

# Module 6 Profiling in Linux

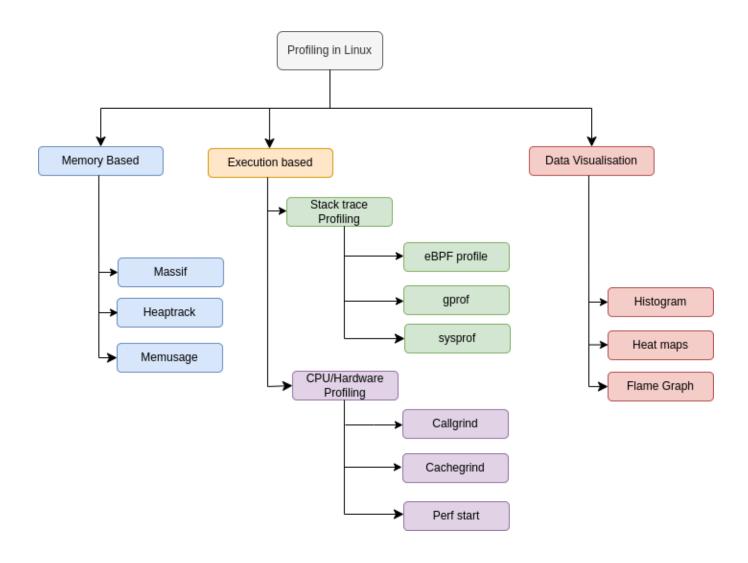


#### Introduction

- Profiling is the process of collecting and analyzing performance data from user applications or system-wide to understand their behavior and identify performance bottlenecks.
- A profiling tool uses sampling techniques to collect data at defined intervals.
- Profiling provides an aggregated overview of system performance to identify bottlenecks, while tracing captures detailed, sequential event data to analyze execution flow and debug specific issues.
- Profiling can be done on different aspects of the system such as:
  - Function exection time
  - o Function call count
  - Memory usage
  - Hardware and Software events



## Profiling landscape in Linux





#### **Data Visualization**



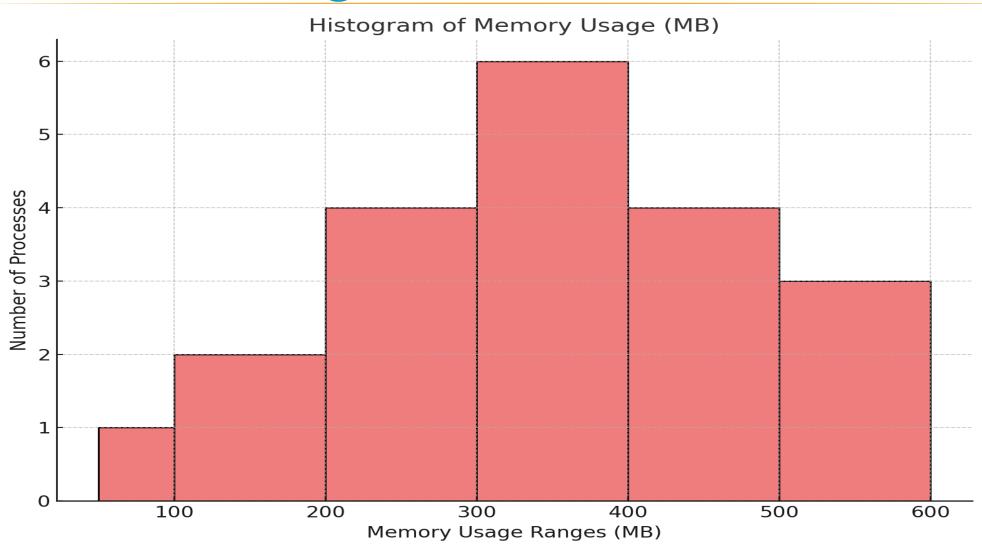
#### Histogram

- A histogram is a type of graph that shows how data is distributed by grouping it into intervals or "bins."
- It helps understand the following:
  - o Frequency of data points within specific ranges.
  - o Understand patterns or trends in data.
  - o Identify the shape of the data distribution (e.g., normal, skewed).
  - o Spot outliers or unusual data points.
- It can be used to visualize the performance data collected by profiling tools, providing insights into resource utilization and identifying areas for optimization.
- Example: Analyzing the memory usage of processes running on a Linux server over a certain period.

```
Memory usage (in MB) for 20 processes: [50, 120, 180, 200, 210, 230, 250, 300, 320, 330, 350, 370, 390, 410, 430, 450, 470, 500, 520, 550]
```



## Histogram Visualization



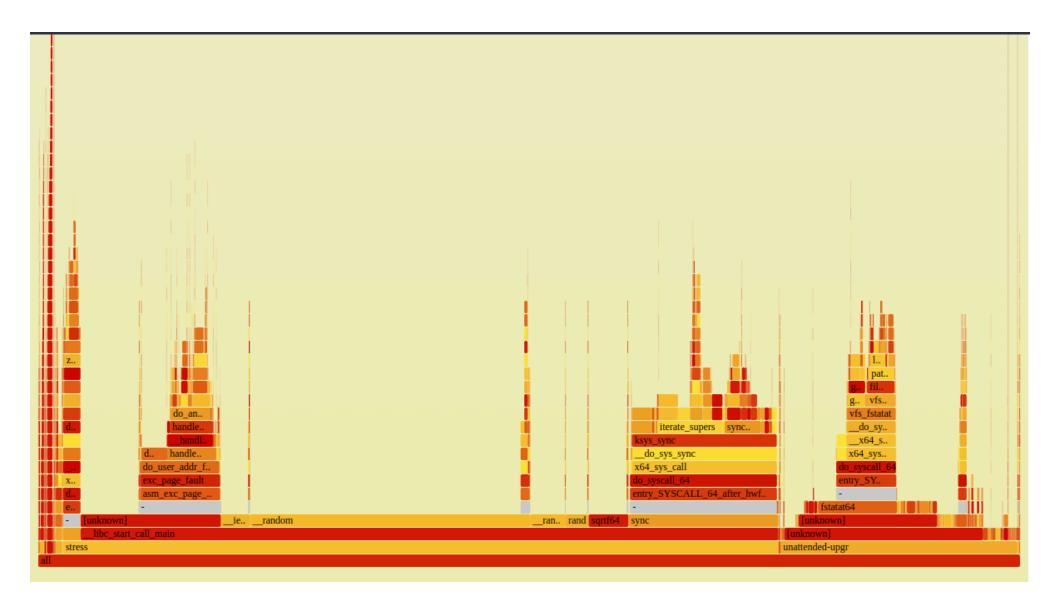


## FlameGraph

- Flame graphs are a type of visualization that represents sampled stack traces of a program's execution.
- Flame graphs are commonly used in performance analysis to identify performance bottlenecks, CPU hotspots, and inefficient code paths.
- The colors often represent different functions or function categories, aiding in quick identification.
- The vertical axis (Y-axis) represents the stack depth, showing the sequence of function calls counting from zero at the bottom.
- The horizontal axis (X-axis) represents stack profile population, sorted alphabetically. Each stack frame is represented by a horizontal bar, with wider bars indicating functions that consume more CPU time.
- Please refer to the instructions for flame graph generation.



## FlameGraph Visualization





#### Memory profiling

- Memory profiling tools highlight inefficient memory usage patterns such as excessive allocations, unnecessary copying of data, and redundant memory accesses.
- Helps in finding and fixing memory leaks by tracking all memory allocations and deallocations, ensuring that memory is properly released.
- Enhances cache utilization by promoting better memory locality, allowing data to be accessed more quickly from the CPU cache.
- Reduces the need for frequent page swaps by maintaining a smaller working set in memory, thus minimizing paging and disk I/O.



#### Massif

- Massif is a memory profiler tool in Valgrind, used to analyze heap allocations, stack usage, and overall memory consumption of a program.
- It provides detailed analysis of heap memory usage, aiding in the identification of memory leaks and inefficient memory usage patterns.
- Massif tool can be invoked using following command line.

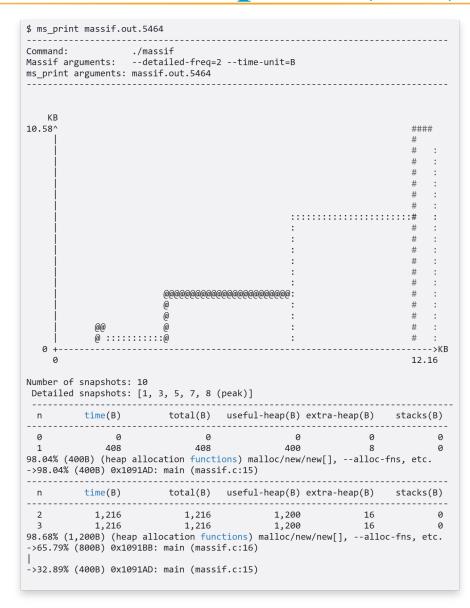
```
valgrind --tool=massif --time-unit=B <Application Program>
```

- Upon completion, Valgrind does not print summary statistics; instead, all of Massif's profiling data is written to a file, typically named massif.out.{pid} (filename can be changed using --massif-out-file option).
- You can then use the ms\_print tool to visualize a heap allocation graph.

```
ms_print massif.out.5464
```



#### Massif report (1/2)





#### Massif report (2/2)

```
n time(B) total(B) useful-heap(B) extra-heap(B) stacks(B)
______
4 1,624 808 800 8 0
5 3,632 2,816 2,800 16 0
99.43% (2,800B) (heap allocation functions) malloc/new/new[], --alloc-fns, etc.
->71.02% (2,000B) 0x10918F: f (massif.c:10)
| ->71.02% (2,000B) 0x109204: main (massif.c:26)
->28.41% (800B) 0x1091BB: main (massif.c:16)
->00.00% (0B) in 1+ places, all below ms_print's threshold (01.00%)
n time(B) total(B) useful-heap(B) extra-heap(B) stacks(B)
6 7,640 6,824 6,800 24 0
7 11,648 10,832 10,800 32 0
99.70% (10,800B) (heap allocation functions) malloc/new/new[], --alloc-fns, etc.
->73.86% (8,000B) 0x10917A: g (massif.c:5)
| ->36.93% (4,000B) 0x109194: f (massif.c:11)
| | ->36.93% (4,000B) 0x109204: main (massif.c:26)
 ->36.93% (4,000B) 0x109209: main (massif.c:28)
->18.46% (2,000B) 0x10918F: f (massif.c:10)
->18.46% (2,000B) 0x109204: main (massif.c:26)
->07.39% (800B) 0x1091BB: main (massif.c:16)
->00.00% (0B) in 1+ places, all below ms print's threshold (01.00%)
n time(B) total(B) useful-heap(B) extra-heap(B) stacks(B)
______
99.70% (10,800B) (heap allocation functions) malloc/new/new[], --alloc-fns, etc.
->73.86% (8,000B) 0x10917A: g (massif.c:5)
->36.93% (4,000B) 0x109194: f (massif.c:11)
 ->36.93% (4,000B) 0x109204: main (massif.c:26)
 ->36.93% (4,000B) 0x109209: main (massif.c:28)
->18.46% (2,000B) 0x10918F: f (massif.c:10)
 ->18.46% (2,000B) 0x109204: main (massif.c:26)
->07.39% (800B) 0x1091BB: main (massif.c:16)
->00.00% (0B) in 1+ places, all below ms_print's threshold (01.00%)
 n time(B) total(B) useful-heap(B) extra-heap(B) stacks(B)
 9 12,456 10,024 10,000 24 0
```

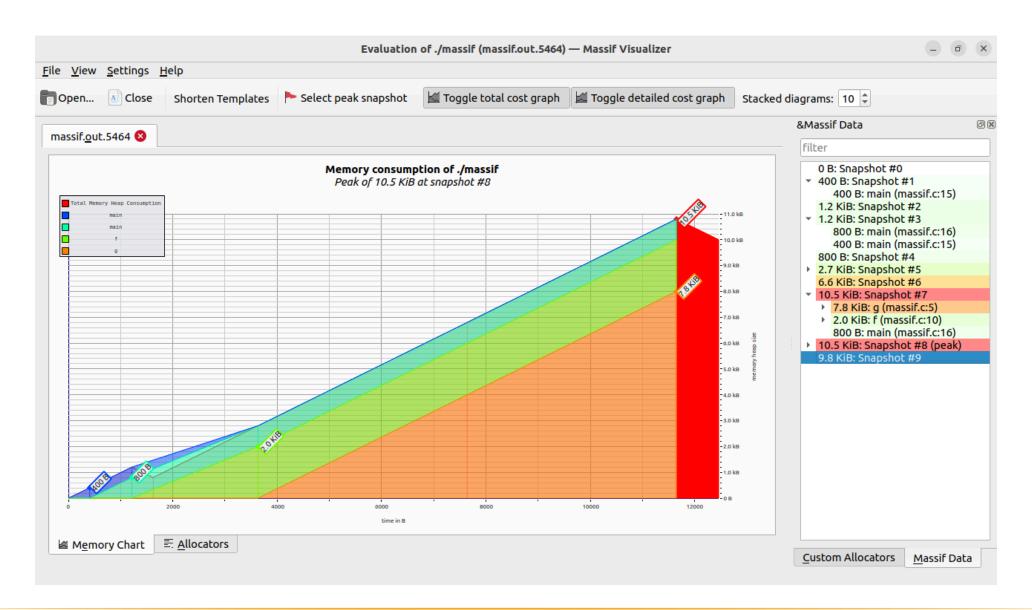


#### Massif result analysis

- We are using massif sample program for this demonstration.
- Each vertical bar represents a snapshot, indicating the memory usage at a specific point in time.
- The columns are further divided into:
  - Normal snapshots, which record basic information and are represented using the ':' character.
  - **Detailed snapshots**, denoted by the '@' character, represent information about where allocations occurred for these snapshots.
  - Peak snapshots, represented by the '#' character, record the point where memory consumption was greatest.
- The graph is followed by the information for each snapshot.
- Normal snapshots provide basic memory usage counts, while detailed snapshots include an allocation tree indicating code sections responsible for heap memory allocation.



#### Massif Visualizer



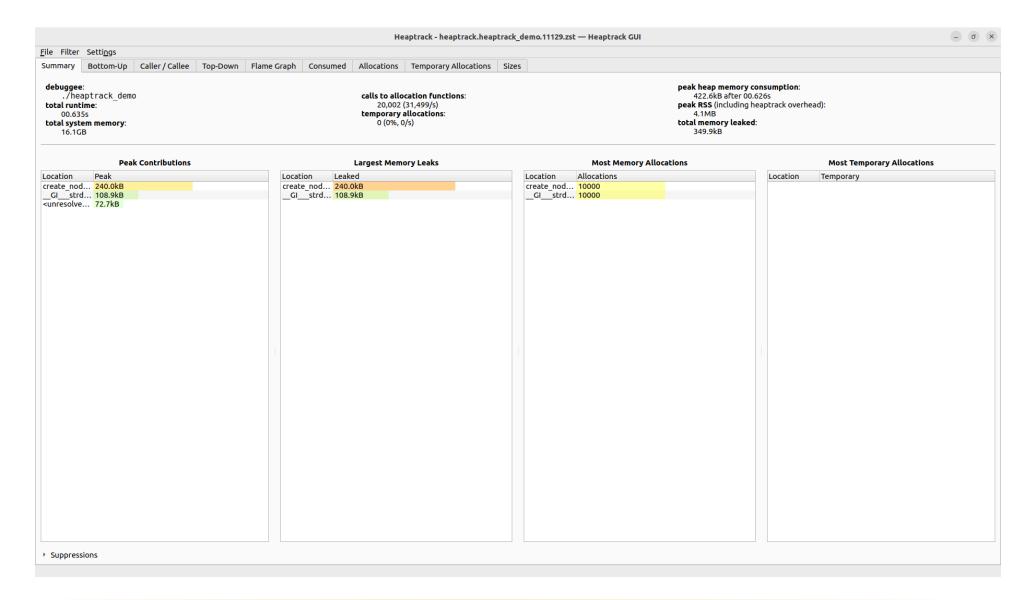


#### Heaptrack

- Heaptrack traces all memory allocations and annotates them with stack traces.
- Uses LD\_PRELOAD to intercept memory allocators.
- Analysis tools help identify hotspots for optimization, memory leaks, allocation hotspots, and temporary allocations.
- Heaptrack offers more detailed allocation context compared to Massif and incurs low overhead.
- Execute using heaptrack {program}.
- After execution, heaptrack generates a heaptrack. {APP}. {PID}.zst file that can be analyzed using heaptrack\_print or heaptrack\_gui.
- We are using heaptrack sample program for the demonstration.



## Heaptrack GUI





#### Memusage

- memusage profiles the memory usage of user-space applications.
- It utilizes the libmemusage.so library by preloading it into the caller's environment via the LD\_PRELOAD environment variable.
- libmemusage.so library traces memory allocation by intercepting memory allocators like malloc, calloc, free, realloc, mmap, mremap and munmap.
- The output data can be collected in a textual form, or in a PNG file for graphical representation.
- memusage is a part of libc-devtools package on Ubuntu.
- Demonstration uses the following memusage example.



#### Memusage example

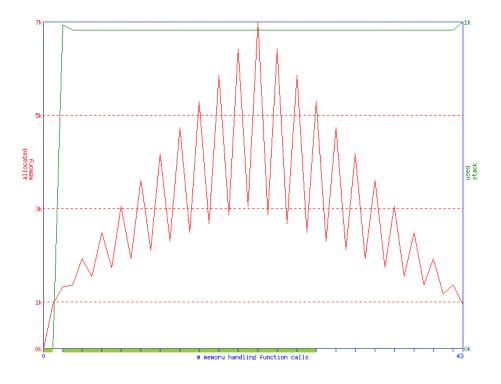
```
manas@sandbox:~/work/memusage$ memusage --data=memusage.dat ./userapp
Memory usage summary: heap total: 45464, heap peak: 7464, stack peak: 1968
        total calls total memory failed calls
malloc
                           1424
realloc
               40
                                              (nomove:37, dec:19, free:0)
                          44040
calloc
  free
                            440
Histogram for block sizes:
 240-255
                        2% ========
 400-415
 432-447
 640-655
 832-847
1024-1039
                    1 2% ========
1040-1055
1232-1247
                    2 4% =============
1440-1455
1632-1647
1840-1855
                    2 4% ==============
2032-2047
2240-2255
2832-2847
3440-3455
4032-4047
4640-4655
5232-5247
 5840-5855
6432-6447
                        2% =======
```



#### Memusage graphical representation

- Convert the collected data into a PNG format

```
```bash
manas@sandbox:~/work/memusage$ memusagestat memusage.dat memusage.png
```
```





# Execution based profiling



#### eBPF profiler

- profile-bpfcc is an eBPF-based CPU profiler that captures stack trace samples at regular intervals.
- It helps analyze CPU usage by identifying which code is executing and measuring its impact, covering both user-space and kernel execution.
- Compared to other profilers, eBPF is efficient and low-overhead. It counts stack trace frequencies in the kernel and passes only unique stacks and their counts to user space, reducing kernel-to-user transfers.
- By default, it samples at 49 Hz across all CPUs which is adjustable via a command-line option. Frequencies like 49 Hz or 99 Hz help avoid lock-step sampling issues compared to 50 or 100 Hz.
- The profiler is a part of bpfcc-tools package.



## eBPF profiler usage

- The profiler can provide output in folded format directly. We can further pipe the output to flamegraph.pl directly to skip the intermediate file.
- The command samples at 49 Hz (-F 49), includes kernel annotations (-a), separates user and kernel stacks with a delimiter (-d), outputs in folded format (-f), and runs for 30 seconds.

```
# git clone https://github.com/brendangregg/FlameGraph

# sudo apt-get install bpfcc-tools

# cd FlameGraph

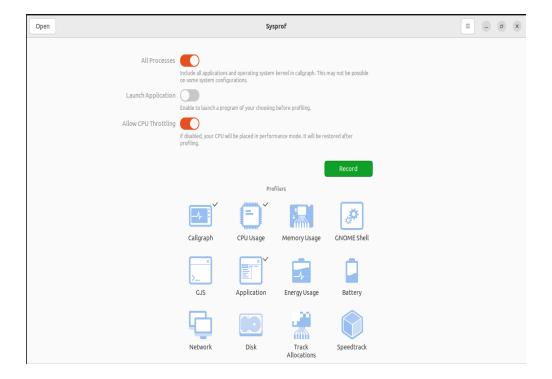
# Profiler output piped into flamegraph.pl
# profile-bpfcc -F 49 -adf 30 | ./flamegraph.pl > profile.svg

# firefox profile.svg
```



## Sysprof (1/2)

- Sysprof is a Linux sampling CPU profiler which records a stack trace of a process's activity multiple times per second.
- Sysprof is a system-wide Linux profiler that includes the kernel and all userspace applications.
- On Ubuntu platform, install it using sudo apt install sysprof.



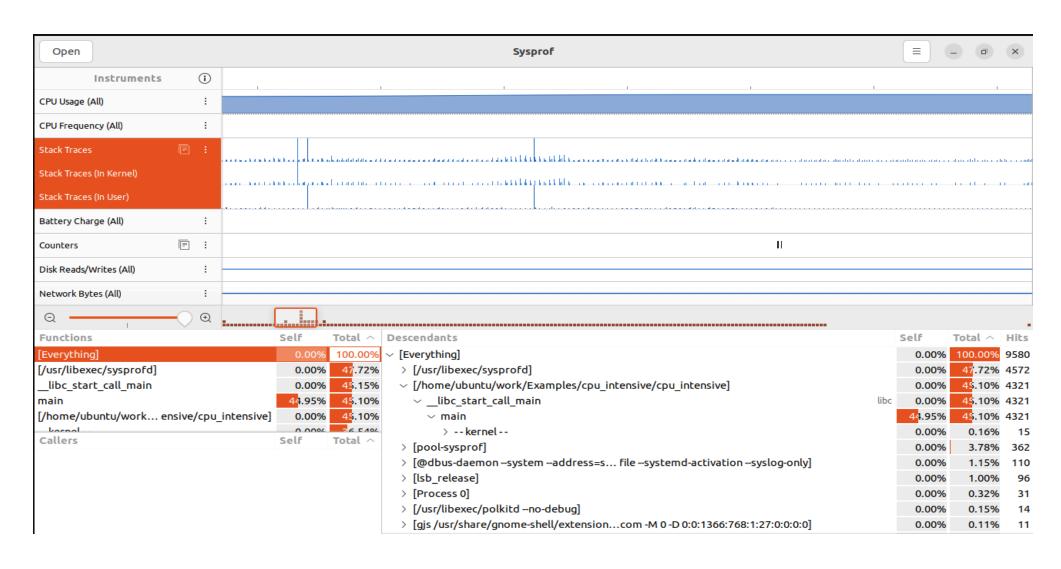


# Sysprof (2/2)

- Sysprof supports profiling of entire systems or specific processes. It also provides the flexibility to launch an application before recording the profile.
- Sysprof analysis includes CPU usage, thread execution, and function call stacks, helping identify performance bottlenecks and optimize applications for better performance.
- It uses frame pointer unwinding to record the stack trace. Ensure to use the fno-omit-frame-pointer compiler option to instruct the compiler to store the stack frame pointer in a register.
- Following links provide detailed insight on frame pointers and working of profiler: Link-1, Link-2 and Link-3.



#### Sysprof example





## GNU Profiler (gprof) [1/2]

- gprof is a profiling tool used to analyze the performance of user space application in terms of function call frequencies and execution times.
- It is typically used with programs written in C or C++ compiled with the lag to generate profiling information.

```
$ gcc -o memory main.c -g -pg
```

• After compiling the program with -pg, execute it to generate a gmon.out file. Then, run gprof with the executable and gmon.out as arguments.

```
# Call graph output
$ gprof memory --graph gmon.out
# Flat Profile output
$ gprof memory -z gmon.out
```

• It produces a report showing the time spent in each function and the number of times each function was called.



## GNU Profiler (gprof) [2/2]

- The generated report includes a flat profile (summary of each function) and a call graph (visual representation of function calls).
- Gprof helps identify bottlenecks and areas for optimization in the program's execution.
- It may not be suitable for multithreaded or heavily optimized programs; other tools like perf may be more appropriate in such cases.
- Refer to the gprof manual for detail on usage and limitation.



## Callgrind (1/2)

- callgrind is a part of the Valgrind tool suite which records the call history among functions in a program's run as a call-graph.
- The default collected data includes the number of executed instructions, their association with source lines, the caller-callee relationship between functions, and the counts of such calls.
- Callgrind propogates costs across function boundaries.
  - o e.g: function foo calls bar, the costs from bar are added into foo's costs.
  - Provides an inclusive costs of entire program, where the cost of each function includes the costs of all functions it called, directly or indirectly.
- Callgrind usage: valgrind --tool=callgrind [callgrind options] yourprogram [program options]



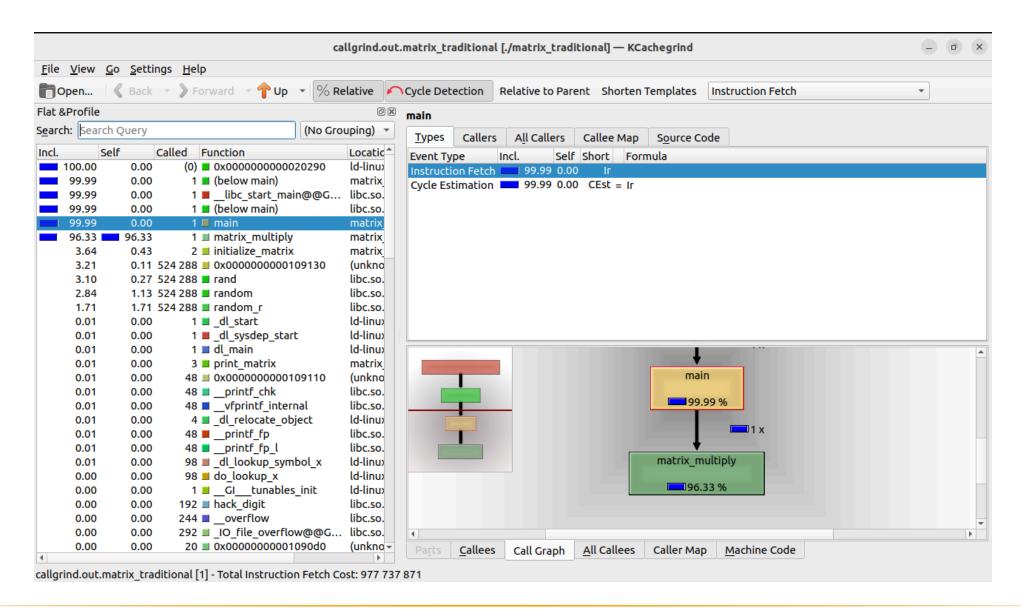
## Callgrind (2/2)

```
ubuntu@sandbox:~/work/Examples/matrix/$ valgrind --tool=callgrind ./matrix_traditional
==7474== Callgrind, a call-graph generating cache profiler
==7474== Copyright (C) 2002-2017, and GNU GPL'd, by Josef Weidendorfer et al.
==7474== Using Valgrind-3.18.1 and LibVEX; rerun with -h for copyright info
==7474== Command: ./matrix_traditional
...
==7474== Events : Ir
==7474== Collected : 977737730
==7474==
==7474== I refs: 977,737,730
```

- We are using the matrix multiplication example for the illustration given above.
- The detailed profiling data is written to a file named callgrind.out.{pid}
- callgrind\_annotate is the command line tool which reads in the profile data, and prints a sorted lists of functions, optionally with source annotation.
- kcachegrind can provide the callgrind report in GUI form.
- For more details on advance usage, please refer to callgrind manual.



#### **KCachegrind**





## Cachegrind (1/2)

- Cachegrind is a cache profiler tool included in the Valgrind suite and it simulates program interaction with the CPU cache hierarchy.
- Cachegrind collects precise and reproducible profiling data.
  - **Precise**: Cachegrind precisely counts program instructions and provides detailed data at file, function, and line levels.
  - **Reproducible**: Instruction counts are reliably reproducible, often perfectly so. This enables precise measurement of even minor program changes.
- The cache simulator, simulates a machine with split (Independent instruction and data cache) L1 cache with a unified L2 cache.
- For the accurate results, the program should be compiled with debugging symbols and optimization turned ON.



## Cachegrind (2/2)

- Usage: \$ valgrind --tool=cachegrind ./program
- The detailed profiling data is written to a file named cachegrind.out.{pid}
- Cachegrind output details cache misses, hit rates, and access patterns, aiding optimization with visualized hotspots.
- cg\_annotate is a command line tool which summarizes cache-related information per function and annotates source code with color-coded markers for cache events.
- kcachegrind can provide the cachegrind report in GUI form.
- For more details on advance usage, please refer to cachegrind manual.



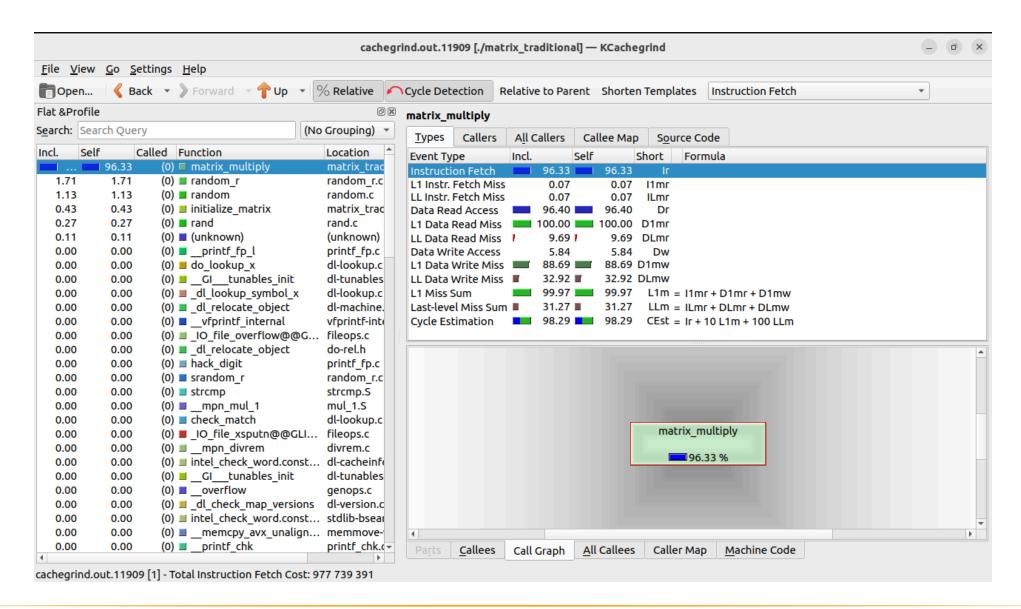
#### Cachegrind usage

• We are using the matrix multiplication example for the illustration given below.

```
ubuntu@sandbox:~/work/Examples/matrix$ valgrind --tool=cachegrind ./matrix traditional
==11909== Cachegrind, a cache and branch-prediction profiler
==11909== Copyright (C) 2002-2017, and GNU GPL'd, by Nicholas Nethercote et al.
==11909== Using Valgrind-3.18.1 and LibVEX; rerun with -h for copyright info
==11909== Command: ./matrix traditional
==11909==
==11909==
==11909== I refs: 977,739,391
==11909== I1 misses:
                          1,532
==11909== LLi misses: 1,525
==11909== I1 miss rate: 0.00%
==11909== LLi miss rate:
                          0.00%
==11909==
==11909== D \text{ refs}: 282,943,530 (278,458,337 rd + 4,485,193 wr)
==11909== D1 misses: 134,808,940 (134,513,355 rd + 295,585 wr)
==11909== LLd misses: 51,364 ( 1,590 rd + 49,774 wr)
==11909== D1 miss rate: 47.6% ( 48.3% + 6.6% )
==11909== LLd miss rate: 0.0% ( 0.0% + 1.1% )
==11909==
==11909== LL refs: 134,810,472 (134,514,887 rd + 295,585 wr)
==11909== LL misses:
                         52,889 ( 3,115 rd + 49,774 wr)
==11909== LL miss rate:
                            0.0% (
                                    0.0\% + 1.1\% )
```



#### **KCachegrind**





## Perf stat (1/2)

- perf stat is a Linux performance measurement tool that collects and reports detailed hardware and software performance counters such as CPU cycles, instructions, cache accesses, and branch mispredictions.
- By collecting and displaying detailed performance metrics, perf stat aids in the performance tuning and optimization of applications.
- Use the perf list command to view all available hardware and software events that can be monitored.

```
ubuntu@sandbox:~/work/linux-kernel/linux-6.2/tools/perf$ ./perf list
List of pre-defined events (to be used in -e or -M):
 branch-instructions OR branches
                                                      [Hardware event]
 branch-misses
                                                      [Hardware event]
 cgroup-switches
                                                      [Software event]
                                                      [Software event]
 context-switches OR cs
 L1-dcache-load-misses
                                                      [Hardware cache event]
 L1-dcache-loads
                                                      [Hardware cache event]
  cache-misses OR cpu/cache-misses/
                                                       [Kernel PMU event]
```



#### Perf stat (2/2)

• perf stat provides options to capture specific events for a command.

```
perf stat -e cycles:uk dd if=/dev/zero of=/dev/null count=100000
```

- We can also use modifiers to count events in userspace (:u) or kernel space (:k).
- If there are more events than counters, the kernel uses multiplexing to give each event a chance to access the monitoring hardware.
- This leads the tool to scale the count and provide an estimate based on total time enabled vs time running.
- Reduce the number of events with each run to avoid scaling.
- Refer to perf wiki for more details.
- The standard Linux perf tool package offers basic support. Following these instructions, building your own version is recommended.



#### Perf stat example

```
root@sandbox:/home/ubuntu/work/linux-kernel/linux-6.2/tools/perf# ./perf stat -B dd if=/dev/zero of=/dev/null count=1000000
1000000+0 records in
1000000+0 records out
512000000 bytes (512 MB, 488 MiB) copied, 1.76965 s, 289 MB/s
 Performance counter stats for 'dd if=/dev/zero of=/dev/null count=1000000':
                                                             0.998 CPUs utilized
         1,770.13 msec task-clock
                       context-switches
               14
                                                           7.909 /sec
                       cpu-migrations
                                                             2.825 /sec
                       page-faults
                                                            45.194 /sec
                       cycles
    5,47,18,32,553
                                                             3.091 GHz
    2,93,68,65,191
                      instructions
                                                        # 0.54 insn per cycle
      50,06,69,465
                     branches
                                                        # 282.843 M/sec
         61,40,189
                       branch-misses
                                                           1.23% of all branches
      1.774383620 seconds time elapsed
      0.781372000 seconds user
      0.989737000 seconds sys
```

- In the example above, perf stat captures software events like context switches and generic hardware events like CPU cycles.
- The derived metrics such as "Instruction per cycle" are presented after the '#' sign.



# **CPU Frequency Profiling**



## Understanding CPU Frequency Scaling

- Modern CPUs dynamically adjust frequency to balance power and performance.
- Frequency is scaled based on workload demands and system policy higher frequency = more performance, lower = more power savings.
- Linux uses scaling governors: powersave, performance and schedutil. The OS governor's job is to provide hints or requests to the underlying hardware based scalar.
- Hardware scalar (HWP) offload frequency decisions to the CPU itself. Intel Speed Shift is a prime example of HWP.
- CPU frequency impacts:
  - Response time
  - Instruction throughput
  - o Power and thermal envelope



## Tools for CPU Frequency Observation

- cpupower: A CLI tool to view and set CPU scaling governors or frequency limits system-wide, such as performance or powersave.
- cpufreq-info: Displays per-core scaling governor, available frequencies, driver (e.g., intel\_pstate), and current policy. It is useful to verify CPU capabilities and scaling behavior.
- s-tui: An interactive terminal-based UI that monitors real-time CPU frequency, utilization, temperature, and power. It is ideal for visualizing the impact of workload on CPU scaling.
- turbostat: Provides per-core CPU MHz, Bzy\_MHz (active cycles), package power (PkgWatt), and temperature (PkgTemp). It's excellent for measuring fine-grained hardware-level CPU performance under load.
- perf stat: Part of the Linux perf toolset. It tracks instructions, cycles, IPC, user/sys time, and cache events. Best suited for profiling workload efficiency and execution cost.



## Demonstration: Powersave vs Performance

• Workload: Synthetic CPU stress test using sysbench or stress-ng.

| Metric         | Expectation in Powersave | Expectation in Performance |
|----------------|--------------------------|----------------------------|
| CPU Frequency  | Low, dynamic             | High, locked at max        |
| Instructions   | Similar                  | Similar                    |
| Cycles         | Higher (slower)          | Lower (faster)             |
| IPC            | Lower                    | Higher                     |
| Total CPU Time | Longer                   | Shorter                    |
| Power/Temp     | Lower                    | Higher                     |

- Set scaling governer:
  - o **Performance**: sudo cpupower frequency-set -g performance
  - o Powersave: sudo cpupower frequency-set -g powersave
- Refer to this helper script to dump the CPU frequency results.



#### Demonstration: Key observation

- Higher and more stable frequencies in performance mode.
- Lower CPU time, higher IPC in perf stat.
- Higher Avg\_MHz, Bzy\_MHz and higher temparature in turbostat.
- CPU governor has a significant impact on workload efficiency.
  - o performance mode reduces runtime by keeping frequency high
  - o powersave mode may hurt latency-sensitive tasks
- Use frequency profiling to:
  - o Optimize performance vs power trade-offs
  - o Debug unexpected slowdowns in real-world deployments



#### References (1/2)

- Histogram Statquest
  - https://www.youtube.com/watch?v=qBigTkBLU6g
- Flamegraph -Brendan Gregg
  - https://www.youtube.com/watch?v=VMpTU15rIZY
- eBPF Brendan Gregg
  - https://www.youtube.com/watch?v=HKQR7wVapgk
- sysprof
  - https://fedoramagazine.org/performance-profiling-in-fedora-linux/
- Debugging with Frame pointers
  - https://developers.redhat.com/articles/2023/07/31/frame-pointersuntangling-unwinding#
  - https://blogs.oracle.com/linux/post/unwinding-stack-frame-pointers-andorc



#### References (2/2)

- Perf tool building instructions Manas Marawaha.
  - https://medium.com/@manas.marwah/building-perf-tool-fc838f084f71
- SIMD and AVX2 instruction set
  - https://www.codeproject.com/Articles/874396/Crunching-Numbers-with-AVX-and-AVX2
  - https://en.ittrip.xyz/c-language/c-matrix-simd-optimization
  - https://v0dro.in/blog/2018/05/01/building-a-fast-matrix-multiplicationalgorithm/