

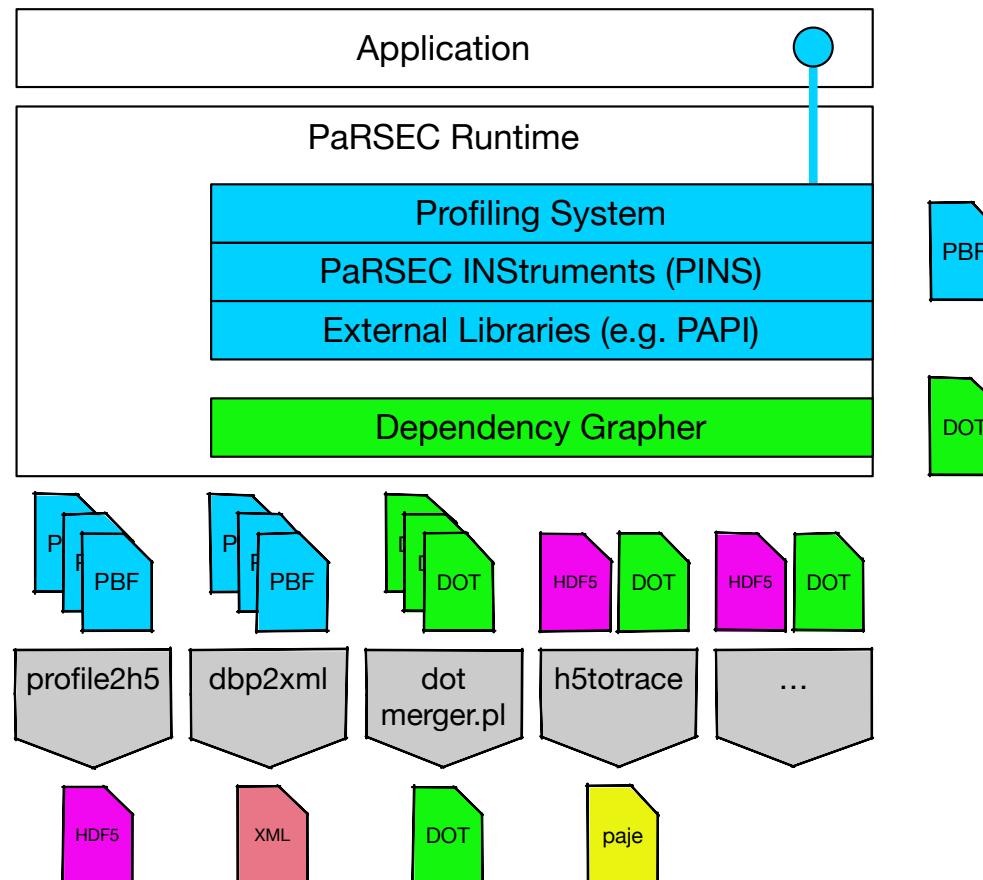
Profiling Tools in PaRSEC

PaRSEC User Group Meeting



Profiling System

- Two main runtime components:
 - Dependency Grapher produces a DOT of the DAG of tasks.
 - Cost is non-negligible (I/O at the preamble and epilogue of each task)
 - Profiling System keeps a track of the execution
 - Significant effort to keep that cost down
 - Modular design: goal is to allow power users to inject their own instrumentation if needed
 - Provide interface to PAPI
 - Each component produces one file per rank
- A collection of user-level tools to process the produced files



Compile PaRSEC with Profiling

- Two options:
 - PARSEC_PROF_TRACE
 - Enables ‘profiling’: events can be logged in a binary format
 - Can also use OTF2 format, if libotf2 is available, but functionality is reduced
 - PARSEC_PROF_GRAPHER
 - Enables dependency graphing
- Dependencies:
 - Python 2.7.15 or later and cython 0.21.2 for most profiling tools (optional, but recommended)
 - PAPI for the pins_papi MCA module (optional)

```

parsec/build/fast-profiling > ../../configure \
    --prefix=$HOME/parsec/install-dir/fast-profiling \
    -DPARSEC_PROF_TRACE=ON \
    -DPARSEC_PROF_TRACE_SYSTEM="PaRSEC Binary Tracing
Format" \
    -DPARSEC_PROF_GRAPHER=ON

[...]

-- Found Python: /nfs/apps/spack/opt/spack/linux-
centos7-x86_64/gcc-7.2.0/python-2.7.15-
aczt3ejnkaiikvwqqy7btuccpm2bpgnm/bin/python2.7 (found
version "2.7.15") found components: Interpreter
Development
-- Cython version 0.29.15 found
-- Found Cython: /nfs/apps/spack/opt/spack/linux-
centos7-x86_64/gcc-7.2.0/python-2.7.15-
aczt3ejnkaiikvwqqy7btuccpm2bpgnm/bin/cython (Required is
at least version "0.21.2")
-- -- Found Component 'pins'
-- The PAPI Library is found at /spack/opt/spack/linux-
scientific7-x86_64/gcc-7.3.0/papi-5.6.0-
dybzvixufstomkhem7ayjmyjdpfprcsc/lib/libpapi.so
-- ---- Module 'papi' is ON
-- Profiling uses PaRSEC Binary Tracing Format

```

Run PaRSEC with profiling

- MCA parameter: `profile_filename` -- where to store the PBF files.
 - This is a template name, files will be named `<profile_filename>-<rank>.prof-<random>`
- PaRSEC parameter: `--dot [<filename>]`
 - This parameter is for PaRSEC, not for the app
 - `<filename>` is also a template: the files will be named `<filename>-<rank>.dot`
- Only one of them may be specified
- It is critical for performance that these files are as local as possible (/tmp, scratch directories, etc...)
 - By default, the tracing system uses trunc/mmap to resize the events files, and this sometimes fails on NFS directories. A compilation variable, `PARSEC_PROFILING_USE_MMAP` can be undefined in `profiling.c` if this prevents profiling to execute
 - 3 MCA parameters exist to tune the profiling:
 - `profile_buffer_pages`: how many pages per buffer per thread are allocated
 - `profile_file_resize`: increment at which event files are resized (in number of buffers)
 - `profile_show_profiling_performance`: displays at the end of the execution the time spent in profiling routines

```
> mpirun -map-by node -np 2 \
  ./tests/stencil/testing_stencil_1D \
  -- --mca profile filename \
  /scratch/shared/herault/stencil
[****] TIME(s)          0.00158 : Stencil N= 8 NB= 4
M= 8 MB= 4 PxQ= 1 2 KPxKQ= 1 1 Iteration= 10
Radius= 1 Kernel type= 0 Number_of_buffers= 2
cores= 20 : 0.001213 gflops
```

```
> ls /scratch/shared/herault/
stencil-1.prof-Hx6eWl    stencil-0.prof-Hx6eWl
```

```
> mpirun -map-by node -np 2 \
  ./tests/stencil/testing_stencil_1D \
  -- --dot /scratch/shared/herault/stencil
W@000000 /!\ DEBUG LEVEL WILL PROBABLY REDUCE THE
PERFORMANCE OF THIS RUN /!\
[****] TIME(s)          0.00745 : Stencil N= 8 NB= 4
M= 8 MB= 4 PxQ= 1 2 KPxKQ= 1 1 Iteration= 10
Radius= 1 Kernel type= 0 Number_of_buffers= 2
cores= 20 : 0.000258 gflops
```

```
> ls /scratch/shared/herault/
stencil-0.dot    stencil-0.prof-Hx6eWl
stencil-1.dot    stencil-1.prof-Hx6eWl
```

Implementation

- Trace collection is per-stream
- 1 Computing thread in PaRSEC = 1 Profile Stream
- Comm. Thread has 3 Profile Streams (task notification, payload send, payload receive)
- CUDA devices create one stream per CUDA stream
- Most tracing operations on streams are independent: no atomics.
- Each stream appends events of variable sizes on its own even buffer
- Buffer management - MMAP
 - 1 additional thread is created to resize the backend file, and pre-allocate 1 buffer / stream in advance in that file
 - goal is to minimize wait time to acquire a new buffer
 - cost is a file size that is overestimated, and I/O in parallel with computation
 - this thread holds 99% of its time in blocking FS calls or idle on semaphores
- Buffer management – append to file
 - Buffer allocation is centralized
 - buffers are dumped on file one after the other, in a serial manner

User API

- Most calls are done by the runtime system, without requiring anything from the user:
 - trace of internal events (memory allocations, notifications, memory transfers)
 - trace of inter-process messages
- DSLs also automatically decorate their code with tracing calls
 - trace of start and end of task execution
- High-level programs can (optionally) add information to the trace
 - On any rank:
 - `void profiling save_dinfo(const char *key, double value);`
 - `void profiling save_iinfo(const char *key, int value);`
 - `void profiling save_uint64info(const char *key, uint64_t value);`
 - `void profiling save_sinfo(const char *key, const char *svalue);`
 - NB: if two ranks define the same key, the HDF5 will only show one of the values

PaRSEC INStrumentation (PINS)

- In addition to the profiling system, the runtime has hooks placed at many critical steps of a task lifecycle:
 - first time a task is discovered
 - every time one of its input flow becomes ready
 - when the task becomes ready to execute
 - when it prepares its input data
 - when it starts (possibly multiple times) executing, every time it returns from the execution
 - when it is released
- Each of these places can become a logged event using the PINS system
- PINS MCA components can then decorate each event with more information.
 - Example: PINS PAPI module:
-mca mca_pins papi --mca pins_papi 1 --mca pins_papi_event "S*:C*:PAPI_L1_DCM"



Tools

- Some tools are binary executables
 - parsec-dbp2xml, parsec-dbp2mem, parsec-dbpinfo, ...
- Others are Python scripts
 - profile2h5, h5totrace, ...
- Some are Perl scripts
 - parsec-dotmerger
- Although it is possible to use them within the build/source directories, the easiest way to use them is to install and load the environment provided in the bin/subdirectory after install in bash.env or csv.env

```
> make -j 20 install  
> export PARSEC_ROOT=[...]  
> . $PARSEC_ROOT/bin/bash.env  
> which parsec-ptgpp  
[...]/bin/parsec-ptgpp
```

Tool: **parsec-dotmerger**

- Takes a set of per-node DOT files, and merges them in one DOT file
- DOT is the format used by graphviz (<https://www.graphviz.org/>) to visualize graphs (in our case the DAG of tasks)
- **parsec-dotmerger** has options to
 - Select what nodes or edges to ignore
 - Select the content, form and color of nodes and edges
 - See **parsec-dotmerger -h** for a full list of options

```
> $PARSEC_ROOT/bin/parsec-dotmerger \
  /scratch/shared/herault/stencil-0.dot \
  /scratch/shared/herault/stencil-1.dot > \
  stencil.dot

> dot -Tpdf -o stencil.pdf stencil.dot
```



Tool: profile2h5

- The binary profiling format is needs to be assembled into a portable file.
- parsec-dbp2xml provides only rudimentary information (tasks start and end dates)
- the recommended approach is to convert the parsec binary format files into pandas dataframes stored in an HDF5 file
- This is the role of profile2h5
 - Note: profile2h5 will check if a target hdf5 file already exists. If it exists, it will not re-generate it
 - i.e. if you want to re-generate an hdf5 from a new profile with the same name, you need to move or delete the existing hdf5 file.
- profile2h5 depends on pandas, with pytables for HDF5 support. Minimal versions are:
 - pandas: 0.24.2
 - numpy: 1.16.6
 - tables: 3.5.1
- All these tools can be installed in the user directory using pip

```
> cd /scratch/shared/herault/
> profile2h5 stencil-* .prof-Hx6eW1
Processing ['stencil-0.prof-Hx6eW1',
'stencil-1.prof-Hx6eW1']
Generated: stencil-ap-Hx6eW1.h5
```

```
> file stencil-ap-Hx6eW1.h5
stencil-ap-Hx6eW1.h5: Hierarchical Data
Format (version 5) data
```

HDF5 Profiling files

- Hierarchy of pandas DataFrame
 - Most data is in events
 - Meta-data is in information, nodes, streams, and event_names/event_types
 - information holds a key-value store that is application-specific
 - event_names/event_types are key-value stores to identify the events
 - nodes and streams are DataFrames to identify the processes and the streams in each process
 - Streams are profiling streams: there might be more than one per thread (e.g. 3 for the comm. thread, one per CUDA stream, etc..)

errors

event_attributes

event_convertors

event_names

event_types

events

information

nodes

streams

HDF5 Profiling: /nodes

- Basic information about the nodes
 - id is the value used in other DataFrames for node_id
 - HWLOC-XML is a dump of the hwloc topology loaded on the node, if hwloc is available.
 - sched is the scheduler loaded at init time
 - nb_cores is the number of computing threads (including the main thread) used by this run
 - MEMORY_USAGE and MEMORY_USAGE_list hold some statistics of mempool memory usage internal of PaRSEC (data repositories and tasks contexts)

/nodes	CMDLINE
	DEVICE_MODULES
	DIMENSION
	GIT_BRANCH
	GIT_HASH
	HWLOC-XML
	MEMORY_USAGE
	MEMORY_USAGE_list
	cwd
	error
	exe
	exe_abspath
	filename
	hostname
	id
	nb_cores
	nb_vps
	sched

HDF5 Profiling: /streams

- begin/end/duration: times related to the stream (in unit of the realtime timer of the machine that did the run, usually nanoseconds)
- boundto: binding information (core). Displayed as a float. NaN for non-binding
- description: human-readable information about the stream
- node_id/th_id/vp_id: the identifiers of the node/thread/virtual process that hosts this stream
- stream_id: the identifier used in other DataFrames to relate to this stream

/streams

begin

end

duration

boundto

description

node_id

stream_id

th_id

vp_id

HDF5 Profiling: /information

- Key/Value store: used by the application to store meta-information
 - For the stencil example, here is the information stored by the application. They overlap some information stored in the DataFrame /nodes.
 - This is purely user-defined.

/information	CMDLINE
	DEVICE_MODULES
	DIMENSION
	GIT_BRANCH
	GIT_HASH
	cwd
	error
	exe
	exe_abspath
	last_error
	nb_cores
	nb_nodes
	nb_vps
	sched
	worldsize

HDF5 Profiling: /event_types

- /event_types: key-value store that maps event names (string) to event type (number)
- /event_names: key-value store that maps event types (number) to event names (string)

/event_types	N/A
	TASK_MEMORY
	Device delegate
	cuda
	movein
	moveout
	prefetch
	cuda_mem_alloc
	cuda_mem_use
	MPI_ACTIVATE
	...
/event_names	0
	1
	2
	3
	4
	5
	6
	7
	8
	...

HDF5 Profiling: /events

- Large DataFrame holding all ‘events’
- Rows in this dataframe are a pair of event: an event.begin and an event.end.
- Each event pair happens on a given stream (stream_id), on a given node (node_id), and has a unique type, a begin time and an end time.
- Some events are related to a given taskpool (e.g. execution of a task) (taskpool_id)
- Other events are related to a data (e.g. a GPU memory transfer) and have a data_collection_unique_key and a data_collection_data_id.
- Some events are network-related (e.g. communication) and have a source (src), a destination (dst), and sometimes a task identifier (did – type identifier of the task/tid – task identifier in this type).
- Each event type may define additional columns in this table, and the values in these columns make sense only for these event types.

/events	begin
	end
	stream_id
	node_id
	type
	taskpool_id
	data_collection_unique_key
	data_collection_data_id
	data_collection_padding
	src
	dst
	did
	tid
	... [taskpool-specific]

HDF5 Profiling: example.

```
>>> import pandas as pd
>>> import numpy as np
>>> t=pd.HDFStore('stencil-ap-Hx6eWl.h5')
>>> t.events.taskpool_id.unique()
array([2, -1], dtype=object)
>>> m = t.events[t.events.taskpool_id == 2]
>>> m['duration'] = m['end']-m['begin']
>>> m['duration'] = m['duration'].astype(np.int)
>>> m[['duration','type']].groupby(['type']).describe()

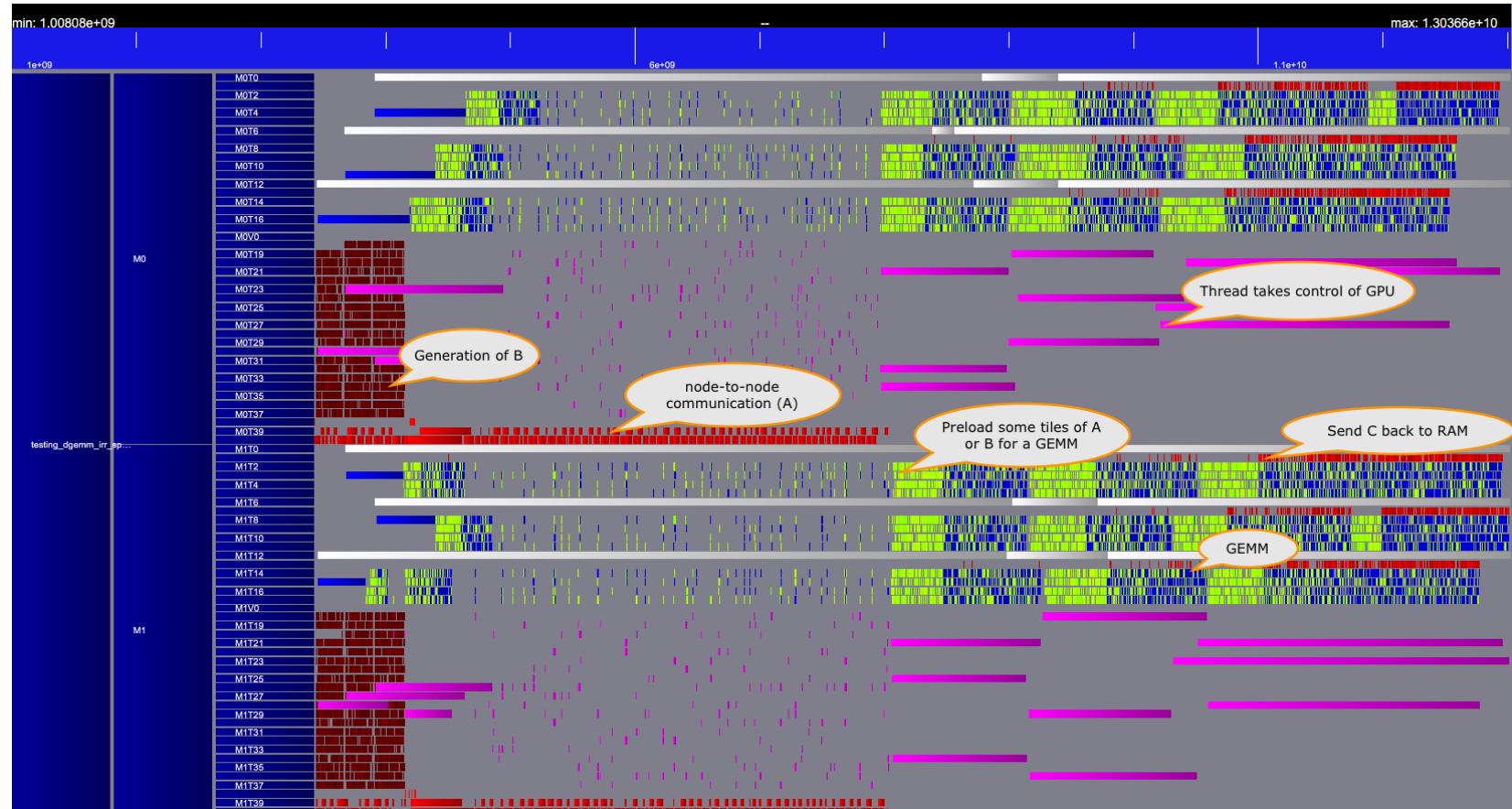
      duration
      count      mean       std      min     25%     50%     75%      max
type
15          22.0  780.590909  297.918758   269.0    624.5    724.0    980.0   1373.0
>>> t.event_names[15]
'task'
```

HDF5 Profiling: example.

```
>>> c=t.events[((t.events.type >= t.event_types['MPI_ACTIVATE']) &  
                 (t.events.type <= t.event_types['MPI_DATA_PLD_RCV']))]  
  
>>> c['duration']=c['end'].astype(np.int)-c['begin'].astype(np.int)  
  
>>> c[['duration','node_id']].groupby('node_id').describe()  
  
           duration  
  
          count      mean       std      min     25%     50%     75%      max  
  
node_id  
0        16.0  1.070306e+04  1.199261e+04  1461.0  2082.25  2739.0  16566.50  33097.0  
1        16.0  1.587329e+06  3.404030e+06  1308.0  1914.75  4253.0  8713.75  8449498.0
```

Tool: h5totrace

- Converts an HDF5 profiling database into a PAJE trace (<http://paje.sourceforge.net/>)
- Can be visualized with Paje (old), or Vite (still in dev.): <http://vite.gforge.inria.fr/>
- See h5totrace – help for list of options



- counter: some events (e.g. memory allocation events) are better represented as a line that accumulates values than a block. This option allows to select which event names are considered counters
- ignore-type / –ignore-stream: allows to trim the trace to make it easier to process or visualize
- list: just list the event types in the HDF5 (useful for –ignore-type)
- dot / –dot-DAG : optional, can take the DOT information to add arrows between tasks to represent the DAG of dependencies on top of the Gantt diagram
- COMM: represents communications between nodes with arrows

Ad-hoc visualizations

