## **Featherlight Reuse-distance Measurement**

#### **APPENDIX**

## A. TIME REUSE TO STACK REUSE

RDX allocates several bins to include every sampled reuse pair with time distance falling into different ranges. Common to prior work [1, 2], the histogram uses the logarithmic scale for the bin size as the horizontal axis and the percentage of total sampled reuse pairs as the vertical axis.

With the time distance histogram, RDX leverages the existing technique [1] to approximate the reuse distance histogram. This section only shows the intuitive idea of this approximation. Considering a memory access sequence aXXXXa, the time distance of a is 5 while we have no idea how many distinct elements between a. If three time reuses are of distance 1 and one reuses is of distance 5, we definitely know there should be only two distinct elements between a by exhaustive searching.

However, this greedy method usually does not work for calculating the reuse distance of a common application, especially directly calculating the reuse distance of the specific reuse instance of an element. Thus we seek to a statistical model to calculate the probability of a specific reuse distance instead of obtaining the reuse distance of a specific reuse instance. Assuming that the probability of a data element is independent from others, whether a data element is accessed in a given time interval  $\Delta$  is actually a Bernoulli process. Then the probability of having k distinct data elements in this  $\Delta$  interval is

$$P(k, \Delta) = \binom{N}{k} P_{interval}(\Delta)^k (1 - P_{interval}(\Delta))^{(N-k)}$$

where  $P_{interval}(\Delta)$  is the probability for a data element to appear in the interval  $\Delta$  and N is total number of distinct data elements. To calculate  $P_R(k)$  (the probability of having reuse distance k for the entire program), we need to consider all the possibilities of having reuse distance of k from the interval length ranging between 1 and T, where T is the total number of memory accesses. The probability of having time interval  $\Delta$  can be obtained from the time reuse histogram, denoted as  $P_T(\Delta)$ . Thus we have

$$P_R(k) = \sum_{\Delta=1}^{T} P(k, \Delta) P_T(\Delta)$$
 (5)

Since  $P_{interval}(\Delta)$  can be derived from  $P_T(\Delta)$  and the details can be found in [3], Equation (5) transforms time reuse histogram to stack reuse histogram.

Their later work [1] further developed the algorithm by extending each bar width of both time and stack reuse histograms from 1 to any arbitrary number, which is utilized in our work. Their evaluation shows that this model gives more than 99% accuracy as to cache block granularity.

## B. INPUT OF SPEC CPU BENCHMARKS

We use ref input to run all SPEC CPU benchmarks but some have multiple inputs, which are distinguished with numerical suffixes such as gcc-1, gcc-2, etc. The actual mapping of these names are shown in Figure 13 and Figure 14.

## C. STACK REUSE HISTOGRAMS OF SPEC CPU 2006 BENCHMARKS WITH OFF-BY-ONE PROBLEM

In SPEC CPU2006, bwaves (Figure 15), leslie3d (Figure 16), sphinx3 (Figure 17), GemsFDTD (Figure 18), lbm (Figure 19), and cactusADM (Figure 20) have the off-by-one problem, which can be inferred from their stack reuse histograms.

## D. STACK REUSE HISTOGRAMS OF SPEC CPU2017

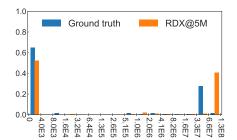
We have plotted the stack reuse histograms of perlbench\_r (Figure 21, Figure 22 and Figure 23), gcc\_r (Figure 24, Figure 25, Figure 26, Figure 27 and Figure 28), bwaves\_r (Figure 29, Figure 30, Figure 31 and Figure 32), mcf\_r (Figure 33), cactuBSSN\_r (Figure 34), namd\_r (Figure 35), povray\_r (Figure 36), lbm\_r (Figure 37), omnetpp\_r (Figure 38), wrf\_r (Figure 39), xalancbmk\_r (Figure 40), x264\_r (Figure 41, Figure 42 and Figure 43), blender\_r (Figure 44), cam4\_r (Figure 45), deepsjeng\_r (Figure 46), imagick\_r (Figure 47), leela\_r (Figure 48), nab\_r (Figure 49), exchange2\_r (Figure 50), fotonik3d\_r (Figure 51), roms\_r (Figure 52), xz\_r (Figure 53, Figure 54 and Figure 55), perlbench\_s (Figure 56, Figure 57 and Figure 58), gcc\_s (Figure 59, Figure 60 and Figure 61), bwaves\_s (Figure 62 and Figure 63), mcf\_s (Figure 64), cactuB-

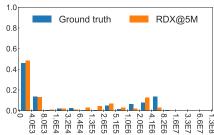
Benchmark	Input argument
perlbench-1	-I./lib checkspam.pl 2500 5 25 11 150 1 1 1 1
perlbench-2	-I./lib diffmail.pl 4 800 10 17 19 300
perlbench-3	-I./lib splitmail.pl 1600 12 26 16 4500
bzip2-1	input.source 280
bzip2-2	chicken.jpg 30
bzip2-3	liberty.jpg 30
bzip2-4	input.program 280
bzip2-5	text.html 280
bzip2-6	input.combined 200
gcc-1	166.i -o 166.s
gcc-2	200.i -o 200.s
gcc-3	c-typeck.i -o c-typeck.s
gcc-4	cp-decl.i -o cp-decl.s
gcc-5	expr.i -o expr.s
gcc-6	expr2.i -o expr2.s
gcc-7	g23.i -o g23.s
gcc-8	s04.i -o s04.s
gcc-9	scilab.i -o scilab.s
gamess-1	< cytosine.2.config
gamess-2	< h2ocu2+.gradient.config
gamess-3	< triazolium.config
gobmk-1	quietmode gtp" < 13x13.tst
gobmk-2	quietmode gtp" < nngs.tst
gobmk-3	quietmode gtp" < score2.tst
gobmk-4	quietmode gtp" < trevorc.tst
gobmk-5	quietmode gtp" < trevord.tst
soplex-1	-s1 -e -m45000 pds-50.mps
soplex-2	-m3500 ref.mps
hmmer-1	nph3.hmm swiss41
hmmer-2	fixed 0mean 500num 500000sd 350seed 0 retro.hmm
h264ref-1	-d foreman_ref_encoder_baseline.cfg
h264ref-2	-d foreman_ref_encoder_main.cfg
h264ref-3	-d sss_encoder_main.cfg
astar-1	BigLakes2048.cfg
astar-2	rivers.cfg

Figure 13: The input arguments of benchmarks which have several ref inputs for SPEC CPU 2006.

Benchmark	Input argument
perlbench r-1	-I./lib checkspam.pl 2500 5 25 11 150 1 1 1 1
	-I./lib diffmail.pl 4 800 10 17 19 300
	-I./lib splitmail.pl 6400 12 26 16 100 0
gcc_r-1	qcc-pp.c -03 -finline-limit=0 -fif-conversion -fif-
	conversion2 -o gcc-pp.opts-O3finline-limit_0fif-
	conversion -fif-conversion2.s
gcc_r-2	gcc-pp.c -02 -finline-limit=36000 -fpic -o gcc-pp.opts-
	02 -finline-limit 36000 -fpic.s
gcc_r-3	gcc-smaller.c -03 -fipa-pta -o gcc-smaller.opts-03fipa-
	pta.s
gcc_r-4	ref32.c -05 -o ref32.opts-05.s
gcc_r-5	ref32.c -03 -fselective-scheduling -fselective-scheduling2
	-o ref32.opts-03fselective-schedulingfselective- scheduling2.s
bwaves r-1	bwaves 1 < bwaves 1.in
	bwaves 2 < bwaves 2.in
	bwaves_3 < bwaves_3.in
	bwaves 4 < bwaves 4.in
x264 r-1	pass 1stats x264_stats.logbitrate 1000frames
x204_r-1	1000 -o BuckBunny_New.264 BuckBunny.yuv 1280x720
x264_r-2	pass 2stats x264_stats.logbitrate 1000dumpyuv
	200frames 1000 -o BuckBunny_New.264 BuckBunny.yuv 1280x720
	seek 500dumpyuv 200frames 1250 -o
x264_r-3	BuckBunny_New.264 BuckBunny.yuv 1280x720
xz_r-1	cld.tar.xz 160
	19cf30ae51eddcbefda78dd06014b4b96281456e078ca7c13e1c0c9e6a
	aea8dff3efb4ad6b0456697718cede6bd5454852652806a657bb56e07d
vz r-9	61128434b474 59796407 61004416 6 cpu2006docs.tar.xz 250
	055ce243071129412e9dd0b3b69a21654033a9b723d874b2015c774fac
	1553d9713be561ca86f74e4f16f22e664fc17a79f30caa5ad2c04fbc44
	7549c2810fae 23047774 23513385 6e
xz_r-3	input.combined.xz 250
	a841f68f38572a49d86226b7ff5baeb31bd19dc637a922a972b2e6d125 7a890f6a544ecab967c313e370478c74f760eb229d4eef8a8d2836d233
	d3e9dd1430bf 40401484 41217675 7
perlbench s-1	-I./lib checkspam.pl 2500 5 25 11 150 1 1 1 1
perlbench_s-2	-I./lib diffmail.pl 4 800 10 17 19 300
$perlbench\_s-3$	-I./lib splitmail.pl 6400 12 26 16 100 0
gcc_s-1	gcc-pp.c -05 -fipa-pta -o gcc-pp.opts-05fipa-pta.s
	gcc-pp.c -05 -finline-limit=1000 -fselective-scheduling -
$gcc\_s-2$	fselective-scheduling2 -o gcc-pp.opts-05finline-
	limit_1000fselective-schedulingfselective- scheduling2.s
	gcc-pp.c -05 -finline-limit=24000 -fgcse -fgcse-las -
gcc_s-3	fgcse-lm -fgcse-sm -o gcc-pp.opts-05finline-
	limit_24000fgcsefgcse-lasfgcse-lmfgcse-sm.s
bwaves_s-1	bwaves_1 < bwaves_1.in
bwaves_s-2	bwaves_2 < bwaves_2.in
	pass 1stats x264_stats.logbitrate 1000frames
x264_s-1	1000 -o BuckBunny New.264 BuckBunny.yuv 1280x720
x264_s-2	pass 2stats x264_stats.logbitrate 1000dumpyuv
	200frames 1000 -o BuckBunny_New.264 BuckBunny.yuv 1280x720
	seek 500dumpyuv 200frames 1250 -o
x264_s-3	BuckBunny_New.264 BuckBunny.yuv 1280x720
v7 e_1	cpu2006docs.tar.xz 6643
	055ce243071129412e9dd0b3b69a21654033a9b723d874b2015c774fac
	1553d9713be561ca86f74e4f16f22e664fc17a79f30caa5ad2c04fbc44
xz_s-2	7549c2810fae 1036078272 1111795472 4 cld.tar.xz 1400
	19cf30ae51eddcbefda78dd06014b4b96281456e078ca7c13e1c0c9e6a
	aea8dff3efb4ad6b0456697718cede6bd5454852652806a657bb56e07d
	61128434b474 536995164 539938872 8

Figure 14: The input arguments of benchmarks which have several  $\mathit{ref}$  inputs for SPEC CPU 2017.





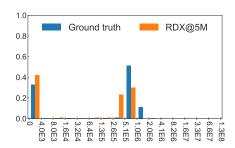


Figure 15: Stack reuse histogram of bwaves.

Figure 16: Stack reuse histogram of leslie3d.

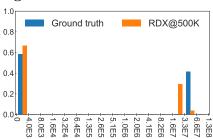


Figure 17: Stack reuse histogram of sphinx3.

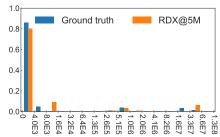


Figure 18: Stack reuse histogram of GemsFDTDs.

Figure 19: Stack reuse histogram of 1bm.

Figure 20: Stack reuse histogram of cactus ADM.

SSN\_s (Figure 65), 1bm\_s (Figure 66), omnetpp\_s (Figure 67), wrf\_s (Figure 68), xalancbmk\_s (Figure 69), x264\_s (Figure 70, Figure 71 and Figure 72), cam4\_s (Figure 73), pop2\_s (Figure 74), deepsjeng\_s (Figure 75), imagick\_s (Figure 76), leela\_s (Figure 77), nab\_s (Figure 78), exchange2\_s (Figure 79), fotonik3d\_s (Figure 80), roms\_s (Figure 81), xz\_s (Figure 82 and Figure 83) from SPEC CPU2017.

## E. CASE STUDY: OPTIMIZING LULESH

LULESH [4], a UHPC application benchmark developed by Lawrence Livermore National Laboratory (LLNL), solves a simple Sedov blast problem with analytic answers. We launch it with 32 threads monitored by RDX and Figure 84 shows our graphic user interface highlighting the locations with high latency. The interface consists of three panels. The top one displays the source code; the bottom left one shows the application contexts with functions, loops and statements; the bottom right panel displays metrics related to the contexts.

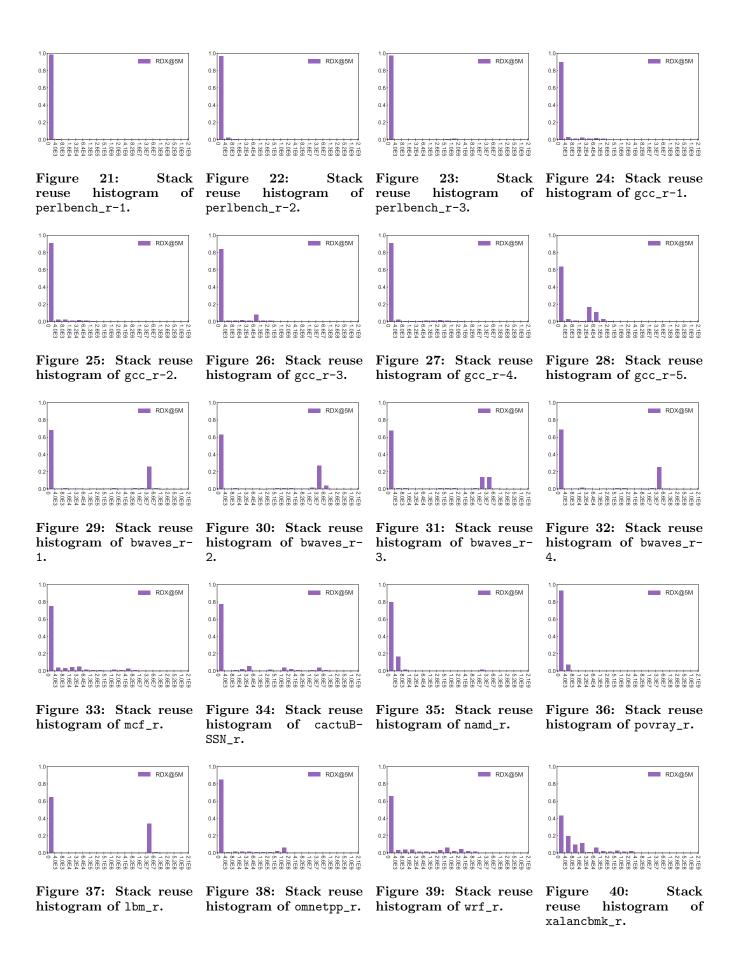
We start to look for the locations with high latency. A loop region (line:1284) accounts for 7.9% of the total latency as shown in Figure 84. With the knowledge of the problematic location, we correlate all the reuse pairs with this region. We find that the application stores the arrays x8n (line:1525), y8n (line:1526) and z8n (line:1527) in the loop (line:1509) and later uses them many times in the loop (line:1284) inside the function CalcFBHourglassForceForElems. Figure 85 shows a reuse pair of the variable z8n. The time reuse distance is  $\sim 4 \times 10^7$  for x8n, y8n and z8n and RDX reports high access latencies for these arrays.

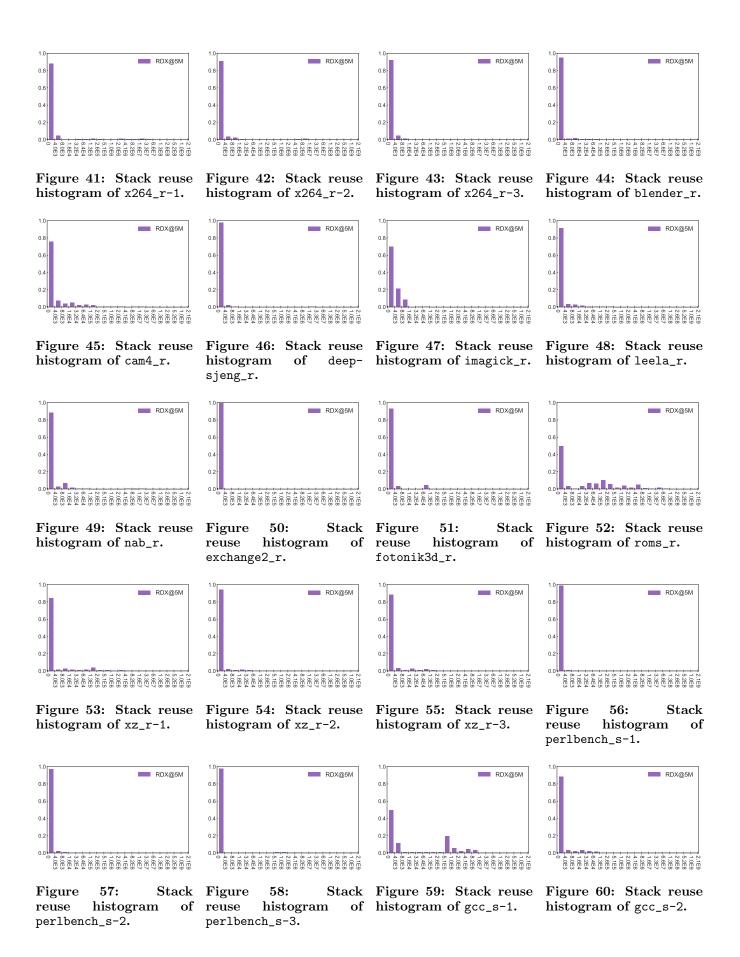
To enhance the temporal locality, we fused the two loops (line:1509 and 1284). The fusion is straightfor-

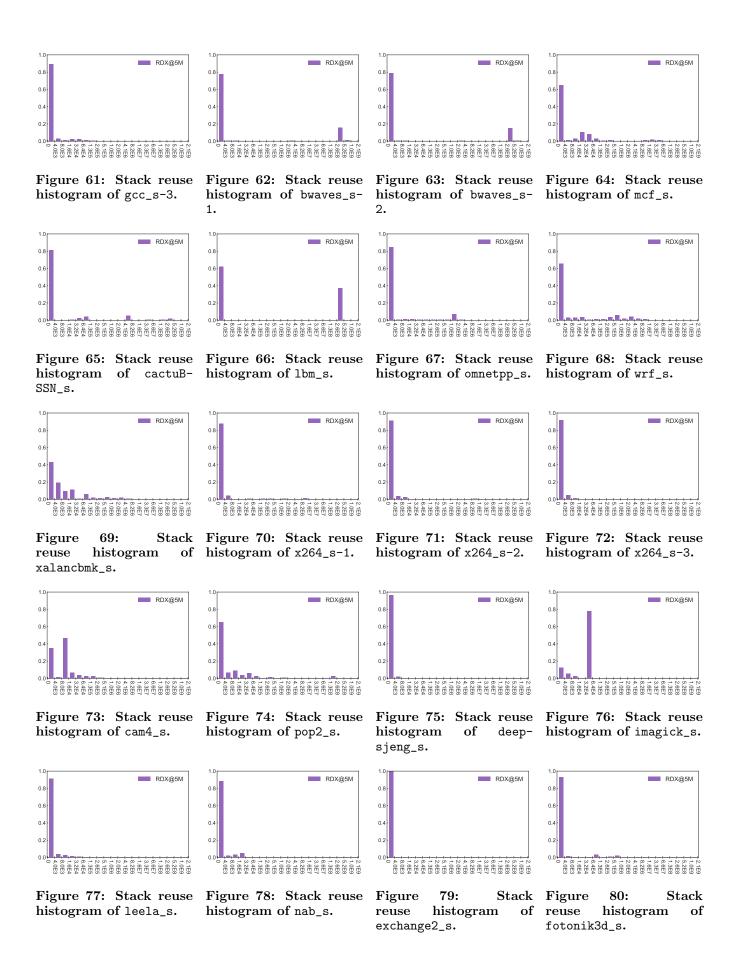
ward and safe: they share the same loop bounds and stride and do not change the dependence direction after fusing. The fusion optimization achieves a  $1.54 \times$  speedup.

# F. COMPARISON WITH THREADSPOTTER

ThreadSpotter introduces a large overhead (GeoMean  $\sim 17\times$ ) shown in Figure 86. Furthermore, the time reuse accuracy of ThreadSpotter is always lower than RDX as shown in Figure 87. In certain cases, ThreadSpotter has significantly low accuracy, as low as 37% for astar. Figure 88 shows the histogram of astar, where ThreadSpotter incorrectly attributes more on the longer reuses and misses many short-distance reuses.







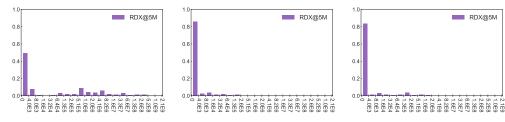


Figure 81: Stack reuse histogram of roms\_s.

Figure 82: Stack reuse histogram of xz\_s-1.

Figure 83: Stack reuse histogram of xz\_s-2.

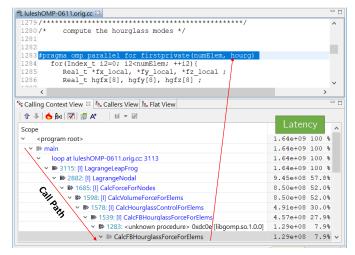


Figure 84: Locate a loop region with high latency in LULESH.

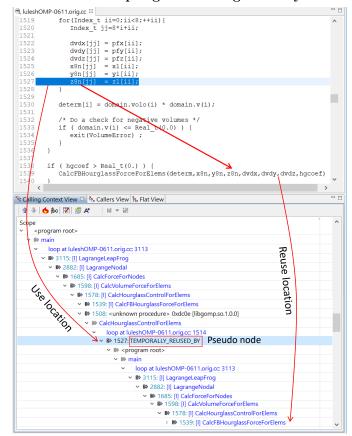


Figure 85: One reuse pair of the variable z8n is related to the problematic region shown in Figure 84.

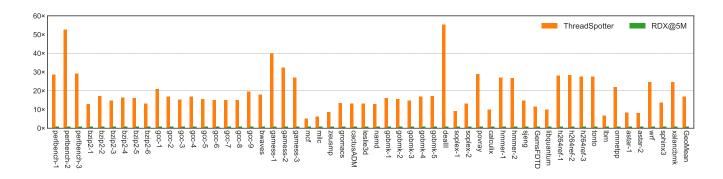


Figure 86: Time overhead comparison between ThreadSpotter (running in default sampling rate) and RDX for SPEC CPU2006 benchmarks. One benchmark may have multiple lines due to runs with different inputs.

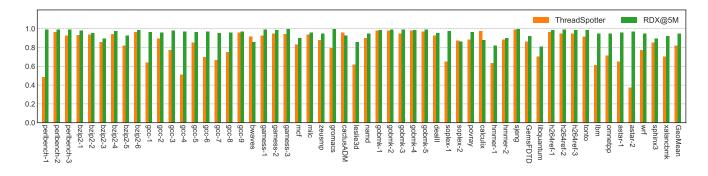


Figure 87: Time reuse accuracy comparison between ThreadSpotter (running in default sampling rate) and RDX for SPEC CPU2006 benchmarks. One benchmark may have multiple lines due to runs with different inputs.

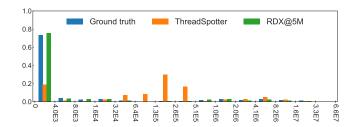


Figure 88: Time reuse histogram of astar (input 2) from SPEC CPU2006 benchmark.

## 7. REFERENCES

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