

Open Source Similarity Digests DFRWS August 2016

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Notes on this class



- Discuss how Similarity Digests work
 - Ssdeep
 - TLSH
 - Sdhash
 - Nilsimsa
- Have practice using them
- Cover important cases

The notes include slides labeled FULL DETAILS – we will not go over these in the class

Acknowledgement:

Work done with Scott Forman, Chun Cheng, Yanggui Chen & Vic Hargrave Also thanks to Jennifer Rihn for notes on the class



What are Similarity Digests?



- Traditional hashes (such as SHA1 and MD5) have the property that a small change to the file being hashed results in a completely different hash
- Similarity Digests and Locality Sensitive Hashes (LSH) have the property that a small change to the file being hashed results in a small change to the hash
 - You can measure the similarity between 2 files by comparing their digests



Similarity Digests



- Similar files / images / documents
 - Spam / attachments
 - Malware families
- Does not solve
 - Packing issues
 - Encryption
 - Compression (zip files, jpg, gif, etc)
 - Encoding
- Use Security / Forensic knowledge to extract the required content
 - Then use Similarity Digests



Similarity Digests



4 pieces of open source software

- 1. Ssdeep is the Industry standard (in Virus Total and NIST)
- 2.TLSH is Trend Micro's LSH
 - Less vulnerable to attack
 - Enables fast search

3.Sdhash

Literature says the Sdhash is better than Ssdeep

4. Nilsimsa

Proposed for spam signatures



Licenses



- 1. Ssdeep GPL
- 2. TLSH Apache
- 3. Sdhash Apache
- 4. Nilsimsa Various

The Apache license – an important detail.

Variants must include NOTICE.txt



NOTICE.txt



- == NOTICE file for use with the Apache License, Version 2.0, ==
- == in this case for the Trend Locality Sensitive Hash distribution. ==

Trend Locality Sensitive Hash (TLSH)
Copyright 2010-2016 Trend Micro

This product includes software developed at Trend Micro (http://www.trendmicro.com/)

Refer to the following publications for more information:

Jonathan Oliver, Chun Cheng and Yanggui Chen,
"TLSH - A Locality Sensitive Hash"
4th Cybercrime and Trustworthy Computing Workshop, Sydney, November 2013
https://github.com/trendmicro/tlsh/blob/master/TLSH_CTC_final.pdf

Jonathan Oliver, Scott Forman and Chun Cheng,
"Using Randomization to Attack Similarity Digests"

Applications and Techniques in Information Security. Springer Berlin Heidelberg, 2014. 199-210.

https://github.com/trendmicro/tlsh/blob/master/Attacking_LSH_and_Sim_Dig.pdf



Log into AWS



If you use Cygwin

- (1) Put instance1.pem into some folder sim digest
- (2) In shell / Cygwin
 - \$ cd sim digest
 - \$ ssh -i instance1.pem ec2-user@ec2-a-b-c-d-201.ap-southeast-2.compute.amazonaws.com

Where a-b-c-d is replaced with your allocated IP # Do not use "." use "-" in between the numbers

- (3) In AWS
 - \$./alloc.sh YOUR NAME
 - \$ cd Similarity_Digest_YOUR_NAME



Exercise 1A: Calculating Digests



chp1.txt - Chapter 1 of Pride and Prejudice

It is a truth universally acknowledged, that a single man in possession of a good fortune, must be in want of a wife.

. .

When she was discontented, she fancied herself nervous. The business of her life was to get her daughters married; its solace was visiting and news.

- \$ cd Exercise1
- \$./exercise1A.sh



Exercise 1B: Comparing Files



chp1.txt - Chapter 1 of Pride and Prejudice

It is a truth universally acknowledged, that a single man in possession of a good fortune, must be in want of a wife.

. .

When she was discontented, she fancied herself nervous.

The business of her life was to get her daughters married; its solace was visiting and news.

chp1-.txt - Chapter 1 of Pride and Prejudice with last line removed

It is a truth universally acknowledged, that a single man in possession of a good fortune, must be in want of a wife.

. .

When she was discontented, she fancied herself nervous.



Exercise 1B: Comparing Files Trend University is to Train

- \$./exercise1B.sh
- ../bin/tlsh -f chp1.txt -c chp1-.txt chp1.txt
- ../bin/sdhash -q chp1.txt chp1-.txt chp1.txt|chp1-.txt | 100
- ../bin/ssdeep -d -l chp1.txt chp1-.txt chpl-.txt matches chpl.txt (100)
- ../bin/nilsimsa ut -v1 -f chp1.txt -c chp1-.txt chp1.txt



Exercise 1B: Comparing Files Trend University is to Train

- \$./exercise1B.sh random.txt
- ../bin/tlsh -f chp1.txt -c random.txt 324 chp1.txt
- ../bin/sdhash -t -1 -g chp1.txt random.txt chp1.txt|random.txt|000
- ../bin/ssdeep -a -d -l chp1.txt random.txt random.txt matches chp1.txt (0)
- ../bin/nilsimsa ut -v1 -f chp1.txt -c random.txt chp1.txt



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Score Ranges

TLSH: distance score

0 perfect match

1.. 100 near perfect (1) to weak match (100)

2000 very distant files

Ssdeep: similar score

Sdhash: similarity score

0 no match

1.. 99 match

100 perfect match

Nilsimsa: similarity score

D perfect disagreement

no similarity

perfect match



Exercise 1C

Copy chp1.txt to 1c.txt

```
$ cp chp1.txt 1c.txt
```

Do a small change to 1c.txt

```
$ vi 1c.txt
    (if you do not like vi - then use nano)
$ ./exercise1B.sh 1c.txt
```

Exercise: Modify 1c.txt so that

```
TLSH (chp1.txt, 1c.txt) = 1 \text{ or } 2
```

Ssdeep(chp1.txt, 1c.txt) =
$$99 \text{ or } 100$$

Sdhash(chp1.txt, 1c.txt) =
$$99 \text{ or } 100$$

Nilsimsa(chp1.txt,
$$1c.txt$$
) = 255 or 256



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Exercise 1D

Simple transforms

What will some standard transformations do?

sort

fmt

rot13 a->n b->o c->p ... z->m

lowercase A->a B->b ...

\$./exercise1D.sh



Exercise 1E



Simple encodings

Encode a file and a close variant.

File1 => base64 file1.base64

File1+ => base64 file1+.base64

\$./exercise1E.sh



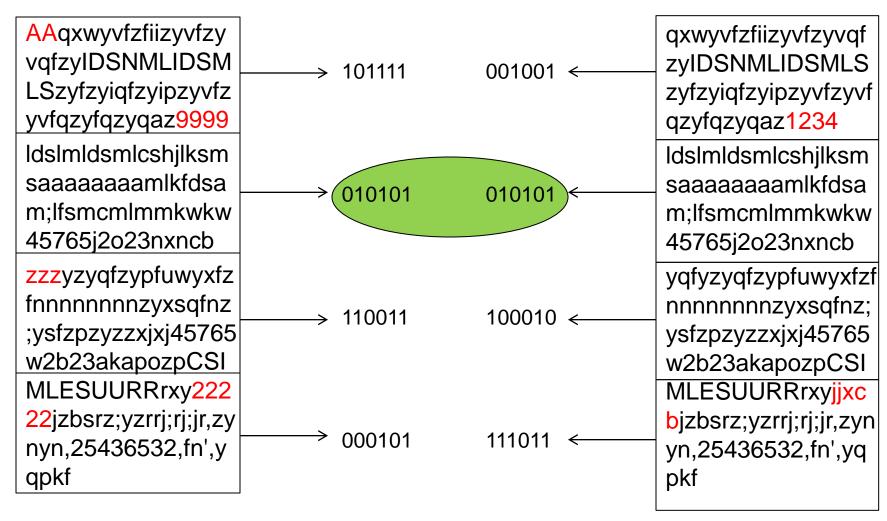




	Context Triggered Piecewise Hashing	Feature Extraction	Locality Sensitive Hashes
Example	Ssdeep	Sdhash	Nilsimsa, TLSH
Creating the digest	Cut up a file into segments Create a checksum for each segment The digest is the concatenation of the checksums	Extract relatively long features (64 bytes) which are "interesting" The digest is the encoded features	Extract many very small features (3 bytes) Put the features into a histogram The digest is the encoded histogram
Matching Digests	If enough checksums match – then the files match	If enough encoded features match – then the files match	Score the distance between the histograms

Ssdeep







Locality Sensitive Hashes (Nilsimsa, TLSH)

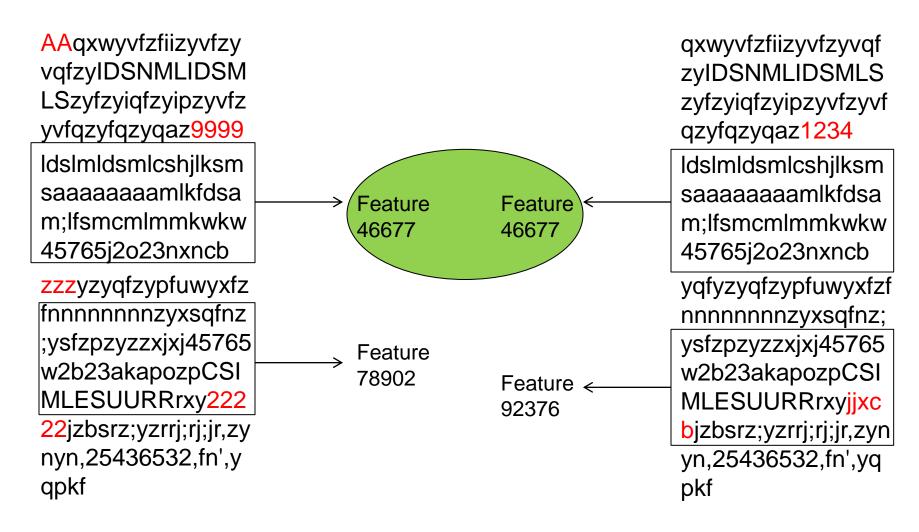


AAqxwyvfzfiizyvfzy qxwyvfzfiizyvfzyvqf vqfzvIDSNMLIDSM ZVIDSNMLIDSMLS LSzyfzyiqfzyipzyvfz zyfzyiqfzyipzyvfzyvf **Bucket** Bucket ∠ yvfqzyfqzyqaz9999 qzyfqzyqaz1234 56 IdsImIdsmlcshilksm IdsImIdsmlcshjlksm saaaaaaamlkfdsa saaaaaaamlkfdsa m:Ifsmcmlmmkwkw m:lfsmcmlmmkwkw 45765j2023nxncb 45765j2o23nxncb zzzyzyqfzypfuwyxfz yqfyzyqfzypfuwyxfzf **Bucket Bucket** fnnnnnnnzyxsqfnz nnnnnnnzyxsqfnz; 89 89 ;ysfzpzyzzxjxj457)65 ysfzpzyzzxjxj45765 w2b23akapozpCSI w2b23akapozpCSI MLESUURRrxy222 **MLESUURRrxyjixc** 22jzbsrz;yzrrj;rj;jr,zy bjzbsrz;yzrrj;rj;jr,zyn nyn,25436532,fn',y yn,25436532,fn',yq qpkf pkf



Sdhash (feature extraction)







Processing Directories



We have set up for you 8 commands

Lists the digests for a directory of files nil_list DIR tlsh_list DIR ssdeep_list DIR sdhash list DIR

Does a scoring comparison for every pair of files in a directory nil_score DIR tlsh_score DIR ssdeep_score DIR sdhash_score DIR



Processing Directories



```
alias nil_list="/home/ec2-user/bin/nilsimsa_ut -v1 -r"
alias tlsh_list="/home/ec2-user/bin/tlsh -r"
alias ssdeep_list="/home/ec2-user/bin/ssdeep -r -l"
alias sdhash_list="/home/ec2-user/bin/sdhash -r"

alias nil_score="/home/ec2-user/bin/nilsimsa_ut -xref -v1 -r"
alias tlsh_score="/home/ec2-user/bin/tlsh -xref -r"
alias ssdeep_score="/home/ec2-user/bin/ssdeep -r -l -d -a"
alias sdhash score="/home/ec2-user/bin/sdhash -r -g -t -999"
```







Folder Name	Manipulation	Image 1	Image 2	
Angled	Image rotation	DISCOUNTED PHARMA BUY BULK AND SAVE 80%!!! WE SHIP WORLDWIDE! OUR SITE VERIFIED BY VISA! CLICK HERE	DISCOUNTED PHARMA BUY BULK AND SAVE 80%!!! WE SHIP WORLDWIDE! OUR SITE VERIFIED BY VISA!	

Pharmacy erectile Changing image height and width; Changing background colour





Pharmacy_Move Changing image height and width; Adding dots, and dashes.



Working with Image Files



Exercise2/Images_sorted_1000/Pharmacy_cialis_softtabs







0e6f3429_0.gif



0f05d804_0.gif



03303bb4_0.gif

084a03d7_0.gif

01047e2a_0.gif



Exercise 2A Working with Image Files



```
$ cd Exercise2
```

Use the commands

nil_list, tlsh_list, ssdeep_list, sdhash_list
nil_score, tlsh_score, ssdeep_score, sdhash_score
to inspect the digests and similarity scores of
Images_sorted_1000/Pharmacy_cialis_softtabs

Can the digests determine that the images are similar?



Limitation of Similarity Digests

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Similarity Digests cannot identify files as being similar if they are

- -Encrypted
- -Compressed
- -Packed malware
- -Encoded

-...

You have to unpack, un-compress or decrypt first.

What do we need to do with gif, jpeg files?



Exercise 2B Working with Image Files



Each of the image files has been converted (using CxImage library) to a .bmp file (a image bitmap – identical to the in memory representation of images)

.bmp files can be compared

Use *_list, *_score commands to inspect the digests and similarity scores of Images_sorted_1000/Pharmacy_cialis_softtabs_bmp



Exercise 2B Working with Image Files



```
Select a folder of images.
      Which method(s) works on that folder?
      Which method(s) fails on that folder?
If you are having problems with sdhash - you might have to
      $ export LC ALL="en US.UTF-8"
A.Angled bmp/
B. Pharmacy erectile dys bmp/
C.Pharmacy Move 2col bmp/
$ cd Exercise2
$ tlsh score
                 Images sorted 1000/Angled bmp
$ ssdeep score
                 Images sorted 1000/Angled bmp
$ sdhash score
                 Images sorted 1000/Angled bmp
```



Exercise 2C Working with Image Files



Use the digests to work out what type of images are in Random_Images/

```
$ cd Exercise2
```

\$./exercise2C.sh

ssdeep

Usage: ssdeep [-m file] [FILES]

-m - Match FILES against known hashes in file

tlsh

Usage: tlsh -l listdigests -c file



Exercise 2C (cont) Working with Image Files



Need a hint?

```
$ cd Exercise2
$ cp ../Answers/answer2C.sh .
$ ./answer2C.sh | less
```



Collisions / False Positive Matches



Collision

When 2 distinct files have the same digest or hash

False Positive Match

When the score is a match, but we consider file1 not similar to file2

Ssdeep(file1, file2) > 0

Sdhash(file1, file2) > 0

TLSH(file1, file2) <= 100

Nilsimsa(file1, file2) >= 220



Collisions / False Positive Matches



Exercise 2D

Do any of the methods suffer from collisions in the collection of image files?

Find one.

Exercise 2E

Do any of the methods suffer from false positive matches in the collection of image files?

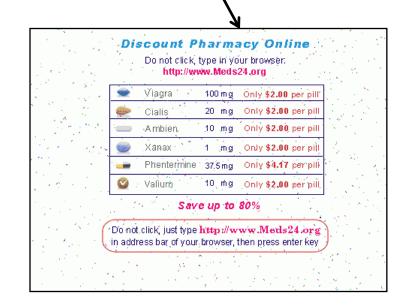
Find one.



Exercise 2D Nilsimsa Collision









Exercise 2E TLSH False Positives



```
../bin/tlsh
```

- -c Images_sorted_1000/Pharmacy_Viagra_Pro_bmp/01b2bb87_0.bmp
- -f Images_sorted_1000/Pharmacy_power_pack_bmp/01a07331_0.bmp

78 Images sorted 1000/Pharmacy power pack bmp/01a07331 0 bmp

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Boost Sexual Performance

Fuller_Harder Erections

Increase Stamina Endurance

Quicker Recharges

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Comparison on Image files



Number images	<u>TLSH (100)</u>	Sdhash(1)		<u>Nilsimsa (240 /</u> <u>256)</u>
20	80.0%	3.7%	0.0%	100.0%
3	33.3%	33.3%	0.0%	100.0%
11				
63	12.1%	11.2%	1.0%	61.2%
10	64.4%	62.2%	4.4%	64.4%
6	100.0%	100.0%	6.7%	100.0%
8	57.1%	3.6%	7.1%	57.1%
ľ	20 3 11 5 147 22 22 63 10 6	images TLSH (100) 20 80.0% 3 33.3% 11 100.0% 5 100.0% 147 22.1% 22 0.0% 22 90.5% 63 12.1% 10 64.4% 6 100.0%	images TLSH (100) Sdhash(1) 20 80.0% 3.7% 3 33.3% 33.3% 11 100.0% 100.0% 5 100.0% 100.0% 22 0.0% 0.0% 22 90.5% 100.0% 63 12.1% 11.2% 10 64.4% 62.2% 6 100.0% 100.0%	images TLSH (100) Sdhash(1) Ssdeep(1) 20 80.0% 3.7% 0.0% 3 33.3% 33.3% 0.0% 11 100.0% 100.0% 100.0% 5 100.0% 100.0% 9.6% 22 0.0% 0.0% 0.0% 22 90.5% 100.0% 10.8% 63 12.1% 11.2% 1.0% 10 64.4% 62.2% 4.4% 6 100.0% 100.0% 6.7%

Comparison on Image files

911

False Positives



0.0%

4.6%

<u>Folder Name</u>	Number images	TLSH (100)	Sdhash(1)		Nilsimsa (240 / 256)
Pharmacy pop a pill	5	80.0%	100.0%	60.0%	100.0%
Pharmacy power pack	41	47.8%	47.8%	20.7%	55.9%
Pharmacy research	3	0.0%	33.3%	33.3%	100.0%
Pharmacy Viagra Pro	11	. 32.7%	38.2%	29.1%	54.5%
Pharmacy Viagra Pro2	7	42.9%	42.9%	42.9%	42.9%
Software OEM	6	66.7%	66.7%	66.7%	66.7%
Software SOBAKA	11	. 100.0%	100.0%	100.0%	100.0%
StockSpam CYTV	105	1.7%	1.4%	0.0%	27.3%
StockSpam EXVG	389	1.2%	2.8%	0.6%	100.0%

0.007%

0.0%



```
$ cd Exercise3/
$ ./exe match.sh
```

Start with default thresholds:

- TLSH <= 100
- Sdhash >= 1
- Ssdeep >= 1

Exercise 3A

For each method, find a better threshold for a match





In the paper, I suggest for executable files:

TLSH <= 52

Near 52:

- about half the pairs are related,
- · about half the pairs are unrelated
- Sdhash >= 13

Near 13:

- about half the pairs are related,
- about half the pairs are unrelated
- Ssdeep >= 1 (unclear what is happening between 1 -25)





Exercise 3B

```
$ cd Exercise3
$ ./exercise3B.sh
```

There are groups of similar looking executables.

Some groups include

debconf-*

dpkg-*

gnome-*

How effective are the methods at identifying these groups?





Exercise 3C

There are some unexpected matches.

For example, are the executable pairs

uniq wc

du diff

similar?

If so, why?

If not, why not?



Evaluation: Random Changes



- 500 lines of Pride and Prejudice
- 200 different version each more different than the previous
- Random changes
 - inserting a new word
 - ii. deleting an existing word
 - iii. swapping two words
 - iv. substituting a word for another word
 - v. replacing 10 occurrences of a character for another character
 - vi. deleting 10 occurrences of a character



Exercise 4A



pp.0 is the first 500 lines of Pride and Prejudice

\$ cd Exercise4

\$./pp.sh

starting file: pp_changes/pp.0

Iteratively makes 500 random changes creating files pp_changes/pp.1001 pp_changes/pp.1500



Exercise 4A



pp.Oreason gives an explanation of each change

- SWAP-line 251-252 and line 451-452
- DELETE-word 'much' [pos=3882,len=4]
- SUBST-word 'when' [pos=6811,len=4] for 'him' [pos=10012,len=4]

pp.txt gives the score for each iteration compared to the original file

tlsh, sdhash, ssdeep

2,096,97

5,095,94

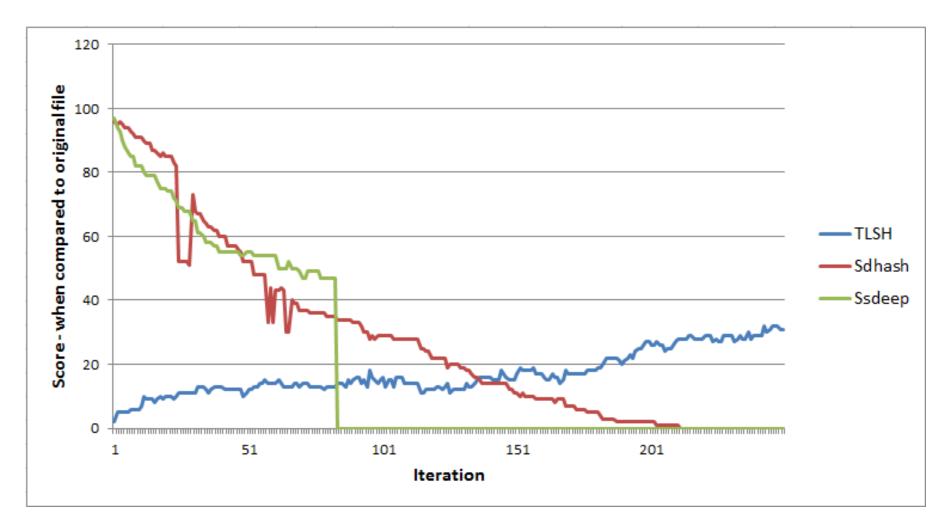
5,096,93

At which iteration does TLSH, Sdhash and Ssdeep break?



Exercise 4A: 500 lines of P&P

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Exercise 4A: 500 lines of P&P



First 500 lines after 200th iteration

"I hope Mr. Bingley will like it, Lizzy."

Author: Jane Austen

Date: August 26, 2008 [EBook #1342]

Release Date: June, 1998

[Last updated: October 12, 2012]

Language: English

*** START OF THIS PROJECT GUTENBERG EBOOK PRIDE AND PREJUDICE ***

- Ssdeep has failed at iteration 84
- Sdhash has failed at iteration 212



Exercise 4B

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Copy pp.0 to pp.4B Modify pp.4B so that

- Ssdeep(pp.0, pp.4B) = 0
- Sdhash(pp.0, pp.4B) = 0
- TLSH (pp.0, pp.4B) > 100

```
$ cd Exercise4
$ mkdir 4B
$ cp pp_changes/pp.0 4B
$ cp pp_changes/pp.0 4B/pp.4B
$ vi 4B/pp.4B
$ tlsh_score 4B
$ sdhash_score 4B
$ ssdeep score 4B
```



Exercise 4C



Create a version of pp.0 which

- Cannot be detected by the similarity digests
- •A human reader would not notice the difference

Copy pp.0 to pp.4C Modify pp.4C so that

- •Ssdeep(pp.0, pp.4C) = 0
- •Sdhash(pp.0, pp.4C) = 0
- •TLSH (pp.0, pp.4C) > 100

```
$ cd Exercise4
```

- \$ mkdir 4C
- \$ cp pp changes/pp.0 4C
- \$ cp pp changes/pp.0 4C/pp.4C
- \$ vi 4C/pp.4C
- \$ tlsh score 4C
- The text is easily readable by a person
- •\$ diff -w 4C/pp.0 4C/pp.4C
 - Produces no output

At this point you have "broken" the digests



Conclusions



- Similarity Digests are a great starting place for quickly finding similar content
 - Might want / need to adapt approaches
 - Similarity digests are a general tool. For specific applications (images) consider specific solutions
- Need to consider thresholds for these hashes
 - Each application may needs its own threshold
- When considering you problem / your application
 - An adversary may be deliberately morphing / obfuscating the file / content
 - Consider attacking your own application
- The different Similarity Digests have different strengths
 - Complex applications may require hybrid approaches



Papers



Introduction to TLSH

Oliver, J., Cheng, C., Chen, Y.: TLSH - A Locality Sensitive Hash. 4th Cybercrime and Trustworthy Computing Workshop, Sydney, November 2013

https://github.com/trendmicro/tlsh/blob/master/TLSH_CTC_final.pdf

Vulnerability Paper

Oliver, J., Forman, S., and Cheng, C.: Using Randomization to Attack Similarity Digests. ATIS 2014, November, 2014, pages 199-210.

https://github.com/trendmicro/tlsh/blob/master/Attacking_LSH_and_Sim_Dig.pdf

Open sources on Github

https://github.com/trendmicro/tlsh/



Papers



SdHash

"Data fingerprinting with similarity digests"

Vassil Roussev

Sixth IFIP WG 11.9 International Conference on Digital Forensics, Hong Kong, China, January 4-6, 2010

http://roussev.net/pdf/2010-IFIP--sdhash-design.pdf

<u>Ssdeep</u>

"Identifying almost identical files using context triggered piecewise hashing" Jesse Kornblum

Journal Digital Investigation: The International Journal of Digital Forensics & Incident Response archive

Volume 3, September, 2006 Pages 91-97

http://dfrws.org/2006/proceedings/12-Kornblum.pdf

Source code for SSDEEP: http://ssdeep.sourceforge.net/



Papers (cont.)



Nilsimsa

Source code for Nilsimsa http://ixazon.dynip.com/~cmeclax/nilsimsa.html

"An open digest-based technique for spam detection"

E. Damiani1, S. De Capitani di Vimercati1, S. Paraboschi2, P. Samarati

Proceedings of the 2004 international workshop on security in parallel and distributed systems. 2004.

Comparison Paper

http://spdp.di.unimi.it/papers/pdcs04.pdf

"An evaluation of forensic similarity hashes"

Vassil Roussev

Journal Digital Investigation: The International Journal of Digital Forensics & Incident Response archive Volume 8, August, 2011

Pages S34-S41



End Session



Thank you



FULL DETAILS Ssdeep



- 1. Use a rolling hash to split the document into segments,
- 2. Produce a 6 bit value for each segment by hashing the segment,
- 3. Concatenate the base64 encoded 6 bit values from step (2) to form the output signature.

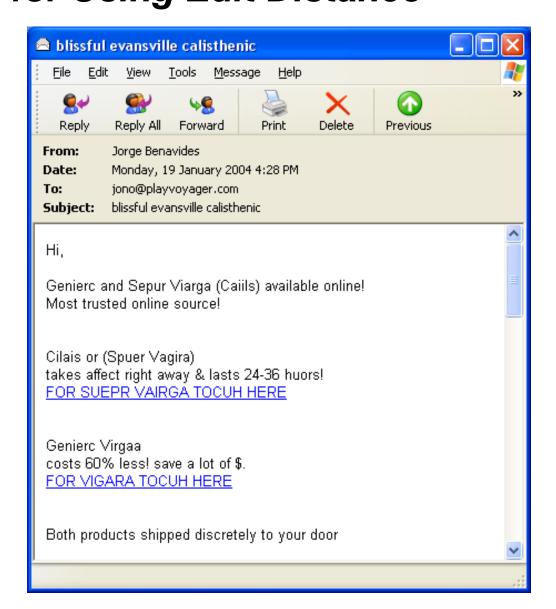
Similarity(digest1, digest2) = 100 - edit distance(digest1 and digest2)

Calculate the probability of the two strings being aligned using Lloyd Allison's Dynamic Programming Algorithm (DPA)

Ref: Dynamic Programming Algorithm for Sequence Alignment http://www.csse.monash.edu.au/~lloyd/tildeStrings/Notes/DPA/



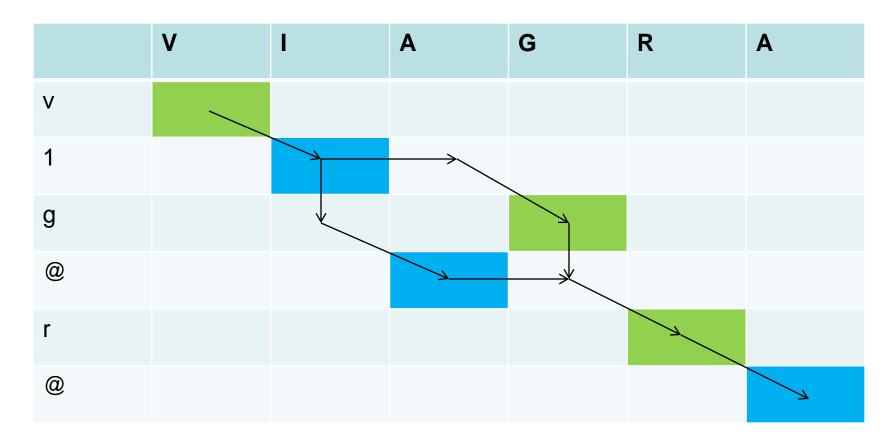
FULL DETAILS Motivation for Using Edit Distance TREND UNIVERSIT





FULL DETAILS Calculating Edit Distance





Ref: "Using Lexigraphical Distancing to Block Spam" Jonathan Oliver, MIT Spam Conference 2005.





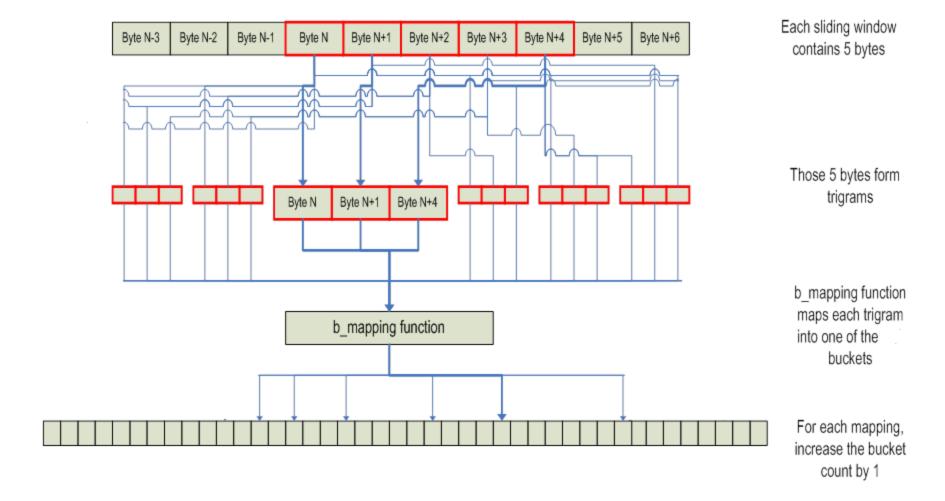
Deeper Explanation of How TLSH works



FULL DETAILS Algorithm to determine TLSH

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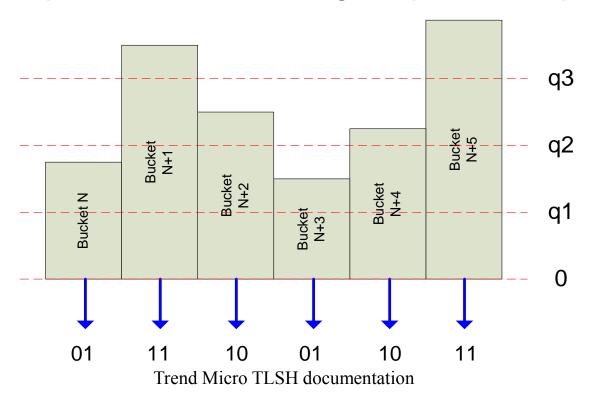




FULL DETAILS Algorithm to determine TLSH



- TLSH uses a 4-way to reflect the differences between different histograms
- The q2 point is at the median bucket count
- The q1 and q3 are the lower and higher quartiles respectively





FULL DETAILS The Distance Function



- To calculate the TLSH distance score we iterate through the buckets, scoring the distance between each bucket value
- For digest1 and digest2 we have bucket values in the range 0 .. 3



How EXACTLY does TLSH work?



• Login to AWS

```
$ cd Similarity_Digests_YOURNAME
$ ./bin/tlsh_unittest_verbose -f Exercise1/chp1.txt |
less
```



How EXACTLY does TLSH work?



Chp1.txt

It is a truth universally acknowledged, that a single man in possession

Append BF BB EF to first window

Append BF BB to second window

WINDOW A B C D E

I t i s

WIN1 E D C

WIN2 E D B

WIN3 E C B

WIN4 E C A

WIN5 E D A

WIN6 E B A



How EXACTLY does TLSH work?



- low quartile=90
- median=103
- high quartile=118

•	bucket[0]=87	=>	Emit	00
---	--------------	----	------	----

• bucket[1]=138
$$\Rightarrow$$
 Emit 11 1100 = C

• bucket[3]=148
$$\Rightarrow$$
 Emit 11 1101 = D

bucket[4]=109
$$\Rightarrow$$
 Emit 10

• bucket[7]=110
$$\Rightarrow$$
 Emit 10 1001 = 9

Header

C991C7 1FA380036685B052B9761E3E17F706C1381764C635981FA12A3332EAAC6F96DC



Experiments





Experiments



- Mismatch file set
 - 109 binary malware files (different malware families)
 - 290 randomly constructed HTML fragments
 - 100 pieces of random text from dictionary (no overlap)
 - 79 distinct text files about different topics
- Match file set
 - 3 malware families 20 files each family
- Random created 15 variants of each of the 79 distinct text files
 - 8766 matched file comparisons
 - 55822 different file comparisons



Experiments

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TLSH		Nilsimsa		Sdhash		Ssdeep					
Score	FP rate	Detect	Score	FP rate	Detect	Score	FP rate	Detect	Score	FP rate	Detect
< 300	79.30%	98.8%	> 120	99.86%	100.0%	> 0	0.04711%	37.1%	> 0	0.09966%	31.2%
< 250	69.06%	98.8%	> 130	99.20%	100.0%	> 5	0.02718%	36.6%	> 5	0.09785%	31.2%
< 200	50.10%	98.8%	> 140	98.11%	100.0%	> 10	0.02174%	36.1%	> 10	0.09603%	31.2%
< 150	24.33%	98.1%	> 150	96.98%	100.0%	> 20	0.01812%	35.4%	> 20	0.09422%	31.2%
< 100	6.43%	94.5%	> 160	94.26%	100.0%	> 30	0.01268%	34.4%	> 30	0.05617%	30.9%
< 90	4.49%	92.3%	> 170	89.52%	100.0%	> 40	0.00544%	32.7%	> 40	0.01812%	29.3%
< 80	2.93%	89.0%	> 180	81.38%	100.0%	> 50	0.00362%	29.7%	> 50	0.00362%	27.3%
< 70	1.84%	83.6%	> 190	69.69%	99.7%	> 60	0.00362%	26.0%	> 60	0.00362%	25.9%
< 60	1.09%	76.0%	> 200	54.45%	98.8%	> 70	0.00181%	18.8%	> 70	0.00181%	23.1%
< 50	0.52%	65.3%	> 210	36.73%	96.4%	> 80	0.00181%	12.4%	> 80	0.00000%	16.2%
< 40	0.07%	49.6%	> 220	18.29%	91.9%	> 90	0.00181%	4.6%	> 90	0.00000%	8.8%
< 30	0.00181%	32.2%	> 230	5.52%	72.0%	> 99	0.00000%	1.0%	> 99	0.00000%	3.5%
< 20 0.00181% 17.3%		> 240	1.26%	35.2%							
< 10 0.00181% 6.4% > 2				0.49%	9.5%						