Scalable Data Correlation

Managing TB-scale investigations with similarity digests



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The Current Forensic Workflow



* http://accessdata.com/distributed-processing

Scalable Forensic Workflow



→ We can start working on the case *immediately*.

Current Forensic Processing

- > Hashing/filtering/correlation
- File carving/reconstruction
- > Indexing

The ultimate goal of this work is to make similarity hash-based correlation scalable & I/O-bound.

Motivation for similarity approach: Traditional hash filtering is failing

> Known file filtering:

- Crypto-hash known files, store in library (e.g. NSRL)
- Hash files on target
- Filter in/out depending on interest

> Challenges

- Static libraries are falling behind
 - Dynamic software updates, trivial artifact transformations
 - → We need **version** correlation
- \circ Need to find embedded objects
 - Block/file in file/volume/network trace
- Need higher-level correlations
 - Disk-to-RAM
 - Disk-to-network



Given a fragment, identify source

- Fragments of interest are 1-4KB in size
- Fragment *alignment is arbitrary*

Scenario #2: Artifact Similarity





Similar files (shared content/format)

Similar drives (shared blocks/files)

Given two binary objects, detect similarity/versioning

Similarity here is purely syntactic;

• Relies on commonality of the binary representations.

Solution: Similarity Digests



All correlations based on bitstream commonality

Quick Review: Similarity digests & sdhash

Generating sdhash fingerprints (1)

Digital artifact

(block/file/packet/volume) as byte stream



Generating sdhash fingerprints (2)

Digital artifact



Select characteristic features (statistically improbable/rare)

Generating schash fingerprints (3) Feature Selection Process





Generating sdhash fingerprints (4)



Sequence of Bloom filters (sdbf)

Bloom filter

local SD fingerprint
256 bytes
up to 128/160 features



Based on BF theory, overlap due to chance is analytically predictable.

Additional BF overlap is proportional to overlap in features. BF_{Score} is tuned such that $BF_{Score}(A_{random}, B_{random}) = 0$.

SDBF fingerprint comparison



Scaling up: Block-aligned digests & parallelization

Block-aligned similarity digests (sdbf-dd)



Sequence of Bloom filters (sdbf-dd)

Bloom filter

Iocal SD fingerprint

□ 256 bytes

up to 192 features

Advantages & challenges for blockaligned similarity digests (sdbf-dd)

> Advantages

- Parallelizable computation
- Direct mapping to source data
- Shorter (1.6% vs 2.6% of source)
- → Faster comparisons (fewer BFs)

> Challenges

- Less reliable for smaller files
- Sparse data
- Compatibility with sdbf digests

Solution

- Increase features for sdbf filters: $128 \rightarrow 160$
- Use 192 features per BF for sdbf-dd filters
- Use compatible BF parameters to allow sdbf ⇔ sdbf-dd comparisons

sdhash 1.6: sdbf vs. sdbf-dd accuracy

Query size	FP rate	TP rate	Query size	FP rate	TP rate
1,000	0.1906	1.000	2,000	0.0006	0.997
1,100	0.0964	1.000	2,200	0.0005	1.000
1,200	0.0465	1.000	2,400	0.0001	1.000
1,300	0.0190	1.000	2,600	0.0001	0.997
1,400	0.0098	1.000	2,800	0.0000	1.000
1,500	0.0058	1.000	3,000	0.0000	0.999
1,600	0.0029	0.999	3,200	0.0000	0.998
1,700	0.0023	0.999	3,400	0.0000	0.998
1,800	0.0013	0.999	3,600	0.0000	1.000
1,900	0.0010	0.998	3,800	0.0000	0.998

Sequential throughput: sdhash 1.3

> Hash generation rate

- Six-core Intel Xeon X5670 @ 2.93GHz
 ~27MB/s per core
- Quad-Core Intel Xeon @ 2.8 GHz
 ~20MB/s per core
- Hash comparison
 - 1MB vs. 1MB: 0.5ms
- > T5 corpus (4,457 files, all pairs)
 - \circ 10mln file comparisons in ~ 15min
 - 667K file comps per second
 - Single core

sdhash 1.6: File-parallel generation rates on 27GB real data (in RAM)

Threads	$\mathbf{Time}\;(\mathrm{sec})$	Throughput (MB/s)	Speedup
1	920	29.08	1.00
4	277	96.57	3.32
8	187	143.05	4.92
12	144	185.76	6.39
24	129	207.36	7.13

sdhash 1.6: <u>Optimal</u> file-parallel generation: 5GB synthetic target (RAM)

Threads	$\mathbf{Time}\;(\mathrm{sec})$	Throughput (MB/s)	Speedup
1	177	28.25	1.00
4	50	100.00	3.54
8	30	166.67	5.90
12	24	208.33	7.38
24	19	263.16	9.32

sdhash-dd: Hash generation rates 10GB in-RAM target (RAM)

Threads	$\mathbf{Time}\;(\mathrm{sec})$	Throughput (MB/s)	Speedup
1	374.0	26.74	1.00
4	93.0	107.53	4.02
8	53.0	188.68	7.06
12	44.5	224.72	8.40
24	27.0	370.37	13.85

Throughput summary: sdhash 1.6

Parallel hash generation

- sdbf: file-parallel execution
 - 260 MB/s on 12-core/24-threaded machine
- sdbf-dd: block-parallel execution
 - 370 MB/s (SHA1 —> 330MB/s)
- Optimized hash comparison rates
 24 threads: 86.6 mln BF/s
 1.4 TB/s for small file comparison (<16KB)
 I.e., we can search for a small file in a reference set of 1.4TB in 1s

The Envisioned Architecture



The Current State



Todo List (1)

> libsdbf

- C++ rewrite (v2.0)
- TBB parallelization

> sdhash-file

- More command line options/compatibility w/ssdeep
- Service-based processing (w/ *sdbf_d*)

> GPU acceleration

- sdhash-pcap
 - Pcap-aware processing:
 - payload extraction, file discovery, timelining

Todo List (2)

> sdbf_d

- Persistance: XML
- Service interface: JSON
- Server clustering

> sdbfWeb

Browser-based management/query

≻ sdbfViz

• Large-scale visualization & clustering

Further Development

Integration w/ RDS

- *sdhash-set*: construct *SDBF*s from existing SHA1 sets
 - Compare/identify whole folders, distributions, etc.

Structural feature selection

- E.g., exe/dll, pdf, zip, ...
- Optimizations
 - Sampling
 - o Skipping
 - Under min continuous block assumption
 - Cluster "core" extraction/comparison

Representation

- Multi-resolution digests
- New crypto hashes
- Data offsets

Thank you!

http://roussev.net/sdhash

- wget <u>http://roussev.net/sdhash/sdhash-1.6.zip</u>
- o make
- ./sdhash
- Contact:

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> Reminder

DFRWS'12: Washington DC, Aug 6-8 Paper deadline: Feb 20, 2012 Data sniffing challenge to be released shortly