Organizing and analyzing logdata with entropy

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Outline



Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

Data organization

- Trying to define the browsing problem
- Entropy
- Measuring co-dependence
- Mutual Information
- The tree building algorithm

3 Examples

What is this about?

Why?

- To design a better interface for browsing logs & packets
- A smarter interface that reacts to statistical properties of the data.
 - Show "anomalies" first
 - Show off correlations and where they break

How?

Design the browsing interface around

- Trees: natural for decision & classification
- Basic statistics for frequency distribution and correlation
 - Entropy, conditional entropy, mutual information, ...

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How it started

- My wife ran a Tor node (kudos to Roger)
- Kept getting frantic messages from admins:

Your machine is compromised! There is IRC traffic! (IRC=evil)

- OK, but how would we really check if there is something besides the "normal" Tor mix?
- Ethereal isn't much help: how many page-long filters can you juggle?
- Wanted a tool that made classification simple.

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Disclaimer

- These are really simple tricks.
- Ont a survey of research literature (but see last slides).
 - You can do much cooler stuff with entropy & friends.
- These tricks are for off-line browsing ("analysis"), not IDS/IPS magic.
 - but they might help you understand that magic.

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The UNIX pipe length contest

What does this do?

grep 'Accepted password' /var/log/secure |
 awk '{print \$11}' | sort | uniq -c | sort -nr

/var/log/secure:

Jan	13	21:11:11	zion	sshd[3213]:	Accepted password for root from 209.61.200.11
Jan	13	21:30:20	zion	sshd[3263]:	Failed password for neo from 68.38.148.149
Jan	13	21:34:12	zion	sshd[3267]:	Accepted password for neo from 68.38.148.149
Jan	13	21:36:04	zion	sshd[3355]:	Accepted publickey for neo from 129.10.75.101
Jan	14	00:05:52	zion	sshd[3600]:	Failed password for neo from 68.38.148.149
Jan	14	00:05:57	zion	sshd[3600]:	Accepted password for neo from 68.38.148.149
Jan	14	12:06:40	zion	sshd[5160]:	Accepted password for neo from 68.38.148.149
Jan	14	12:39:57	zion	sshd[5306]:	Illegal user asmith from 68.38.148.149
Jan	14	14:50:36	zion	sshd[5710]:	Accepted publickey for neo from 68.38.148.149

And the question is:

44	68.38.148.149
12	129.10.75.101
2	129.170.166.85
1	66.183.80.107
1	209.61.200.11

Successful logins via ssh using password by IP address

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...where is my WHERE clause?

What is this?

SELECT COUNT(*) as cnt, ip FROM logdata
 GROUP BY ip ORDER BY cnt DESC

var.log.secure

serial	date	time	host	daemon	message	pid	ip	user
10	2006-01-13	21:11:11	zion	sshd[3213]	Accepted password for root from 209.61.200.11	3213	209.61.200.11	root
	2006-01-13	21:30:20	zion	sshd[3263]	Failed password for neo from 68.38.148.149	3263	68.38.148.149	neo
12	2006-01-13	21:34:12	zion	sshd[3267]	Accepted password for neo from 68.38.148.149	3267	68.38.148.149	neo
13	2006-01-13	21:36:04	zion	sshd[3355]	Accepted publickey for neo from 129.10.75.101	3355	129.10.75.101	neo
14	2006-01-14	00:05:52	zion	sshd[3600]	Failed password for neo from 68.38.148.149	3600	68.38.148.149	neo
15	2006-01-14	00:05:57	zion	sshd[3600]	Accepted password for neo from 68.38.148.149	3600	68.38.148.149	neo
16	2006-01-14	12:06:40	zion	sshd[5160]	Accepted password for neo from 68.38.148.149	5160	68.38.148.149	neo
□ 17	2006-01-14	12:39:57	zion	sshd[5306]	Illegal user asmith from 68.38.148.149	5306	68.38.148.149	asmith
18	2006-01-14	14:50:36	zion	sshd[5710]	Accepted publickey for neo from 68.38.148.149	5710	68.38.148.149	neo

cn	t	ip
	44	68.38.148.149
	12	129.10.75.101
	2	129.170.166.85
	1	66.183.80.107
	1	209.61.200.11

(Successful logins via ssh using password by IP address)

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Must... parse... syslog...

Wanted:

Free-text syslog records \rightarrow named fields

Reality check

- printf format strings are at developers' discretion
- 120+ types of remote connections & user auth in Fedora Core

Pattern language

sshd: Accepted %auth for %user from %host Failed %auth for %user from %host Failed %auth for illegal %user from %host ftpd: %host: %user[%pid]: FTP LOGIN FROM %host [%ip], %user

"The great cycle"

	• Filter
	2 Group
	Ount
The case	Sort
Constant of	Sinse Repeat

grep user1 /var/log/messages | grep ip1 | grep ... awk -f script ... | sort | uniq -c | sort -n

SELECT * FROM *logtbl* WHERE user = '*user1*' AND ip = '*ip1*' GROUP BY ... ORDER BY ...

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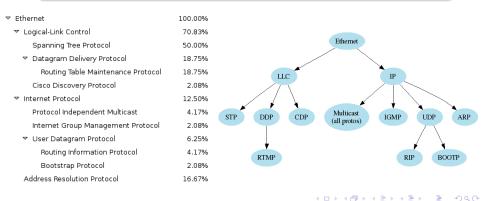
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Can we do better than pipes & tables?

Humans naturally think in classification trees:

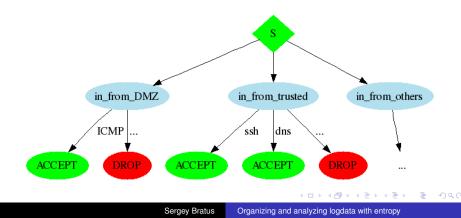
- Protocol hierarchies (e.g., Wireshark)
- Firewall decision trees (e.g., iptables chains)



Can we do better than pipes & tables?

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Pipes, SQL queries \rightarrow branches / paths

Groups \leftrightarrow nodes (sorted by count / weight), records \leftrightarrow leaves.

🖢 TreeView 📃 🗆 🔀										
<u>File Edit I</u> emplate										
113391 Snort portscan alerts ← 113191 dst_port: 1445 src_ip: [55] dst_jp: [75] ← 1701 dst_port: 20 src_ip: [8] dst_jp: [75] ← 1261 dst_port: 21 src_ip: 80.141.141.173 dst_jp: [11] ← 1222 dst_port: 1699 src_ip: 218.103.195.242 dst_jp: [22] ← 1201 dst_port: 1699 src_ip: 129.170.125.243 dst_jp: [8] ← 1151 dst_port: 1433 src_ip: 129.170.125.243 dst_jp: [8] ← 1131 dst_port: 1433 src_ip: 129.170.125.243 dst_jp: [9] ← 1221 dst_port: 1433 src_ip: 129.15.341 dst_jp: [12] ← 123 dst_port: 1 src_ip: 209.15.84, 72 dst_jp: [9] ← 124 dst_port: 1 src_ip: 209.15.84, 72 dst_jp: [9]	Attributes Field # Value _minute 54 type SYN _month Apr loghost annon _hour 19 _line Apr 15 19:54:10 annon snort: 194,208,40									
⊕-[3] dst_port: 8000 src_jp: 194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 8000 src_jp: 194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 8100 src_jp: 194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 8100 src_jp: 194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 8000 src_jp: 194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 8000 src_jp: [194,208,40.120 dst_pp: [2] ⊕-[3] dst_port: 5008,40.120 dst_pp: [2] ⊕-[3] dst_port: 195/5410 annon snott: 194,208,40.120 4743 > 125	src_port 194.208.40.120 arc_port 4743 flags ************************************									
Apr 15 19:55:00 annon snort: 194.208.40.120 4914 -> 12: Apr 15 19:55:06 annon snort: 194.208.40.120 4914 -> 12:										
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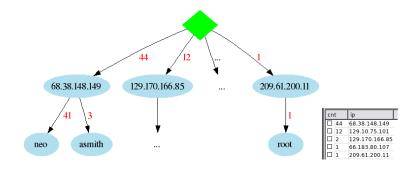
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Pipes, SQL queries \rightarrow branches / paths

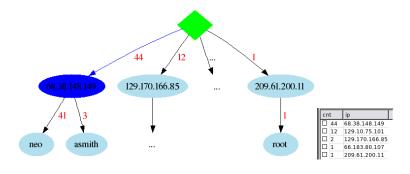
Groups \leftrightarrow nodes (sorted by count / weight), records \leftrightarrow leaves. Queries pick out a leaf or a node in the tree.



grep 68.38.148.149 /var/log/secure | grep asmith | grep ...

Pipes, SQL queries \rightarrow branches / paths

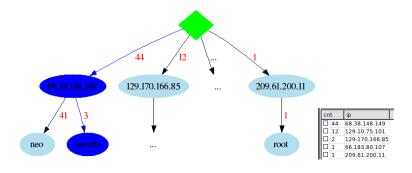
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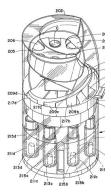
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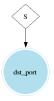
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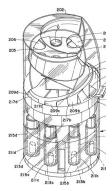
A "coin sorter" for records/packets





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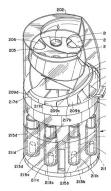
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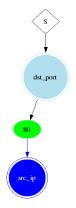
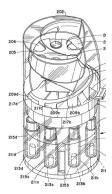
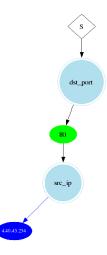


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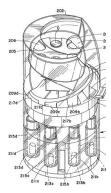
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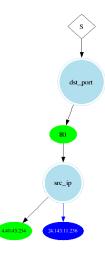




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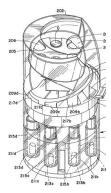


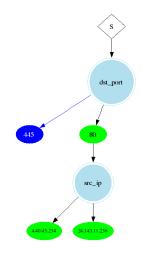


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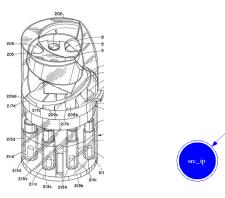


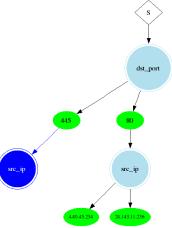


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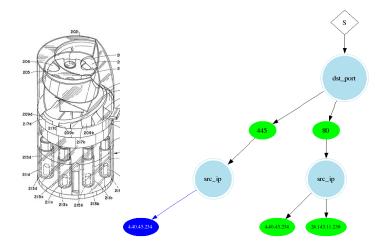
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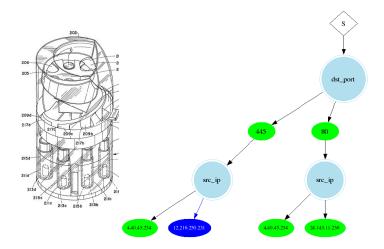
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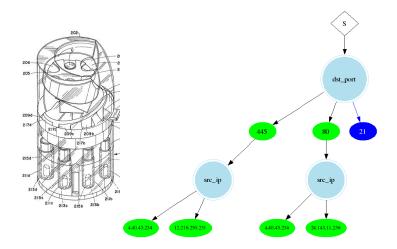
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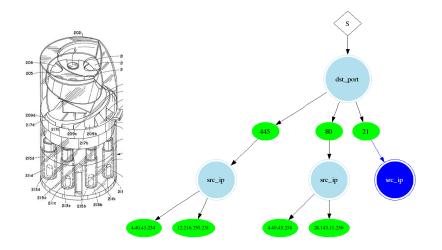
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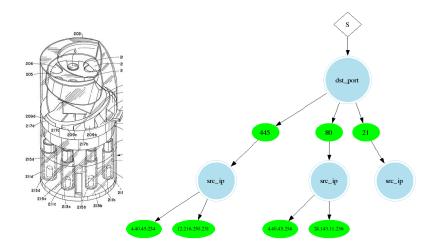
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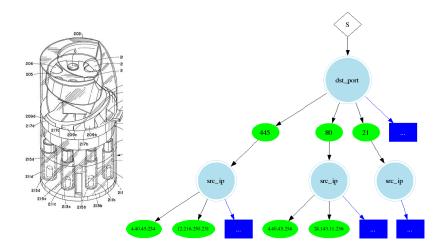
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A "coin sorter" for records/packets



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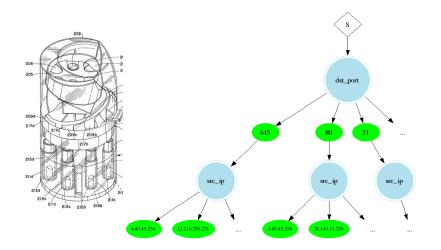


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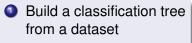


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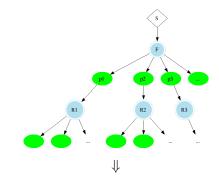
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$Classify \rightarrow Save \rightarrow Apply$

Edit	⊻jew	Help	
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D [146]		Sep 2002	
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Þ 1531		Nov 2002	
↓ [71]		Dec 2002	
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		Feb 2003	
\bigtriangledown B		Mar 2003	
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		Mar 25 2	0:36:48 mystic syslog: LOGIN ON typ0 BY john FROM h000502032ae9.ne.isp.net
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[1]	k	ogin by <mark>john</mark>	from h0010b565bb03.ne.isp.net
17]	FAI	LED LOGIN	%num=((d+) FROM %host FOR %%user, %%reason
- 081	k	ogin by <mark>jos</mark> fr	om h000502032ae9.ne.isp.net
- 131	k	ogin by from	h000502032ae9.ne.isp.net
[1]	k	ogin by <mark>johs</mark>	from h000502032ae9.ne.isp.net
[1]	k	ogin by <mark>john</mark>	[[D from h000502032ae9.ne.isp.net
- 11	k	ogin by <mark>(null</mark>	from h000502032ae9.ne.isp.net
[1]	k	ogin by <mark>r)</mark> fro	mh000502032ae9.ne.isp.net
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- 2 Save template
- Reuse on another dataset



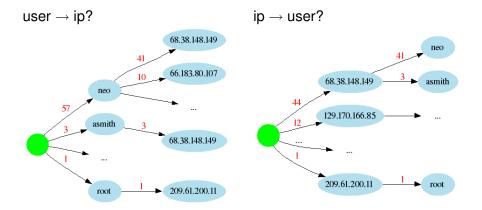
• (617)	617 logins from mediaone.net	Fields 3	Slate	Featur	es 1	pp Index
··· (616)	host: h000502032ae9.ne.mediaone.net user: 7 tty: 5	Feature		ε	# E	ntropy
p (599)	user: josh tty: 4	HINhor	st Nous	er)	CH D	.169/1.18
(R) (E	user: jos tty: ()	HINUS	er Mho	st)	cH 0	.000/1.00
(2) (2)	user: tty:()					
Þ (3)	user: johs tty: ()					
P (1)	user: (null) tty: ()	1				
> (1)	user:r] tty:0					
P 111	user: josh^[[D tty: []	1				
▶ 110	host: we-24-31-59-152.we.mediaone.net user: (oleg) tty: 2					
▶ 111	host: h0010b565bb03.ne.mediaone.net_user: (iosh)_tty: (ttyp0)	1				

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Organizing and analyzing logdata with entropy

Which tree to choose?



Goal: best grouping

How to choose the "best" grouping (tree shape) for a dataset?

Sergey Bratus Organizing and analyzing logdata with entropy

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Outline

Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

Data organization

- Trying to define the browsing problem
- Entropy
- Measuring co-dependence
- Mutual Information
- The tree building algorithm

3 Examples

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Trying to define the browsing problem

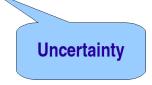
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	1 0.000000	192.168.101.2	192.168.1.40	DNS	Standard query response A 192.168.4.2					
	2 0.000509	192.168.1.40	192.168.4.2	TCP	32790 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0 MS					
	3 2.997399	192.168.1.40	192.168.4.2	TCP	32790 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0 MS					
	4 5.211690	192.168.101.2	192.168.1.40	DNS	Standard query response A 192.168.7.2					
	5 5.211934	192.168.1.40	192.168.7.2	TCP	32791 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0 MS					
	5 5.217693	192.168.7.2	192.168.1.40	TCP	www > 32791 [SYN, ACK] Seq=0 Ack=1 Win=17376 Le					
	7 5.217792	192.168.1.40	192.168.7.2	TCP	32791 > www [ACK] Seq=1 Ack=1 Win=5840 Len=0 TS					
	8 5.217895	192.168.1.40	192.168.7.2	HTTP	GET / HTTP/1.1					
	9 5.419530	192.168.7.2	192.168.1.40	TCP	www > 32791 [ACK] Seq=1 Ack=406 Win=17376 Len=0					
	3 6.634156	192.168.7.2	192.168.1.40	HTTP	HTTP/1.1 200 OK[Unreassembled Packet]					
-	1 6.634441	192.168.1.40	192.168.7.2	TCP	32791 > www [ACK] Seq=406 Ack=1449 Win=8688 Len					
	2 6.635032	192.168.7.2	192.168.1.40	HTTP	Continuation or non-HTTP traffic					
	3 6.635237	192.168.7.2	192.168.1.40	HTTP	Continuation or non-HTTP traffic					
•					<u> </u>					
⊳ Fram	e 1 (112 byt	tes on wire, 112 byt	es captured)		-					
⊳ Raw j	backet data									
⊳ Inte	rnet Protoci	ol, Src Addr: 192.16	8.101.2 (192.168.101.	2), Dst An	dr: 192.168.1.40 (192.168.1.40)					
⊳ User	Datagram Pr	otocol, Src Port: d	omain (53), Dst Port:	32771 (3	(771)					
Doma:	in Name Syst	tem (response)			Ŧ					
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mie: 010	ju.znuu.pcap.	1 30 MB 13:32:40		J. 10203	D: 100030 M: U					

- The lines you need are only 20 **PgDn**s away:
- ...each one surrounded by a page of chaff...
- ...in a twisty maze of messages, all alike...
- ...but slightly different, in ways you don't expect.

Trying to define the browsing problem

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	2 0.000509	192.168.1.40	192.168.4.2	TCP	32790 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0 MS			
	3 2.997399	192.168.1.40	192.168.4.2	TCP	32790 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0 M			
	4 5.211690	192.168.101.2	192.168.1.40	DNS	Standard query response A 192.168.7.2			
	5 5.211934	192.168.1.40	192.168.7.2	TCP	32791 > www [SYN] Seq=0 Ack=0 Win=5840 Len=0			
	6 5.217693	192.168.7.2	192.168.1.40	TCP	www > 32791 [SYN, ACK] Seq=0 Ack=1 Win=17376			
	7 5.217792	192.168.1.40	192.168.7.2	TCP	32791 > www [ACK] Seq=1 Ack=1 Win=5840 Len=0			
	8 5.217895	192.168.1.40	192.168.7.2	HTTP	GET / HTTP/1.1			
	9 5.419530	192.168.7.2	192.168.1.40	TCP	www > 32791 [ACK] Seq=1 Ack=406 Win=17376 Len			
	10 6.634156	192.168.7.2	192.168.1.40	HTTP	HTTP/1.1 200 OK[Unreassembled Packet]			
	11 6.634441	192.168.1.40	192.168.7.2	TCP	32791 > www [ACK] Seq=406 Ack=1449 Win=8688 L			
	12 6.635032	192.168.7.2	192.168.1.40	HTTP	Continuation or non-HTTP traffic			
	13 6.635237	192.168.7.2	192.168.1.40	HTTP	Continuation or non-HTTP traffic			
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Þ Fra	me 1 (112 by	tes on wire, 112 byf	es captured)		4			
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Þ Int	ernet Protoc	ol, Src Addr: 192.16	8.101.2 (192.168.101	2), Dst A	dr: 192.168.1.40 (192.168.1.40)			
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Old tricks

	off Mary Gr.	Çapture Analyze Şta	@ulegd.znb0.pcap	1 • Ethen	NI X
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	4 5.211690	192.168.101.2	192.168.1.40		Standard query response & 192.168.7.2
	5 5.211934	192.168.1.40	192.168.7.2	TCP	32790 > www (S1N) Seq=0 Ack=0 Win=5840 Lan=0 MS
	6 5.217693	192.168.7.2	192.168.1.40	TCP	www > 32791 [S1N, ADX] Seq=0 Ack=1 Win=17376 Le
	7 5.217792	192.168.1.40	192.168.7.2	TCP	32791 > www [ACK] Segr1 Ackr1 Winn5040 Lenn0 TS
	8 5.217895	192.168.1.40	192.168.7.2	HTTP	GET / HTTP/1.1
	9 5.419530	192.168.7.2	192.168.1.40	TCP	www > 32791 (ACK) Seq+1 Ack=406 Win=17376 Len=0
	10 6.634156	192.168.7.2	192.168.1.40	HTTP	HTTP/1.1 200 DK[Urreassembled Packet]
	11 6.634441		192.168.7.2	TCP	32791 > www [ACK] Seq=406 Ack=1449 Win=8688 Len
	12 6.639032	192.168.7.2	192.168.1.40	HTTP	Continuation or non-HTTP traffic
	13 6.635237	192.168.7.2	192.168.1.40	HTTP	Cantinuation or non-HITP traffic
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lorat.	cost take has	1 30 MB 13:32:40		RUNESENE	D: 165636 M: 0

Sorting, grouping & filtering:

- Shows max and min values in a field
- Groups together records with the same values
- Drills down to an "interesting" group

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Key problems:

- Where to start? Which column or protocol feature to pick?
- e How to group? Which grouping helps best to understand the overall data?
- I How to automate guessing (1) and (2)?

Old tricks

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	3 1.999953	192,168,1,60	192,168,7,2	IOMP Echo (ping) request	
	4 2.999925	192.168.1.60	192.168.7.2	ICMP Echo (ping) request	
	5 3.999911	192.168.1.60	192.168.7.2	IOMP Echo (ping) request	
	6 4.082445	192.168.1.120	192.168.5.2	ICMP Echo (ping) request	
	7 4,999888	192.168.1.60	192.168.7.2	ICMP Echo (ping) request	
	8 5.082304	192.168.1.120	192.168.5.2	IOMP Echo (ping) request	
	9 5.999866	192.168.1.60	192.168.7.2	ICMP Echo (ping) request	
	18 6.082219	192.168.1.120	192.168.5.2	ICMP Echo (ping) request	
	11 6.999844	192.168.1.60	192.168.7.2	IOMP Echo (ping) request	
	12 7.082132	192.168.1.120	192.168.5.2	ICMP Echo (ping) request	
	13 7.999822	192.168.1.60	192.168.7.2	IOMP Echo (ping) request	
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	er packet data				
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Sorting, grouping & filtering:

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Key problems:

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Old tricks

•			🖉 ulogd.znb0.pcap + t	thereal	
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	2 1.000017	192.168.1.60	192,168,7.2	ICMP Echo (ping) r	equest
	3 1.999953	192.168.1.60	192,168.7.2	ICMP Echo (ping) r	
	4 2.999925	192.168.1.60	192.168.7.2	ICMP Echo (ping) r	equest
	5 3.999911	192.168.1.60	192.168.7.2	IOMP Echo (ping) r	equest
	6 4.082445	192.168.1.120	192.168.5.2	ICMP Echo (ping) r	equest
	7 4,999888	192.168.1.60	192.168.7.2	ICMP Echo (ping) r	equest
	8 5.082304	192.168.1.120	192,168.5.2	IOMP Echo (ping) r	
	9 5.999866	192.168.1.60	192.168.7.2	ICMP Echo (ping) r	
- 1	0 6.082219	192.168.1.120	192.168.5.2	ICMP Echo (ping) r	equest
1	1 6.999844	192,168,1.60	192,168,7.2	ICMP Echo (ping) r	equest
1	2 7.082132	192,168,1,120	192.168.5.2	ICMP Echo (ping) r	equest
	3 7.999822	192.168.1.60	192.168.7.2	IOMP Echo (ping) r	equest .
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Raw	packet data				
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	ad tob0 or to	3614 KB 04:25:28	100	(0943 D: 60943 M: 0	

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Key problems:

- Where to start? Which column or protocol feature to pick?
- Output: Out
- How to automate guessing (1) and (2)?

Estimating uncertainty

Trivial observations

- Most lines in a large log will not be examined directly, ever.
- One just needs to convince oneself that he's seen everything interesting.
- "Jump straight to the interesting stuff", compress the rest.

Example

The problem:

Must deal with uncertainty vs. redundancy of the data.

Measure it!

There is a **measure** of uncertainty/redundancy: entropy.

Estimating uncertainty

Trivial observations

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Example

АААААААААААААААААААААААААААААААААА<mark>В</mark>ААААААА...

The problem:

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Example

The problem:

Must deal with uncertainty vs. redundancy of the data.

Measure it!

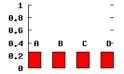
There is a **measure** of uncertainty/redundancy: entropy.

Sergey Bratus Organizing and analyzing logdata with entropy

Entropy intuitions

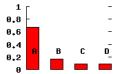
The <u>number of bits</u> to encode a data item under <u>optimal</u> encoding (asymptotically, in a very long stream)

Uniform distribution









BAADBAAAAAABAAACABBABAAAAAAAAA BAAAADBAAC... \Rightarrow 1.42 bits/symbol

Entropy of English: 0.6 to 1.6 bits per char (best compression).

The entropy of English?

Depending on the model, 0.6 to 1.6 bits per character.

letters, unigrams	XFOML RXKHRJFFJUJ ZLPWCFWKCYJ FFJEYVKCQSGHYD QPAAMKBZAACIBZLHJQD ZEWRTZYNSADXESYJRQY WGECIJJ
bigrams	OCRO HLI RGWR NMIELWIS EU LL NBNESEBYATH EEI ALHENHTTPA OOBTTVA NAH BRL OR L RW NILI E NNSBATEI AI NGAE ITF NNR ASAEV OIE BAINTHA HYROO POER SETRYGAIETRWCO
trigrams	ON IE ANTSOUTINYS ARE T INCTORE ST BE S DEAMY ACHIN D ILONSIVE TUCOOWE AT TEASONARE FUSO TIZIN ANDY TOBE SEACE CTISBE
words, unigrams	REPRESENTING AND SPEEDILY IS AN GOOD APT OR COME CAN DIFFERENT NATURAL HERE HE THE A IN CAME THE TO OF TO EXPERT GRAY COME TO FURNISHES THE LINE HAD MESSAGES
bigrams	THE HEAD AND IN FRONTAL ATTACK ON AN ENGLISH WRITER THAT THE CHARACTER OF THIS POINT IS THEREFORE ANOTHER METHOD FOR THE LETTERS THAT THE TIME OF WHOEVER
trigrams*	THE BEST FILM ON TELEVISION TONIGHT IS THERE NO-ONE HERE WHO HAD A LITTLE BIT OF FLUFF

Shannon's experiment

Based on how likely humans are to be wrong when predicting the next letter / word (the average number of guesses made to guess the next letter /word correctly) http://math.ucsd.edu/~crypto/java/ENTROPY/

Automating old tricks (1)

"Look at the <u>most</u> frequent and <u>least</u> frequent values" in a column or list.

- What if there are many columns and batches of data?
- Which column to start with? How to rank them?

It would be nice to begin with "easier to understand" columns or features.

Suggestion:

- Start with a data summary based on the columns with simplest value frequency charts (histograms).
- 2 Simplicity \longrightarrow less uncertainty \longrightarrow smaller entropy.

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Suggestion:

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- **2** Simplicity \longrightarrow less uncertainty \longrightarrow smaller entropy.

Trivial observations, visualized

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3 1. 099953	192.168.1.60	192.168.7.2	
4 2/999925	192.168.1.60	192.168.7.2	
5 3.999911	192.168.1.60	192.168.7.2	
6 4.082445	192.168.1.120	192.168.5.2	
7 4.999888	192.168.1.60	192.168.7.2	
8 5.082304	192.168.1.120	192.168.5.2	
9 5.999866	192.168.1.60	192.168.7.2	
10 6.082219	192.168.1.120		
11 6.999844	192.168.1.60		
12 7 .082132	192.168.1.120		
13 7.999822			
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Outline

Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

Data organization

Trying to define the browsing problem

Entropy

- Measuring co-dependence
- Mutual Information
- The tree building algorithm

3 Examples

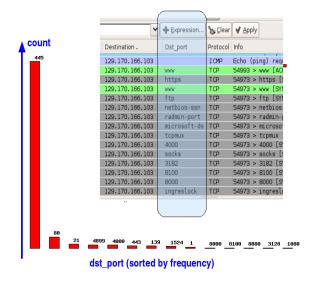
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Start simple: Ranges

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192.188.1.1 - 192.188.10.100	
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File Edit View Go Capture Analyze Statisting Help	1-8100
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1 0.000000 19,168.1.60 192.168.7.2 ICMP Echo (ping) request	
2 1.000017 12 168.1.60 192.168.7.2 ICMP Echo (ping) request	
3 1.999953 1 2.168.1.60 192.168.7.2 ICMP Echo (ping) request	
4 2.999925 1 2.168.1.60 192.168.7.2 ICMP Echo (ping) request	
5 3.999911 92.168.1.60 192.168.7.2 ICMP Echo (ping) request	
6 4.082445 92.168.1.120 192.168.5.2 ICMP Echo (ping) request	
7 4.999888 92.168.1.60 192.168.7.2 ICMP Echo (ping) request	
8 5.082304 192.168.1.120 192.168.5.2 ICMP Echo (ping) request	
9 5.999866 192.168.1.60 192.168.7.2 ICMP Echo (ping) request	
10 6.082219 192.168.1.120 192.168.5.2 ICMP Echo (ping) request	
11 6.999844 192.168.1.60 192.168.7.2 ICMP Echo (ping) request	
12 7.082132 192.168.1.120 192.168.5.2 ICMP Echo (ping) request 13 7.999822 192.168.1.60 192.168.7.2 ICMP Echo (ping) request	
137.999822 192.168.1.60 192.168.7.2 10MP Echo (bind) Peduest	1 - 1
Frame 1 (84 by es on wire, 84 bytes captured)	<u>-</u>
P Frame Dacket data	
 Internet Protocol, Src Addr: 192.168.1.60 (192.168.1.60), Dst Addr: 192.168.7.2 (192.168.7.2) 	
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0000 45 00 00 54 00 00 40 00 3f 01 b2 1a c0 a8 01 3c E.T.@. ?< 0010 c0 a8 07 02 08 00 96 9e 88 03 01 74 12 86 82 40	
0020 49 20 0f 00 08 09 96 96 86 03 0f /4 12 85 82 40	_
0030 14 15 16 17 18 19 1a 1b 1c 1d 1e 1f 20 21 22 23	
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c ip	
2.168.1.1 - 192.168.255.255	
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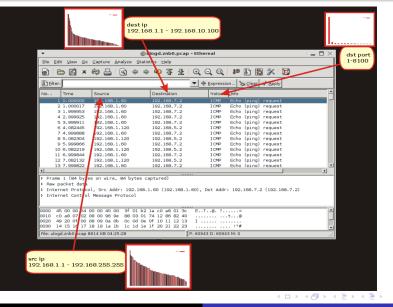
A frequency histogram



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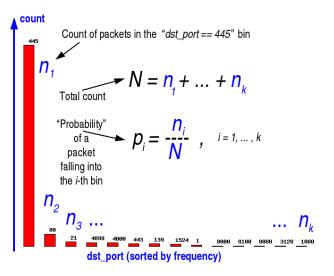
Start simple: Histograms



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Probability distribution



Definition of entropy

Let a random variable X take values x_1, x_2, \ldots, x_k with probabilities p_1, p_2, \ldots, p_k .

Definition (Shannon, 1948)

The entropy of X is

$$H(X) = \sum_{i=1}^{k} p_i \cdot \log_2 \frac{1}{p_i}$$

Recall that the probability of value x_i is $p_i = n_i/N$ for all i = 1, ..., k.

- Entropy measures the <u>uncertainty</u> or lack of information about the values of a variable.
- Entropy is related to the number of bits needed to encode the missing information (to full certainty).

Why logarithms?

Fact:

The least number of bits needed to encode numbers between 1 and N is $\log_2 N$.

Example

- You are to receive one of *N* objects, equally likely to be chosen.
- What is the measure of your uncertainty?

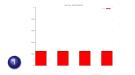
Answer in the spirit of Shannon:

The number of bits needed to communicate the number of the object (and thus remove all uncertainty), i.e. $\log_2 N$.

If some object is more likely to be picked than others, uncertainty decreases.

Log browsing moves Data organization Examples

Entropy on a histogram



Interpretation

Entropy is a measure of uncertainty about the value of X

1 X = (.25 .25 .25 .25) : H(X) = 2 (bits)

2 X = (.5 .3 .1 .1): H(X) = 1.685

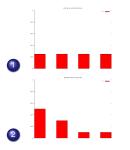
3 X = (.8 .1 .05 .05) : H(X) = 1.022

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(1) $X = (1 \ 0 \ 0 \ 0) : H(X) = 0$

Log browsing moves Data organization Examples

Entropy on a histogram



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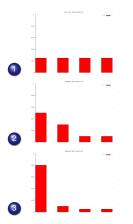
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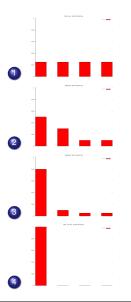
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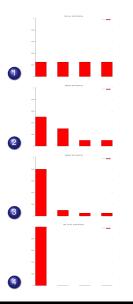
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Entropy on a histogram



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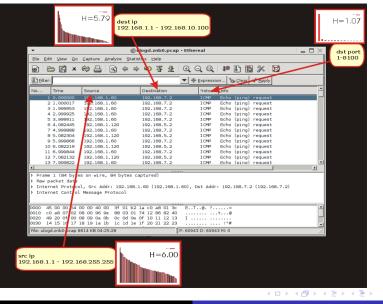
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For only one value, the entropy is 0. When all N values have the same frequency, the entropy is maximal, $\log_2 N$.

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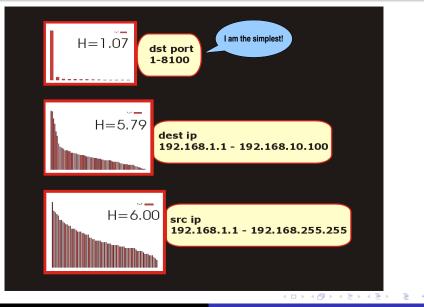
Compare histograms



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Start with the simplest



A tree grows in Ethereal

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Outline

Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

Data organization

- Trying to define the browsing problem
- Entropy

Measuring co-dependence

- Mutual Information
- The tree building algorithm

3 Examples

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Automating old tricks (2)

"Look for correlations. If two fields are strongly correlated on average, but for <u>some</u> values the correlation breaks, look at those more closely".

- Which pair of fields to start with?
- How to rank correlations?

Too many to try by hand, even with a good graphing tool like \underline{R} or Matlab.

Suggestion:

Try and rank pairs before looking, and look at the simpler correlations first.

Simplicity — stronger correlation between features — smaller conditional entropy.

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Examples (1)

Example

Source IP of user logins:

- Almost everyone comes in from a couple of machines
- One user comes in from all over the place. Problem?

Example

Small network, SRC_IP \sim TTL

- On average, src_ip predicts ttl.
- What if a host sends packets with all sorts of ttl?
 - A user just discovered traceroute?
 - What if that machine is a printer or appliance?

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Examples (2)

MUD: Multi-user text adventure (like WoW in ASCII text, only better PvP)

Example

%user gets %obj [%objnum] in room %room

- 2 rooms had by far the largest number of objects picked up.
- Major source of money in the game was: robbers!
 - Stationary camp, safe area, close to cities, easy kill...

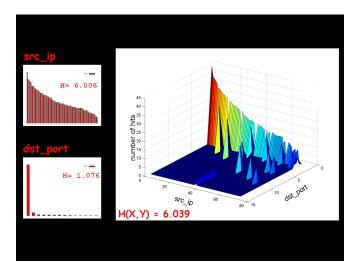
Example

Cheating: player killing by agreement for experience

- A kills B repeatedly, often in the same room. Why?
- A gets experience, warpoints, levels. B is used as a throw-away character, owner of B gets favors.

Log browsing moves Data organization Examples

Histograms 3d: Feature pairs



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Joint Entropy

For fields *X* and *Y*, count # times n_{ij} a pair (x_i, y_j) . is seen together in the same record.

	<i>Y</i> 1	y 2	
<i>x</i> ₁	n ₁₁	n ₁₂	
<i>x</i> ₂	n ₂₁	n ₂₂	
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$$p(x_i, y_j) = \frac{n_{ij}}{N}, \quad (N = \sum_{i,j} n_{ij})$$

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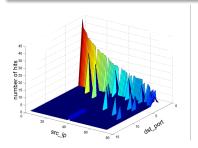
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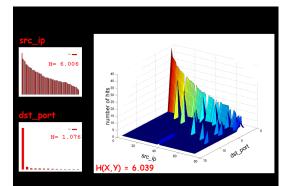
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Log browsing moves Data organization Examples

Measure of mutual dependence

- How much knowing X tells about Y (on average)?
- How strong is the connection?



Compare: H(X, Y) and H(X)

Compare:

H(X) + H(Y) and H(X, Y)

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Independent variables X and Y:

- Knowing X tells us nothing about Y
- No matter what x we fix, the histogram of Y's values co-occurring with that x will be the same shape
- H(X, Y) = H(X) + H(Y)

Dependent X and Y:

- Knowing X tells us something about Y (and vice versa)
- Histograms of ys co-occurring with a fixed x have different shapes
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Outline

Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

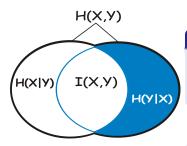
Data organization

- Trying to define the browsing problem
- Entropy
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- Mutual Information
- The tree building algorithm

3 Examples

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Mutual Information



Definition

Conditional entropy of Y given X

$$H(Y|X) = H(X, Y) - H(X)$$

Uncertainty about Y left once we know X.

Definition

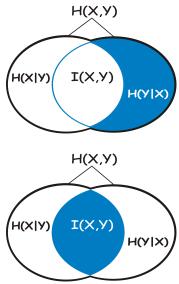
Mutual information of two variables X and Y

I(X; Y) = H(X) + H(Y) - H(X, Y)

Reduction in uncertainty about *X* once we know *Y* and vice versa.

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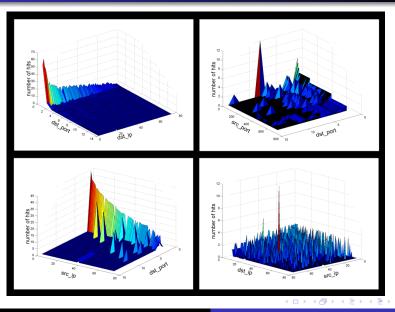
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Log browsing moves Data organization Examples

Histograms 3d: Feature pairs, Port scan

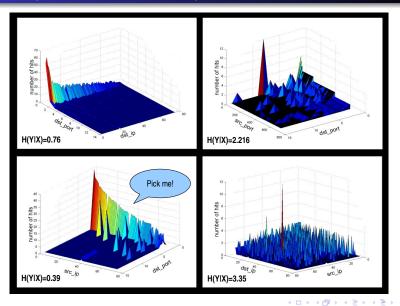


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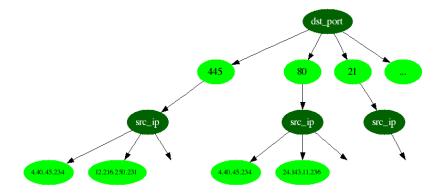
Log browsing moves Data organization Examples

Histograms 3d: Feature pairs, Port scan



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Þ	[1/26]	dst_port: 21	%dst_port %src_ip 76	1 209.15
₽	[1/22]	dst_port: 4899	%dst_port %src_port 687	1 1551,
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₽	[1/15]	dst_port: 443 src_ip: (211.5.239.5) dst_ip: 9 src_port: 9	dst_port 14	1, 21, 8(
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	⊳	[15]	dst_port: 139	_year		2003	
	Þ	[12]	dst_port: 1524	dst_ip	75	129.170.1	
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	Þ	[3, 2]	dst_port: 1080	repeat			

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Outline

Log browsing moves

- Pipes and tables
- Trees are better than pipes and tables!

Data organization

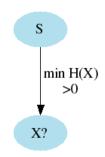
- Trying to define the browsing problem
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3 Examples

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Pick the feature with lowest non-zero entropy ("simplest histogram")

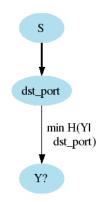
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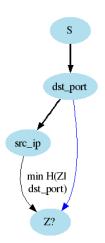
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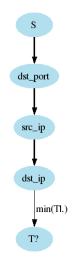
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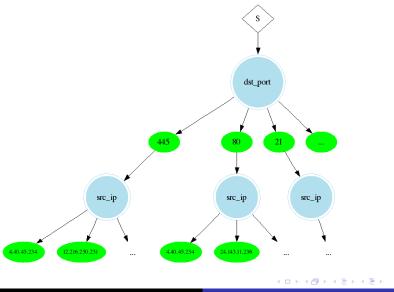


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Þ	[1/26]	dst_port: 21	%dst_port %src_ip 76	1 209.15
₽	[1/22]	dst_port: 4899	%dst_port %src_port 687	1 1551,
₽	[2/20]	dst_port: 4000 src_ip: 2 dst_ip: 8 src_port: 15	dst_ip 75	129.170
₽	[1/15]	dst_port: 443 src_ip: (211.5.239.5) dst_ip: 9 src_port: 9	dst_port 14	1, 21, 8(
₽	[1/15]	dst_port: 139	flags	******S
₽	[1/12]	dst_port: 1524	loghost	annon =
₽	[1/9]	dst_port: 1 src_ip: (209.15.84.72) dst_ip: 9 src_port: 9	program	snort
₽	[1/3]	dst_port: 8100	repeat	
₽	[1/3]	dst_port: 8000	rule_id	732c5ec
₽	[1/3]	dst_port: 8080	serial	-1
Þ	[1/3]	dst_port: 3128	src_ip 71	4.40.45.
₽	[1/3]	dst_port: 1080	src_port 668	1027, 10
			type	SYN
				•
auto	split via m	ninentdep3 without mark	Compute range	s
		autosplit via minentdep3 without mark: OK	Split	

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D			TreeView2 source: Kerf/data/snort2.log				
	Eile	⊑dit ⊻	iew Help				
	- (1	339, <mark>23</mark>]	Snort portscan alerts	Fields	Slate Fea	atures Top	Inc
	Þ	[1135]	dst_port: 445 src_ip: 55 dst_ip: 75 src_port: 100+	Field	#	Value	
	⊳	[70, <mark>1</mark>]	dst_port: 80 src_ip: 8 dst_ip: 30 src_port: 63	_id	100+	e5b80313	
	Þ	[26]	dst_port: 21	_mont	h	Apr	
	Þ	[22]	dst_port: 4899	_progr	am	snort	
	Þ	[20]	dst_port: 4000 src_ip: 2 dst_ip: 8 src_port: 15	_rule_i	d	732c5ed3	
	Þ	[15]	dst_port: 443	_times	tan 100+	Fri 11-Apr-	
	⊳	[15]	dst_port: 139	_year		2003	
	Þ	[12]	dst_port: 1524	dst_ip	75	129.170.1	
	Þ	[9]	dst_port: 1	dst_pd	ort 14	1, 21, 80,	
	⊳	[3, <mark>2</mark>]	dst_port: 8100	flags		*****S*	
	Þ	[3, 2]	dst_port: 8000	loghos	t	annon	
	Þ	[3, 2]	dst_port: 8080	mark		pos	
	Þ	[3, 2]	dst_port: 3128	progra	m	snort	
	Þ	[3, 2]	dst_port: 1080	repeat			

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Quick pair summary

			Tree	View2 sour	ce: Kerf/data	ı/ssh-aı	th-2us	ers			
E	ile	⊑dit	⊻iew Help								
-	[6	17]	617 logins from mediaone.net			Fields	Slate	Fea	tures	Тор	Index
	▽	[606]	host: h000502032ae9.ne.mediaone.net	user: 7 tty	y: 5	Featu	re		E #	Entr	ору
		¢ [589	l user: josh tty: 4			H(%ł	iost %u	ser)	c⊢	0.16	9/1.18
		Þ [8]	user: jos tty: ()			Н(%	ıser∣%h	ost)	c⊢	10.00	0/1.00
		≬ [3]	user: tty:()								
		¢ [3]	user: johs tty: ()								
		≬ [1]	user: (null) tty: ()								
		≬ [1]	user: r] tty: ()								
		≬ [1]	user: josh^[[D tty: ()								
	Þ	[10]	host: we-24-31-59-152.we.mediaone.net	user: (oleg) tty: 2						
	٥	[1]	host: h0010b565bb03.ne.mediaone.net	user: (josh)	tty: (ttyp0)						

One ISP, 617 lines, 2 users, one tends to mistype. 11 lines of screen space.

Quick pair summary

		TreeView2 source: Kerf/data/ssh-au	th-2users
Eile E	dit <u>V</u> ie	w Help	
		617 logins from mediaone.net	Fields Slate Features Top Index
▽ [6	06]	host: h000502032ae9.ne.mediaone.net user: 7 tty: 5	Feature E # Entropy
Þ	[589]	user: josh tty: 4	H(%host %user) CH 0.169/1.18
Þ	[8] .	user: jos tty: ()	H(%user %host) cH 0.000/1.00
Þ	[3]	user: tty: ()	
Þ	[3]	user: johs tty: ()	
Þ	[1]	user: (null) tty: ()	
Þ	[1]	user: r] tty: ()	
Þ	[1]	user: josh^[[D tty: ()	
↓ [1]	0]	host: we-24-31-59-152.we.mediaone.net user: (oleg) tty: 2	
▼ [1]	1	host: h0010b565bb03.ne.mediaone.net user: (josh) tty: (ttyp0)	
~	[1]	user: josh tty: (ttyp0)	
		Jan 10 00:04:14 mystic syslog: LOGIN ON ttyp0 BY josh FROM h0010b	

One ISP, 617 lines, 2 users, one tends to mistype. 11 lines of screen space.

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Novelty changes the order

	TreeView2 source: Kerf/data/snort2b.log					_	
Eile Edit	⊻iew ⊟elp						
▼ [17299]	Snort portscan alerts	A	Fields	Slate	Features	Тор	ndex
♦ [6629]	dst_port: 445 dst_ip: 95 src_ip: 100+ src_port: 100+		Featur	e E	#		
≬ [5428]	dst_port: 139 dst_ip: 94 src_ip: 100+ src_port: 100+		%src_	port	4876/17	299/8.	25
♦ [1743]	dst_port: 80 dst_ip: 94 src_ip: 100+ src_port: 100+	=	%src_	ip	689/172	99/6.1	69
¢ [894]	dst_port: 135 dst_ip: 85 src_ip: 37 src_port: 100+		%dst_	ip	95/1729	9/4.36	1/5
♦ [518]	dst_port:1433 dst_ip:91 src_ip:33 src_port:100+		%dst_	port	35/1729	9/1.78	3/5
Þ [451]	dst_port: 17300 dst_ip: 86 src_ip: 34 src_port: 100+		%flags	3	2/17299	0.004	/1.
∲ [397]	dst_port: 21 dst_ip: 81 src_ip: 23 src_port: 100+		%type		2/17299	0.004	/1
♦ [266]	dst_port:6667 dst_ip:75 src_ip:18 src_port:100+		%logh	ost	1/17299	0.000	/1.
♦ [178]	dst_port: 443 dst_ip: 68 src_ip: 13 src_port: 100+		%seria	al	1/17299	0.000	/1.

TreeView2 source: Kerf/data/snort2b.log		
Eile ⊑dit ⊻iew ⊟elp		
 17299. 1 Snort portscan alerts 	Fields Slate Featu	res Top Index
▷ [17292] flags: *******S* type: (SYN) dst_port: 35 dst_ip: 95 src_ip: 100+	Feature E #	
171 flags: ******SF type: (SYNFIN) dst_port: (21) dst_ip: 7 src_ip: (142.26.217.6	%src_port 487	5/17299/8.25
	%src_ip 689/	17299/6.169
	%dst_ip 95/1	7299/4.361/7

▶ < Ξ >

		-	ew Help		1	_		. [
~	[60943]		do not expand me	Fields	Slate F	ea	atures Top Inc	dex	
			Protocol: TCP Source: 18 Destination: 22 Info: 100+	Featu	re	Е	#	Entropy	
			Protocol: HTTP Source: 13 Destination: 13 Info: 100+	%Info		_	49088/60943	9.744/17050.0	046
			Protocol: IRC Source: 11 Destination: 11 Info: 6	%Des	tination		23/60943	2.254/9.525	
			Protocol: SSHv2 Source: 11 Destination: 11 Info: 84	%Sou	irce		19/60943	2.107/8.225	
	▶ [11/138	3]	Protocol: ICMP Source: 11 Destination: 13 Info: 2	%Prot	tocol		11/60943	0.793/2.211	
	▶ [3/521]		Protocol: MySQL Source: 3 Destination: 2 Info: 100+						
	⊳ [1/96]		Protocol: DNS Source: (192.168.101.2) Destination: 3 Info: 12						
	▶ [5/12]		Protocol: SSH Source: 5 Destination: 4 Info: 7						
	▶ [2/2]		Protocol: DCERPC Source: 2 Destination: 2 Info: 2						
	▶ [1/2]		Protocol: Syslog Source: (192.168.1.40) Destination: (192.168.2.:)	:					
	▶ [1/1]		Protocol: ISystemActivator Source: (192.168.1.50) Destination: (:						
						_			
r				•		_	Load from fi		Þ
1						_			
							Collect from n	ode	
							Compute entr	тору	

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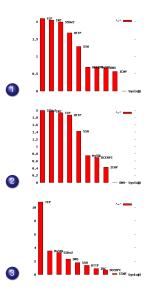
ile	Edit <u>∨</u> iew	/ Help										
Þ	[11/1383]	Protocol: ICMP	Source: 11	Destination: 13	Info: 2	-	Fields	Slate	Features	Top Inde	ex	
Þ	[3/521]			Destination: 2			Feature	,	E #		Entropy	-
	[1/96]			2.168.101.2) D		Info: 12	%Info		12/9	6 :	2.315/10.1	k
	▶ [16/16]	Info: Standar	d query respo	onse PTR blue.roo	otfu.jp		%Dest	nation	3/96		0.675/1.96	-
	▶ [13/13]	Info: Standar	d query respo	onse PTR green.r	ootfu.jp		%Proto		1/96		0.000/1.00	
	▶ [13/13]	Info: Standar	d query respo	onse, No such na	me							
	▷ [13/13]	Info: Standar	d query respo	onse PTR orange.	rootfu.jp		%Sour	ce	1/96		0.000/1.00)(
	▶ [11/11]	Info: Standar	d query respo	onse A 192.168.4	.2							
	▷ [6/6]	Info: Standar	d query respo	onse PTR cyan.ro	otfu.jp							
	▶ [5/5]	Info: Standar	d query respo	onse PTR yellow.r	ootfu.jp		:					
	▶ [5/5]	Info: Standar	d query respo	onse								
	↓ [5/5]	Info: Standar	d query respo	onse A 192.168.2	.2		1					
	▷ [4/4]	Info: Standar	d query respo	onse A 192.168.7	.2							
	▶ [3/3]	Info: Standar	d query respo	onse A 192.168.3	.2							
	▶ [2/2]	Info: Standar	d query respo	onse A 192.168.5	.2							
Þ	[5/12]	Protocol: SSH	Source: 5	Destination: 4	nfo: 7							
	10/01	n i Laoran			(.	-	•					
						Þ			Load	from file		
lit	by "%Info"					•			Collect	from node]	
			split by "%Info	р": ОК					Compi	ite entropy	,	

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Log browsing moves Data organization Examples

Comparing 2nd order uncertainties



Compare uncertainties in each <u>Protocol</u> group:

- Destination: *H* = 2.9999
- Source: *H* = 2.8368
- Info: H = 2.4957

"Start with the simpler view"

Log browsing moves Data organization Examples

Comparing 2nd order uncertainties



Compare uncertainties in each <u>Protocol</u> group:

- **Destination:** H = 2.9999
- Source: *H* = 2.8368
- Info: H = 2.4957

"Start with the simpler view"

File	<u>E</u> dit <u>∨</u> iew	TreeView2 source: xml-packet-summary	.xmi		
_	1/60943, 521	do not expand me	Fields Slate	Features Top	Index
•	[18/48788]	Protocol: TCP Source: 18 Destination: 22 Info: 100+	Feature	E #	Entropy
₽	[13/4378]	Protocol: HTTP Source: 13 Destination: 13 Info: 100+	%Info	49088/6094	13 9.744/17050.
⊳	[11/3435]	Protocol: IRC Source: 11 Destination: 11 Info: 6	%Destination	23/60943	2.254/9.525
⊳	[11/2325]	Protocol: SSHv2 Source: 11 Destination: 11 Info: 84	%Source	19/60943	2.107/8.225
Þ	[11/1383]	Protocol: ICMP Source: 11 Destination: 13 Info: 2	%Protocol	11/60943	0.793/2.211
Þ	[128/521]	Protocol: MySQL Source: 3 Destination: 2 Info: 100+		, , , , , , , , , , , , , , , , , ,	
⊳	[1/96]	Protocol: DNS Source: (192.168.101.2) Destination: 3 Info: 12			
⊳	[5/12]	Protocol: SSH Source: 5 Destination: 4 Info: 7			
⊳	[2/2]	Protocol: DCERPC Source: 2 Destination: 2 Info: 2	:		
Þ	[1/2]	Protocol: Syslog Source: (192.168.1.40) Destination: (192.168.2)	:		
Þ	[1/1]	Protocol: ISystemActivator Source: (192.168.1.50) Destination:			
-				Load from f	
olit	by "%Info"	•		Collect from r	
		Marked nodes: pos 521 neg 0			
				Compute ent	ropy

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File	E	dit <u>∨</u> iew	Help				
-		/521]	TTOCOCOLICIAL SOURCE, 11 DESCINACION, 15 INIO. 2		Fields Slate Fe	eatures Top I	ndex
		[180/180]	Protocol: MySQL Source: 3 Destination: 2 Info: 100+ Info: Response OK	П	Feature E	#	Entropy
		[32/32]	Info: Response Error Code: 417		%Info	128/521	3.516/33.65
		[21/21]	Info: Server Greeting Protocol : 10 ,version: 4.1.1-alpha-log Caps		%Source	3/521	0.748/2.113
	. ₽	[14/14]	Info: Request Command: Quit		%Destination	2/521	0.693/2.000
	⊳	[13/13]	Info: Login Request Caps: 0x2485 ,user: root ,password: ha-log		%Protocol	1/521	0.000/1.000
	Þ	[9/9]	Info: Request Command: Query : show tables				
	⊳	[8/8]	Info: Request Command: Query : show databases				
	Þ	[7/7]	Info: Response Error Code: 428				
	⊳	[6/6]	Info: Server Greeting Protocol : 255[Unreassembled Packet]				
	⊳	[6/6]	Info: Request Command: Init Database : user		1		
	⊳	[6/6]	Info: Response Error Code: 419				
	Þ	[5/5]	Info: Request Command: Query : select user()				
	⊳	[5/5]	Info: Response Error Code: 46c				
	⊳	[4/4]	Info: Request Command: Field List : help_keyword\000				
	Þ	[4/4]	Info: Request Command: Field List : func\000	•	•		
1					_	Load from fi	e
plr	by	"%Info"	split by "%Info": OK	•		Collect from n	ode
			split by "%inio": OK	-		Compute entr	opy

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TreeView2 source: xml-packet-summary.xml			[_ [
jie Edit ⊻jew Help					
Info: Request Command: Field List : host\000		Fields	Slate	Featu	re
Info: Request Command: Field List : db\000		Feature	,	E #	-
Info: Request Command: Field List : tables_priv\000		%Info		- 4	-
Info: Request Command: Field List : user\000		%Desti	nation		
Info: Request Command: Field List : columns_priv\000		%Sour		1	
Info: Request Command: Field List : help_category\000		%Proto		1	1
Info: Request Command: Field List : help_relation\000		7601010	001	1	-/
Info: [TCP Retransmission] Request Command: Quit					
Info: Request Command: Query : set password for 'root'@'green-router.rootfu.jp'=password('schoolofr00t')					
Info: Request Command: Query : update user set password=password('schoolofr00t') where user = 'root@		:			
Info: Request Command: Query : ALTER TABLE user CHANGE COLUMN Password Password LONGTEXT					
Info: Request Command: Field List : psl_story\000					
Info: Request Command: Field List : psl_section_block_lut\000					
Info: Request Command: Field List : db_sequence\000					
Info: Request Command: Field List : psl_quote\000					
Info: Request Command: Field List : psl_commentcount\000					_
Inferior and contract of contraction and the contraction of		•		61	10
lit by "%Info"			ad fro		
Marked nodes: pos 6 neg 0	-	Colle	ect fro	m noc	e
Harked Houss, bus offied o		Corr	npute	entrop	y

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Screenshots (1)

Possible Attributes	Selected Attributes	Order Attributes	
	Protocol attributes	packetlist.protocol	
Ethernet	Packet list attributes	pacacastipiotocol	
Frame	packetlist.protocol - Protocol		
> IP	packets aprotocor i rotocor		
▶ LLC			
▶ NBNS	💠 Add	Add	th Up
▶ STP	-1- <u>1</u> - <u>1</u> -		<u> </u>
▶ UDP			
✓ Packet list attributes			
packetlist.number - Number			
packetlist.time - Time (format as			
packetlist.time_relative - Relative			
packetlist.time_absolute - Absolu			
packetlist.abs_data_time - Absolu			
packetlist.time_delta - Delta time			
packetlist.source_address - Sour	~ 1		
packetlist.resolved_src_addr - Sn	<u> </u>	Pelete Delete	- ⊕ Dov
packetlist.unresolved_src_addr -			
packetlist.harware_src_addr - Ha			
packetlist.resolved_hw_src_addr		Sort by	5
packetlist.unresolved_hw_src_ad			<u>-</u>
· · · ·			
Rename attributes			_
lame:			New Attrib
lias:			Apply
Select Learning Algorithm			
Igorithm: Minimum Entropy Tree (JSV tre			•
'arameters: -max_tree_depth=10 -max_clust	er_size=10 -lower_entropy_threshold=0.000000 -upper_e	ntropy_threshold=0.800000	
🔯 Help		Save 🗁 Open 🗶 Cance	

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Organizing and analyzing logdata with entropy

Screenshots (2)

No	Time	Source	De	estination	Protocol	HwAddr
1	0.000000	192.168	.2.3 19	5.138.145.122	UDP	00:11:50:38:81:70
2	0.031164	192.168	Mark Packet	(togale)	UDP	00:11:50:38:81:70
3	0.061035	192.168	Time Referen		UDP	00:11:50:38:81:70
4	0.072645	195.138	Time Reference		UDP	00:0d:60:76:d9:ce
5	0.090327	195.138	Find in Treevi	ew	UDP	00:0d:60:76:d9:ce
6	0.091400	192.168	7		UDP	00:11:50:38:81:70
7	0.119496	192.168	Apply as Filte		UDP	00:11:50:38:81:70
8	0.121801	195.138	Prepare a Filt	er 🕨	UDP	00:0d:60:76:d9:ce
9	0.149736	192.168			UDP	00:11:50:38:81:70
10	0.159995	195.138			UDP	00:0d:60:76:d9:ce
11	0.177547	192.168	දී Decode As Print		UDP	00:11:50:38:81:70
12	0.193140	195.138			UDP	00:0d:60:76:d9:ce
13	0.208144	192.168	Show Packet	in New Window	UDP	00:11:50:38:81:70
14	0.215444	195.138	.145.122 19	2.168.2.3	UDP	00:0d:60:76:d9:ce

Field name		I	Ethereal formula	Unique values	Entropy -	Values summa	f
tcp.analysis.z	ero_window			2	0.015826	#undef, Zero V	
tcp.options.sa	ck_le			2	0.028560	#undef, 1188	
tcp.analysis.re	transmission			2	0.028560	#undef, Retrar	
•			Split Show distinct values		•		
Value	Count	Fie	Compute condit	ional entropies	Field 1 name	Field 2 name	С
#undef	447						
Retransmissio	n 2	1					
		2					

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Screenshots (3)

⊽ [18944/9]	Min Entropy Tree	Ranges Template Messa	ges		
Þ	[18335/17]	packetlist.protocol UDP	Field name	Ethereal formula Unique values	Entropy Values su	ummary
Þ		packetlist.proto Apply template	frame.number	449	6.107023 10046, 10	0048, 1
Þ	[54/2]	packetlist.protor Save template	ip.dsfield.dscp	1	0.000000 0000 00	= Defa
Þ	[51/3]	packetlist.protor	frame.protocols	2	0.507120 eth:ip:tcp	, eth:ip
Þ	[26/2]	packetlist.protor Show template				F
Þ	[16/2]	packetlist.protor Mark positive	:		Field 1 name Field 2	
Þ	[7/7]	packetlist.protor Mark negative	Value Count	Field name Cond. entropy	Field 1 name Field 2	name
Þ	[5/2]	packetlist.protor Clear markings				
Þ	[1/1]	packetlist.protocol AIM				
					(

Name	Label	Test		
▼ root	Min Entropy Tree	0		
	"packetlist.protocol" %packetlist.protocol	[] [%packetlist.protocol = "		
	"eth.trailer" %eth.trailer			
	"eth.dst" %eth.dst	[%eth.trailer = "#undef"] [] [] [%packetlist.protocol =		
leaf	%line			
leaf	%line			
▼ tcp.ack	"tcp.ack" %tcp.ack			
<u>ا</u>		<u>•</u>		
Change values Label: "packetlist.protocol" %packe	tlist.protocol			
Sort key: @num_leaves				
Sort order: desc				

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Research links

Research on using entropy and related measures for network anomaly detection:

- Information-Theoretic Measures for Anomaly Detection, Wenke Lee & Dong Xiang, 2001
- <u>Characterization of network-wide anomalies in traffic flows</u>, Anukool Lakhina, Mark Crovella & Christiphe Diot, 2004
- Detecting Anomalies in Network Traffic Using Maximum Entropy Estimation, Yu Gu, Andrew McCallum & Don Towsley, 2005

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Summary

Information theory provides useful heuristics for:

- summarizing log data in medium size batches,
- choosing data views that show off interesting features of a particular batch,
- finding good starting points for analysis.

Helpful even with simplest data organization tricks.

In one sentence

 $H(X), H(X|Y), I(X;Y), \ldots$ parts of a complete analysis kit!

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Summary

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Credits & source code

Credits	
Kerf project:	Javed Aslam, David Kotz, Daniela Rus,
	Ron Peterson
Coding:	Cory Cornelius, Stefan Savev
Data & discussions:	George Bakos, Greg Conti,
	Jason Spence, and many others.
Sponsors:	see website

Code

For source code (GPL), documentation, and technical reports:

http://kerf.cs.dartmouth.edu

Thanks!					
	· · · · · · · · · · · · · · · · · · ·	940			
Sergey Bratus	Organizing and analyzing logdata with entropy				