

Reasoning Analytically About Password-Cracking Software

Enze "Alex" Liu, Amanda Nakanishi, Maximilian Golla, David Cash, and Blase Ur

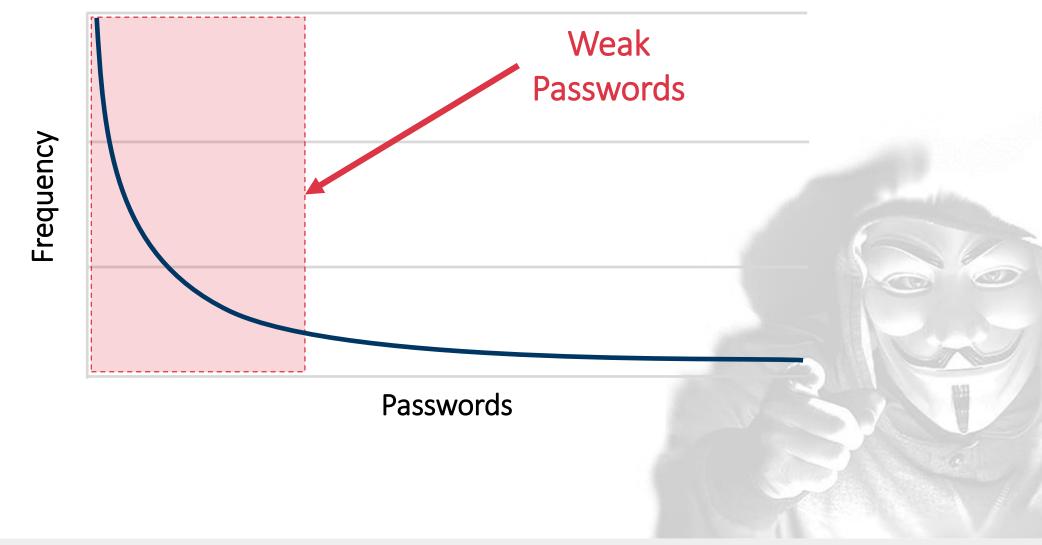


People Choose Weak Passwords



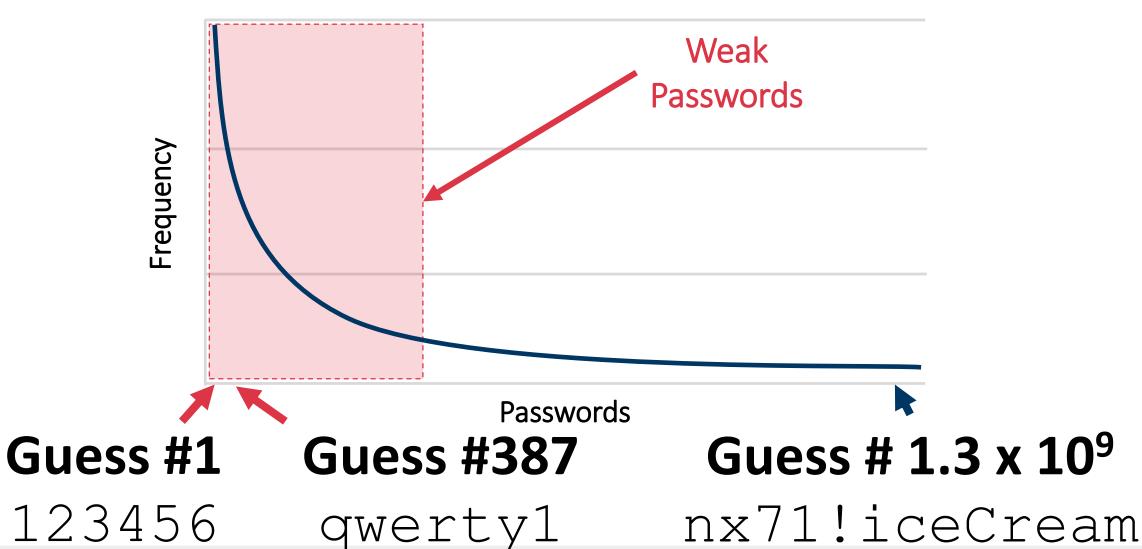


What Makes a Password "Weak"?





What Makes a Password "Weak"?





Guess Number = Approximate Strength

Example:

Johnny14!

Guess #:

390,000



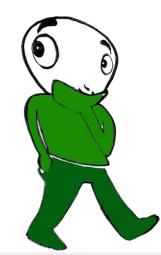
noun

¹ The number of guesses required to guess a password.



Application 1: Strength Meters

Password Strong Strong





Application 2: Proactive Checking

Password123!



Application 2: Proactive Checking



Application 3: Academic Research

Do Users' Perceptions of Password Security Match Reality?

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Although many users create predictable passw tent to which users realize these passwords are is not well understood. We investigate the rela tween users' perceptions of the strength of speci and their actual strength. In this 165-particinant we ask participants to rate the comparative sec fully juxtaposed pairs of passwords, as well a and memorability of both existing passwords password-creation strategies. Participants had se entions about the impact of basing passwords phrases and including digits and keyboard patt words. However, in most other cases, participants of what characteristics make a password secure tent with the performance of current password-c ten win the performance of current password-or We find large variance in participants' understar passwords may be attacked, potentially explainit nonetheless make predictable passwords. We or design directions for helping users make better p

User behavior; perceptions of security; passwor authentication; users' folk models; usable secur

ACM Classification Keywords

For better or worse, passwords remain today's d of user authentication [11]. While the predictal of user authentication [11], while the predictar chosen passwords has been widely documented 74,77,80], very little research has investigated u tions of password security. That is, do users rea selecting terrible passwords and choose to do so or are they unwittingly creating weak passwon

In this paper, we report on a 165-participant s perceptions of password security. Participants 1 perceptions about the security and memorabi

Measuring Password Guessability for an Entire University

Michelle L. Mazurek, Saranga Komanduri, Timothy Vidas, Lujo Bauer Nicolas Christin, Lorrie Faith Cranor, Patrick Gage Kelley*, Richard Shay, and Blase Ur

> Carnegie Mellon University ittsburgh, PA arangak, tvidas, Ibauer, , rshay, bur}@cmu.edu

> > empirical studies of

of access to plaintext

specifically collected

ts. Properties of pass-

in poorly understood.

en-on passwords used

a research university

s of our contributions

swords. We describe the many precautions

malyze how guessable

offline attack by sub-

acking algorithm. We

imber of demographic

For example, we find

ce school make pass-

se of users associated

nd that stronger pass-

ollected in controlled

its. We find more con-

words and passwords

composition policies

s and subsets of passappen to comply with

ors entering them.

*University of New Mexico Albequerque, NM pgk@cs.unm.edu

"I Added '!' at the End to Make It Secure":

Observing Password Creation in the Lab Blase Ur, Fumiko Noma, Jonathan Bees, Sean M. Segreti, Richard Shay, Lujo Bauer, Nicolas Christin, Lorrie Faith Cranor Carnegie Mellon University

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Users often make passwords that are easy for attackers to guess. Prior studies have documented features that lead to easily guessed sswords, but have not probed why users craft weak passwords. plasswords, but lave no protect with users craft weak passwords. To understand the genesis of common password patterns and un-cover average users' misconceptions about password strength, we conducted a qualitative interview study. In our lab, 49 participants each created passwords for fictitious banking, email, and news website accounts while thinking aloud. We then interviewed them about their general strategies and inspirations. Most participants had a well-defined process for creating passwords. In some cases, partic-ioants consciously made weak passwords. In other cases, however, weak passwords resulted from misconceptions, such as the belief that adding "!" to the end of a password instantly makes it secure or that words that are difficult to spell are more secure than easy-tospell words. Participants commonly anticipated only very targeted attacks, believing that using a birthday or name is secure if those data are not on Facebook. In contrast, some participants made se-

Despite decades of research investigating passwords, many users ill make passwords that are easy for attackers to guess [9, 22, 35, 62]. Predictable passwords continue to cause problems, as evidenced by the recent release of celebrities' private photos obtained denoted by the feeth remeased or circumses private pionion conduited in part through a password-guessing attack on Apple's (Cloud [11, 37], While most everyone would prefer a world without the barden of remembering a portfolio of passwords [18, 33], passwords are familiar, easy to implement, and do not require that users carry anything. As a repair of the property the near future [7]. Although expecting users to remember complex and distinct passwords for dozens of accounts is absurd, single sise to reduce this burden. Passwords also remain useful for

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sult in easy-to-guess passwords [9, 22, 35, 62]. While some users may be making informed cost-benefit analyses and creating weak may be making informed cost-benefit analyses and creating weak passwords for low-value accounts, other users may have miscon-ceptions about what makes a good password. Existing security ad-vice [10, 28, 36, 49, 73] and real-time feedback [1, 12, 15, 23, 25, 9, 63] may be insufficient in disabasing users of these misconceptions. To understand where users fall short in their attempts to cre-

ate passwords, we conducted the first qualitative laboratory study are passworts, we connected une irra quantum conversion of the process of passwort creation. Whereas analyses of large sets of passwords can reveal common patterns, a qualitative study is better suited to discern precisely why these patterns appear because researchers can probe the rationale behind behaviors through the control of the context-based follow-up questions. Prior lab studies of password have focused on password management [2, 20, 27, 29, 52, 55], how users cope with password-composition requirements [45,66], nove account systems [19] and the external validity of password stud password systems [19], and me external valuntly or password sun-ies [16]. In this paper, we report on the first lab study focusing ex-clusively on how users craft and compose passwords step-by-step. We conducted in-person lab sessions with 49 participants, each of whom created passwords for a banking website, news website,

and email account in a think-aloud, role-playing scenario. We also xplored participants' general strategies and inspirations. This enasset us to propoint participants: misconceptions and identity strate-gies that seem both usable and secure against large-scale guessing attacks, such as an offline attack [6, 31, 70].

We found that most participants had a well-defined process for

creating passwords. Commonly, participants either had a base word or a systematic human "algorithm" for generating passwords based or a systematic human "algorithm" for generating passwords based on the site. While many strategies led to predictable passwords, some participants successfully mixed unrelated words or crafted unique phrases to create more secure passwords. Some participants desired passwords of different security levels across the three web-sites, yet nearly half did not, indicating that some people may routinely waste effort creating and remembering strong passwords for low-value accounts. Participants struggled to create passwords that matched their desired security levels, sometimes creating strong words that they inteded to be weak, and vice vers

Participants were concerned primarily with targeted attacks on heir passwords, rather than large-scale, automated attacks. As a result, some participants believed the (common) name of their pets or irthdays would be strong passwords because they had not posted that information on their Facebook page, not accounting for the that information on their Facebook page, not accounting for the types of automated guessing attacks often seen in the wild when sites like Linkedin [9], eHarmony [57], Gawker [5], or Adobe [43] had their password databases compromised. We identified numerous other security misconceptions. Most participants knew that dictionary words make bad passwords, yet

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are not made or distributed

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ments of this work must be

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1. INTRODUCTION

Researchers have documented the numerous problems of text passwords for decades - passwords are easy to guess, hard to remember, easily stolen, and vulnerable to observation and replay attacks (e.g., [28, 38]). The research community has invested significant effort in alternatives including biometrics, graphical passwords, hardware tokens, and federated identity; however, text passwords remain the dominant mechanism for authenticating people to computers, and seem likely to remain that way for the fore able future [5,23]. Better understanding of text passwords therefore remains important.

Considerable effort has been spent studying the usage and characteristics of passwords (e.g., [13, 17, 34, 35, 45]), but password research is consistently hampered by the difficulty in collecting realistic data to analyze. Prior password studies all have one or more of the following drawbacks: very small data sets [36], data from experimental studies rather than from deployed authentication systems [31], no access to plaintext passwords [3], self-reported password information [47], leaked data of questionable validity, or accounts of minimal value [26,53]. As a result, the important question of whether the results apply to real, high-value passwords has

In this paper, we study more than 25,000 passwords making up the entire user base of Carnegie Mellon University (CMU). Notably, these passwords are the high-value gatekeeper to most enduser (i.e., non-administrative) online functions within the university, including email, grading systems, transcripts, financial data, health data, payroll, and course content. Furthermore, these passwords were created under a password-composition policy among the stricter of those in common use [18], requiring a minimum of eight characters and four different character classes. Using indirect access to the plaintext of these passwords, we measure their strength. In addition, we obtain contextual information from personnel databases, authentication logs, and a survey about password creation and management, and correlate these factors with password strength. To acquire this data, we established a partnership with the CMU information technology division; the research was also vetted by our Institutional Review Board (IRB). Our approach to analyzing this sensitive data securely provides a blueprint for future research involving security-sensitive data in the wild.

Using this data, we make two important and novel contribu tions to the field of password research. First, we identify interestine trends in password strength, measured as resistance to offline guessing attacks, in which an attacker attempts to recover plaintext passwords from their hashes [2, 6, 44]. Using statistical methods adopted from survival analysis, we find that users associated with science and technology colleges within the university make



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Guess # Depends on Model

We don't think in "cracks," we think in guess numbers!



Password Cracking:

Johnny14! - cracked



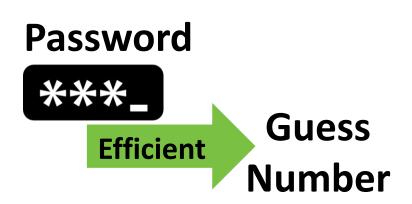
Guess Number:

Depends on "trained" model



Goals For Guess Numbers

- 1. Compute guess numbers efficiently
- 2. Configure guessing method systematically
- 3. Approximate real-world attack







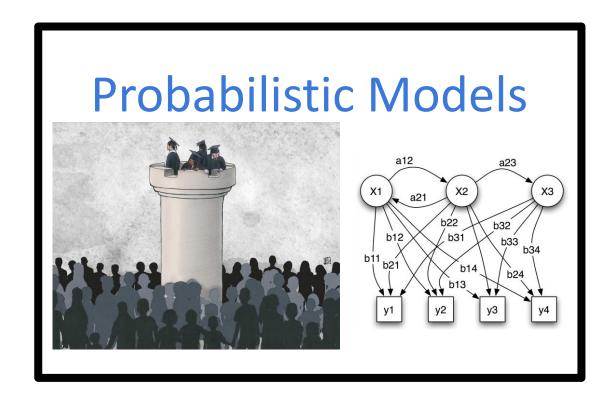


Outline

- State of the art
- 2. How software password-cracking tools work
- 3. Our efficient techniques for guess numbers
- 4. Our techniques for systematic configuration



Password-Cracking Methods







Probabilistic Models



Markov Models [Narayanan and Shmatikov, CCS 2005]

Probabilistic Context-Free Grammars [Weir et al., S&P 2009]

Neural Networks [Melicher et al., USENIX Security 2016]

Guess #
Configuration
Real

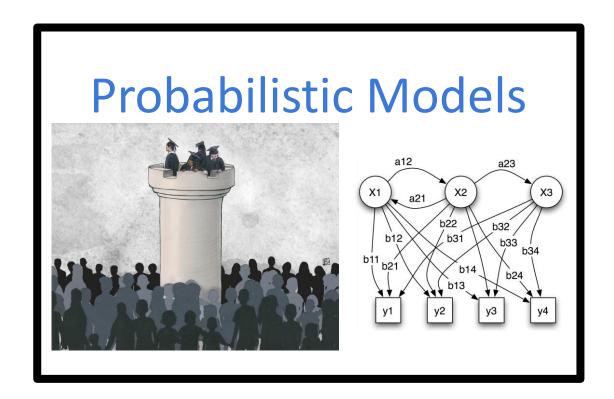








Password-Cracking Methods







Software Tools



John the Ripper

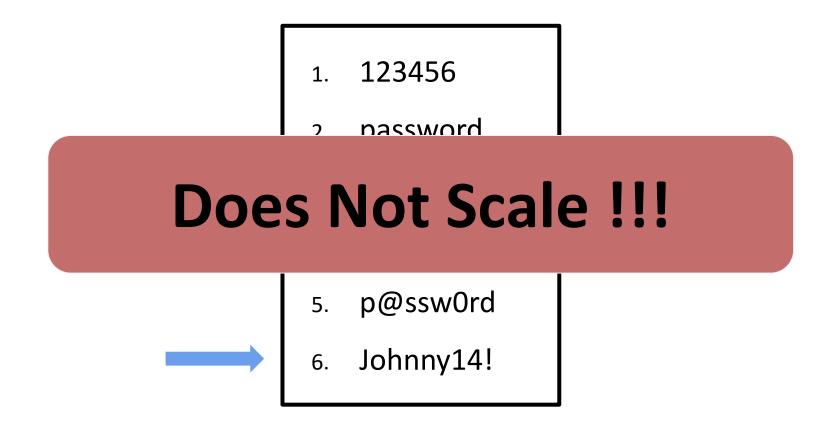


Hashcat





Guess Number by Enumeration





Software Tools



John the Ripper



Hashcat



Guess #
Configuration
Real



Reasoning Analytically About Password-Cracking Software [S&P 2019]

Enze Liu, Amanda Nakanishi, Maximilian Golla[†], David Cash, Blase Ur University of Chicago, † Ruhr University Bochum



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Mangled Wordlist Attack

Wordlist

Super Password Chicago

Rulelist

- 1. Append "1"
- 2. Replace "a" \rightarrow "4"
- 3. Lowercase all



Super1
Password1
Chicago1
Super
P4ssword

Chic4go



Mangled Wordlist Attack

Wordlist

Super Password Chicago

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Super1

Password1

Chicago1

Super

P4ssword

Chic4go

super

password

chicago



Example Wordlists and Rulelists



Wordlist

Rulelist

Linkedin (≈ 60,000,000)

HIBP (≈ 500,000,000)

Korelogic (≈ 5,000)

Megatron (≈ 15,000)

Generated2 (≈ 65,000)

 $10^9 - 10^{15+}$ guesses

+ Professionals' private word/rule lists



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Is This Password in the Guesses?

Chic4go

Guesses

Super1

Password1

Chicago1

Super

P4ssword

Chic4go

super

password

chicago



Is This Password in the Guesses?

Wordlist

Rulelist

Guesses

Super Password

Chicago

- 1. Append "1"
- 2. Replace "a" \rightarrow "4"
- 3. Lowercase all

Super1

Password1

Chicago1

Super

P4ssword

Chic4go

super

password

chicago



Insight

We can work backwards!



Insight

"Rule Reversal"

Marechal (PasswordsCon 2012)
Kacherginsky (PasswordsCon 2013)
and many others



Inversion Process

Rulelist

Password



- Append "1"
 Replace "a" → "4"
 Lowercase all





Chic4go

Inversion Process

Preimages
Rulelist
Password

Chicago
Chic4go

1. Append "1"
2. Replace "a" → "4"
3. Lowercase all



Count Guesses

Wordlist

Super Password Chicago

Rulelist

- 1. Append "1"
- 2. Replace "a" \rightarrow "4"
- 3. Lowercase all

Guesses

Super1
Password1
Chicago1

Super

P4ssword

Chic4go

super

password

chicago



Count Guesses

Wordlist



Rulelist

- 1. Append "1"
- 2. Replace "a" \rightarrow "4"
- 3. Lowercase all

Guesses





Count Guesses

Wordlist

Super Password Chicago

Rulelist

Append "1"
 Replace "a" → "4"
 Lowercase all

Guesses

Super1
Password1
Chicago1
Super
P4ssword
Chic4go
super
password
chicago



Approach

- Invert each password for each rule
 - Identify the first rule, if any, that guesses it
 - Sum guesses made by previous rules
- Count guesses per rule (JtR) / word (Hashcat)
 - Do this once per wordlist / rulelist combo



Why is this non-trivial?



Inverting Passwords

Name	Func tion	Description	Exam- ple Rule	Input Word	Output Word	Note
Nothing	:	do nothing	:	p@ss- W0rd	p@ssW0rd	
Lower- case	I	Lowercase all letters	ı	p@ss- W0rd	p@ssw0rd	
Upper- case	u	Uppercase all letters	u	p@ss- W0rd	P@SSW0RD	
Capital- ize	c	Capitalize the first letter and lower the rest	с	p@ss- W0rd	P@ssw0rd	
Invert Capital- ize	с	Lowercase first found character, uppercase the rest	С	p@ss- W0rd	p@SSW0RD	
Toggle Case	t	Toggle the case of all characters in word.	t	p@ss- W0rd	P@SSw0RD	
Toggle @	TN	Toggle the case of characters at position N	Т3	p@ss- W0rd	p@sSW0rd	*
Reverse	r	Reverse the entire word	r	p@ss- W0rd	dr0Wss@p	
Dupli- cate	d	Duplicate entire word	d	p@ss- W0rd	p@ssW0rdp@ss W0rd	
Dupli- cate N	pΝ	Append duplicated word N times	p2	p@ss- W0rd	p@ssW0rdp@ss W0rdp@ssW0rd	
Reflect	f	Duplicate word reversed	f	p@ss- W0rd	p@ssW0rd- dr0Wss@p	
Rotate Left	{	Rotates the word left.	{	p@ss- W0rd	@ssW0rdp	
Rotate Right	}	Rotates the word right	}	p@ss- W0rd	dp@ssW0r	
Append Charac- ter	\$X	Append character X to end	\$1	p@ss- W0rd	p@ssW0rd1	
Prepend Charac- ter	^X	Prepend character X to front	^1	p@ss- W0rd	1p@ssW0rd	
Truncate left	[Deletes first character	[p@ss- W0rd	@ssW0rd	
Trucate right]	Deletes last character]	p@ss- W0rd	p@assW0r	
Delete @ N	DN	Deletes character at position N	D3	p@ss- W0rd	p@sW0rd	*
Extract range	×NM	Extracts M characters, starting at position N	x04	p@ss- W0rd	p@ss	* #
Omit range	ONM	Deletes M characters, starting at position N	012	p@ss- W0rd	psW0rd	*
Insert @ N	iNX	Inserts character X at position N	i4!	p@ss- W0rd	p@ss!W0rd	*
Over- write @ N	oNX	Overwrites character at position N with X	o3\$	p@ss- W0rd	p@s\$W0rd	*
Truncate @ N	'N	Truncate word at position N	'6	p@ss- W0rd	p@ssW0	*
Replace	sXY	Replace all instances of X with Y	ss\$	p@ss- W0rd	p@\$\$W0rd	
Purge	@X	Purge all instances of X	@s	p@ss- W0rd	p@W0rd	+

Name	Function	Description	Example Rule	Note
Reject less	<n< td=""><td>Reject plains if their length is greater than N</td><td><g< td=""><td>*</td></g<></td></n<>	Reject plains if their length is greater than N	<g< td=""><td>*</td></g<>	*
Reject greater	>N	Reject plains if their length is less or equal to N	>8	*
Reject equal	_N	Reject plains of length not equal to N	_7	*
Reject contain	!X	Reject plains which contain char X	!z	
Reject not contain	/X	Reject plains which do not contain char X	/e	
Reject equal first	(X	Reject plains which do not start with X	(h	
Reject equal last)X	Reject plains which do not end with X)t	
Reject equal at	=NX	Reject plains which do not have char X at position N	=1a	*
Reject contains	%NX	Reject plains which contain char X less than N times	%2a	*
Reject contains	Q	Reject plains where the memory saved matches current word	rMrQ	e.g. for palindrome

Name	Funct ion	Description	Example Rule	Input Word	Output Word	Note
Swap front	k	Swaps first two characters	k	p@ssW0rd	@pssW0rd	
Swap back	K	Swaps last two characters	K	p@ssW0rd	p@ssW0dr	
Swap @ N	*NM	Swaps character at position N with character at position M	*34	p@ssW0rd	p@sWs0rd	*
Bitwise shift left	LN	Bitwise shift left character @ N	L2	p@ssW0rd	p@æsW0rd	*
Bitwise shift right	RN	Bitwise shift right character @ N	R2	p@ssW0rd	p@9sW0rd	*
Ascii increment	+N	Increment character @ N by 1 ascii value	+2	p@ssW0rd	p@tsW0rd	*
Ascii decrement	-N	Decrement character @ N by 1 ascii value	-1	p@ssW0rd	p?ssW0rd	*
Replace N +	.N	Replaces character @ N with value at @ N plus 1	.1	p@ssW0rd	psssW0rd	*
Replace N - 1	,N	Replaces character @ N with value at @ N minus 1	,1	p@ssW0rd	ppssW0rd	*
Duplicate block front	yN	Duplicates first N characters	y2	p@ssW0rd	p@p@ss- W0rd	*
Duplicate block back	YN	Duplicates last N characters	Y2	p@ssW0rd	p@ssW0r- drd	*
Title	E	Lower case the whole line, then upper case the first letter and every letter after a space	E	p@ssW0rd w0rld	P@ssw0rd W0rld	+
Title w/separator	eX	Lower case the whole line, then upper case the first letter and every letter after a custom separator character	e-	p@ssW0rd- w0rld	P@ssw0rd- W0rld	+



Approach to Inverting Passwords

Chic4go

- Represent preimages as ≈ regex
- Few: [{C} {h} {i} {c} {a,4} {g} {o}]
- Many: $4444 \rightarrow [\{a,4\}\{a,4\}\{a,4\}]$
- ("Purge 1" reversed): [{1}* {C} {1}* {h} {1}* {i}
 {1}* {c} {1}* {a,4} {1}* {g} {1}* {o} {1}*]
 - Represent wordlist as trie



Counting Guesses For Each Rule

Wordlist

Rule

Guesses

Super Password Chicago



Reject if no "a"; Replace $a \rightarrow 4$



2

Super Password Chicago



Replace e→ a Reject if no "a"; Replace a→ 4



3



Advantages and Disadvantages

- Method is preferable:
 - Few target passwords
 - Need guess number quickly
- Not preferable:
 - Many target passwords



Fast Guess Number Estimation

LinkedIn + SpiderLabs
$$\equiv 3.01 imes 10^{14}$$
 Guesses

	Enumeration	Our Approach
Size	~ 3 PB	~ 10 GB
Preprocessing	> 2 years	< 1 day
Mean Lookup	???	< 1 second



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Software Tools Depend On

Contents of the wordlist

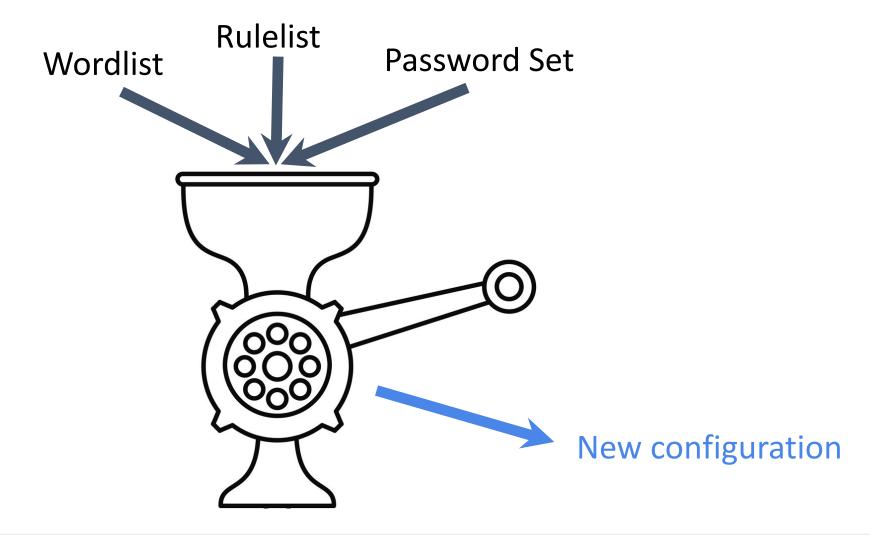
Order of words

Contents of the rulelist

Order of rules



Insight: Data-Driven Configuration





Data-Driven Configuration

Contents of the wordlist

Order of words

Contents of the rulelist

Order of rules



Rule Ordering

Should the rules be in a different order?

Key idea: Order by # cracks per guess

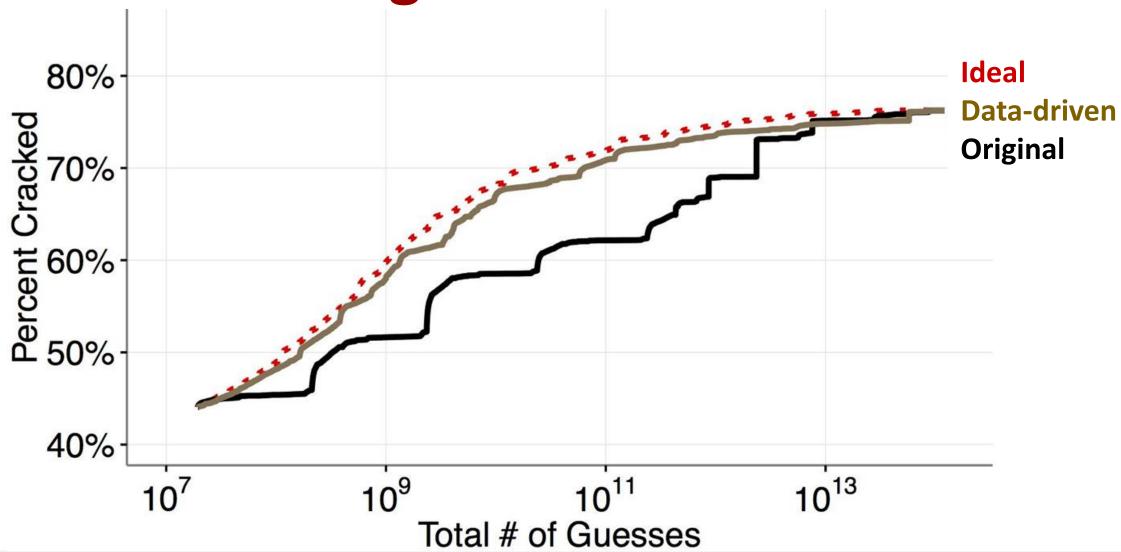
- 1. Append "1"
- 2. Replace "a" \rightarrow "4"
- 3. Lowercase all



- 2. Lowercase all
- Append "1"



Rule Ordering Results

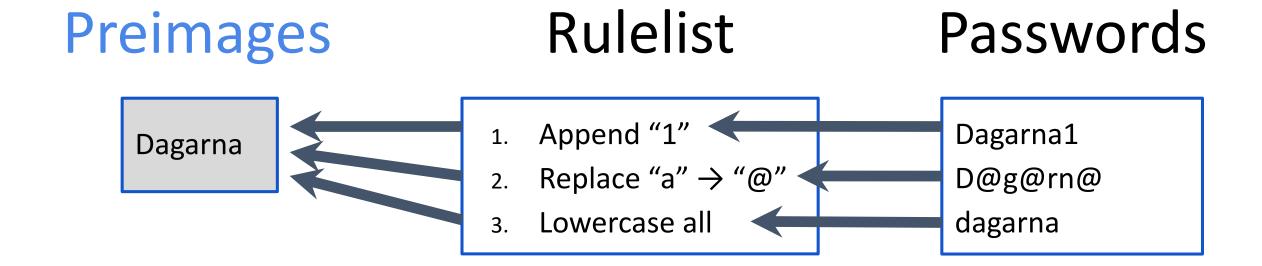




Word Completeness

Should other words be in the wordlist?

Key idea: Add frequent preimage "misses"



Word Completeness (Sample Results)

Category	Examples
Set-specific	bfheros; ilovmyneopets"""







Word Completeness (Sample Results)

Category	Examples
Set-specific	bfheros; ilovmyneopets"""
Meaningful	MaSterBrain; la la la
Short strings	a2; a23; 7a; b2; q2



Takeaway

https://github.com/UChicagoSUPERgroup

Guess Number

Configuration

Analytical Tools

Reasoning Analytically About Password-Cracking Software

Enze "Alex" Liu, Amanda Nakanishi, Maximilian Golla, David Cash, Blase Ur





