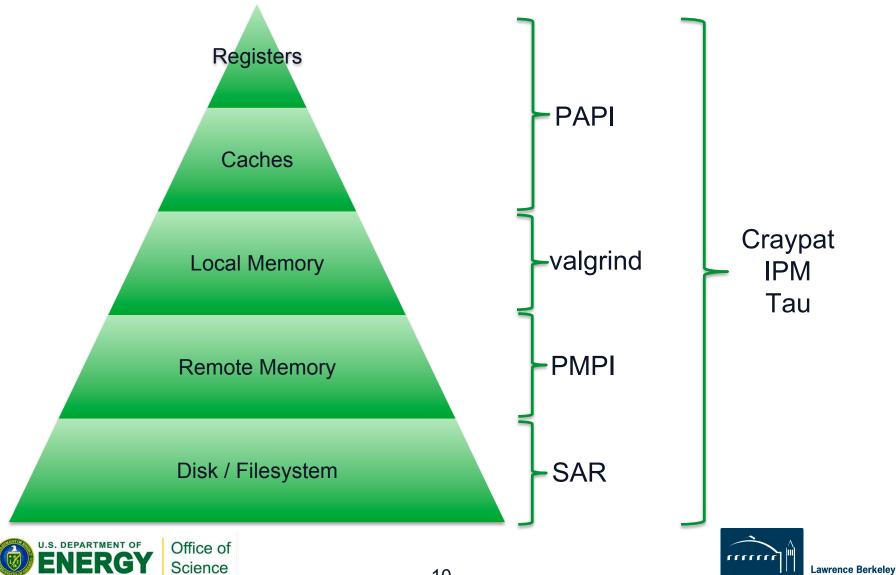








### **Tools are Hierarchical**



**National Laboratory** 

#### **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

#### 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
- Performance monitoring in production







- 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





## **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)

<sup>\*</sup> 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

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

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

Have a nice day ©





# IPM: IPM\_PROFILE=full

#	host : s05601/0060	35314C00_AIX	mpi_tasks	: 32 on 2 nod	es
#	start : 11/30/04/14	:35:34	wallclock	: 29.975184 s	ec
#	stop : 11/30/04/14	:36:00	%comm	: 27.72	
#	gbytes : 6.65863e-01	total	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_FPUO_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





# Advice: Develop (some) portable approaches to performance

- 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?







# **Examples of HPC tool usage**





# Scaling: definitions

- Scaling studies involve changing the degree of parallelism. Will we be change the problem also?
- Strong scaling
  - Fixed problem size
- Weak scaling
  - Problem size grows with additional resources
- Speed up =  $T_s/T_p(n)$
- Efficiency = T<sub>s</sub>/(n\*T<sub>p</sub>(n))

Be aware there are multiple definitions for these terms





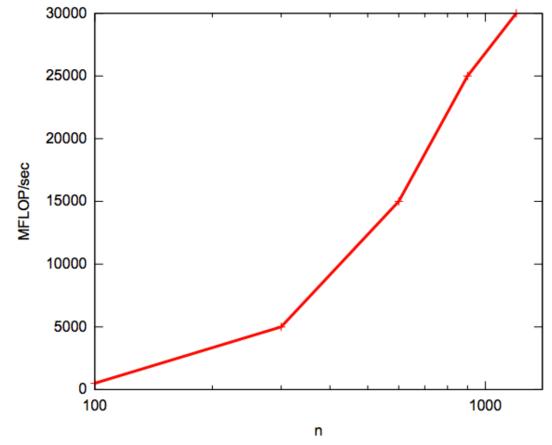
# Conducting a scaling study

# With a particular goal in mind, we systematically vary concurrency and/or problem size

#### **Example:**

How large a 3D (n^3) FFT can I efficiently run on 1024 cpus?

Looks good?







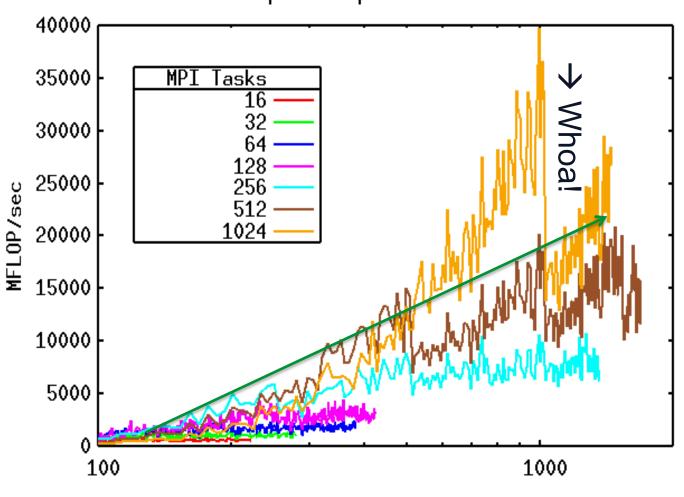
# Let's look a little deeper....





# The scalability landscape





#### Why so bumpy?

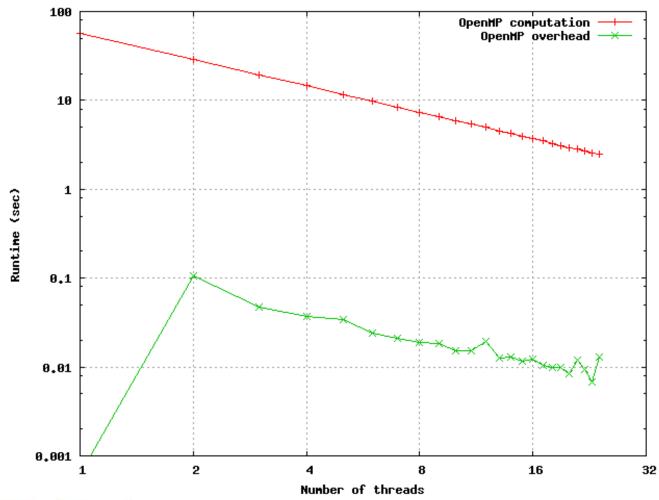
- Algorithm complexity or switching
- Communication protocol switching
- Inter-job contention
- ~bugs in vendor software





# Not always so tricky

#### Main loop in jacobi\_omp.f90; ngrid=6144 and maxiter=20



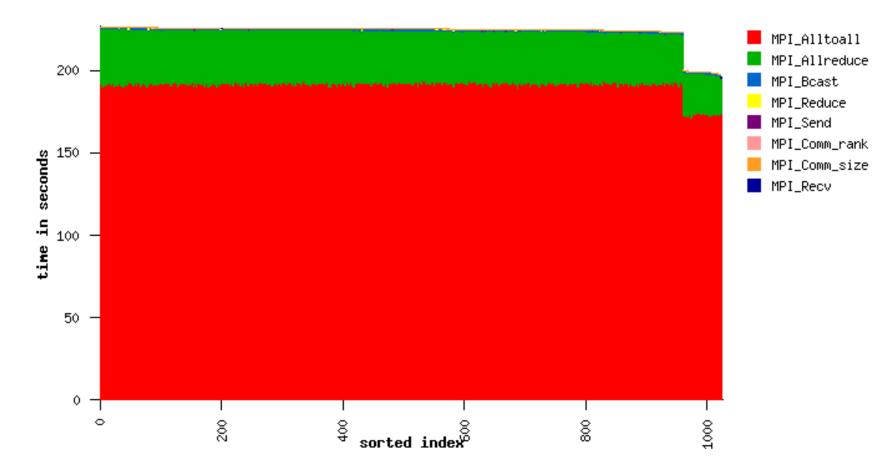






#### **Load Imbalance: Pitfall 101**

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



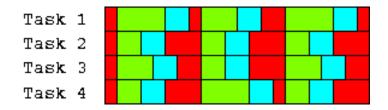
MPI ranks sorted by total communication time





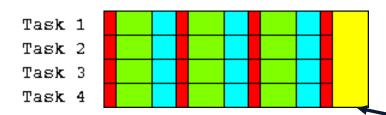
### **Load Balance: cartoon**

#### **Unbalanced:**



# Universal App Sync Flop I/0

#### Balanced:

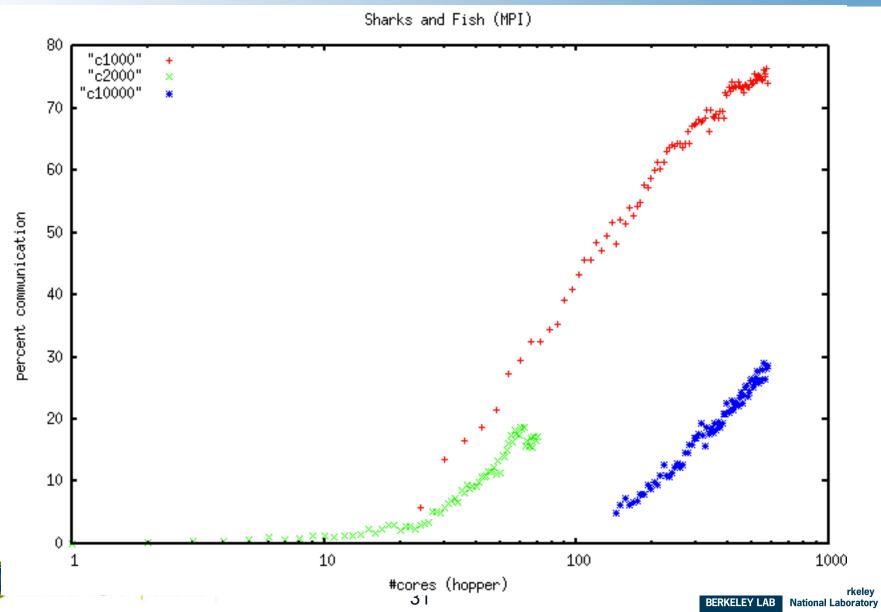


Time saved by load balance



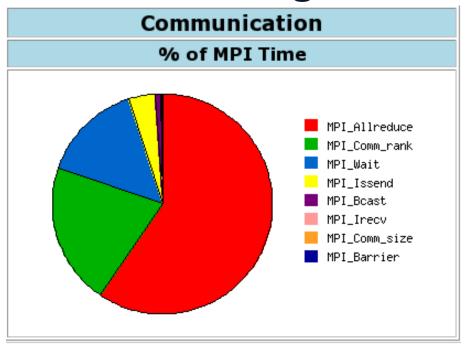


## Too much communication





# Simple Stuff: What's wrong here?



#### Communication Event Statistics (100.00% detail)

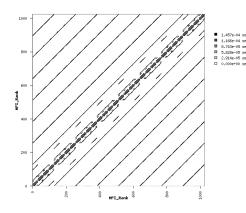
	Buffer Size	Ncalls	Total Time	Min Time	Max Time	%MPI	%Wall
MPI_Allreduce	8	3278848	124132.547	0.000	114.920	59.35	16.88
MPI_Comm_rank	0	35173439489	43439.102	0.000	41.961	20.77	5.91
MPI_Wait	98304	13221888	15710.953	0.000	3.586	7.51	2.14
MPI_Wait	196608	13221888	5331.236	0.000	5.716	2.55	0.72
MPI_Wait	589824	206848	5166.272	0.000	7.265	2.47	0.70



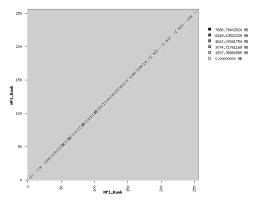
Office of Science



# Not so simple: Comm. topology

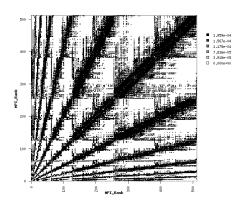


MILC

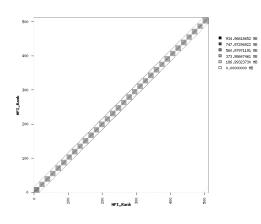


**PARATEC** 

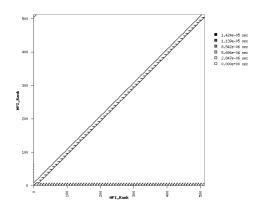




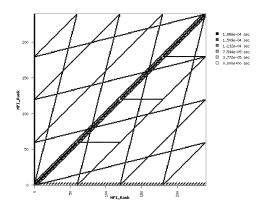
**MAESTRO** 



**IMPACT-T** 



**GTC** 



CAM





# 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

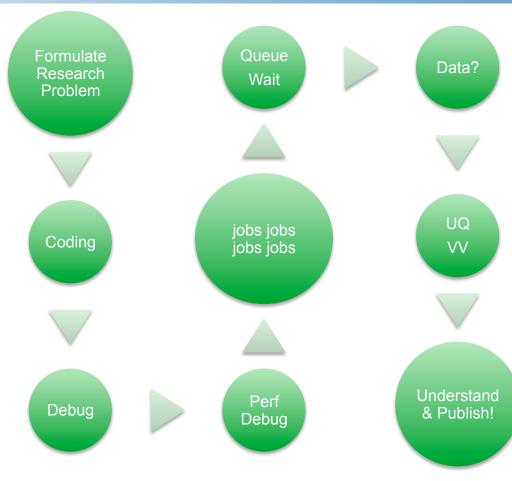
```
#PBS -I mppwidth=4096
aprun -n 512 ./cmd &
aprun -n 512 ./cmd &
...
aprun -n 512 ./cmd &
wait
```

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





#### Thanks!



#### **Contacts:**

help@nersc.gov deskinner@lbl.gov



