

SAuth: protecting user accounts from password database leaks

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Who am I?



- University of Cyprus Computer Science
 - Assistant professor
- SREC (Security Research) group
 - 4 PhD student
 - Several ugrads
 - Research theme is System Security and Privacy
 - More: https://srec.cs.ucy.ac.cy

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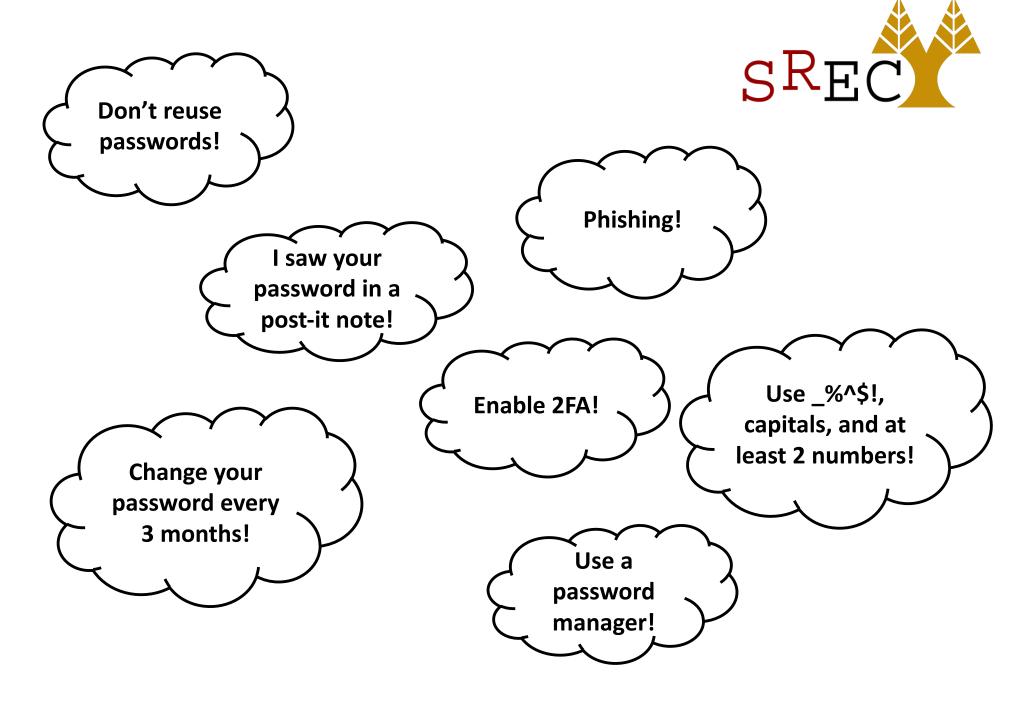




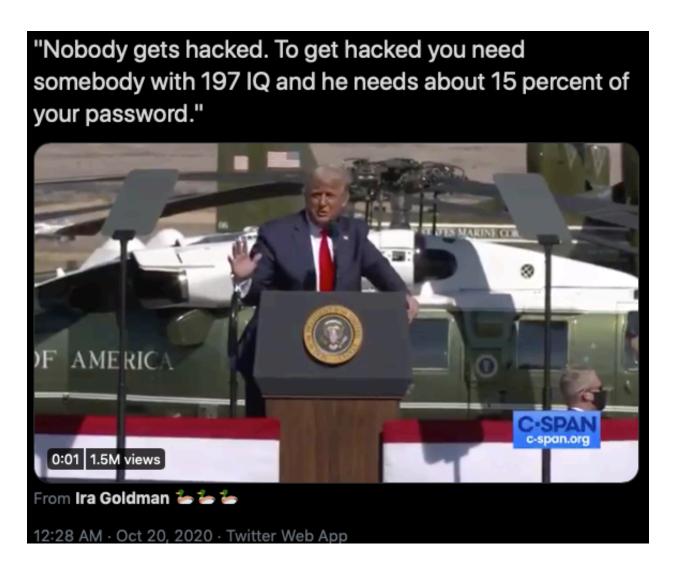




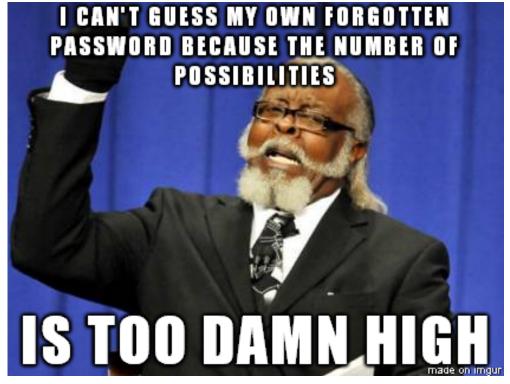
YET, ANOTHER PASSWORDS TALK







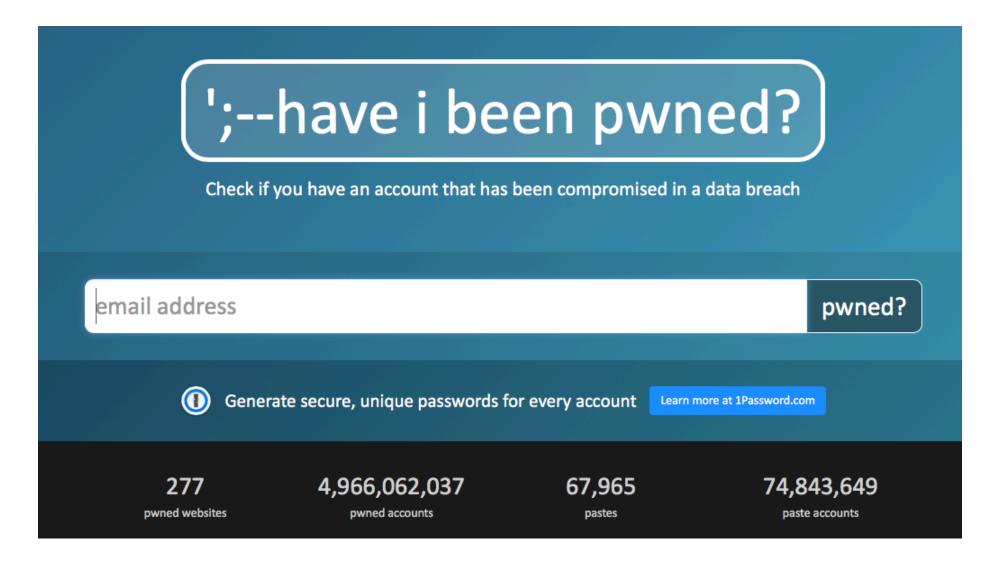




YET, ANOTHER PASSWORDS TALK, BUT NOT A BLAME-THE-USERS TALK

Blame the services!











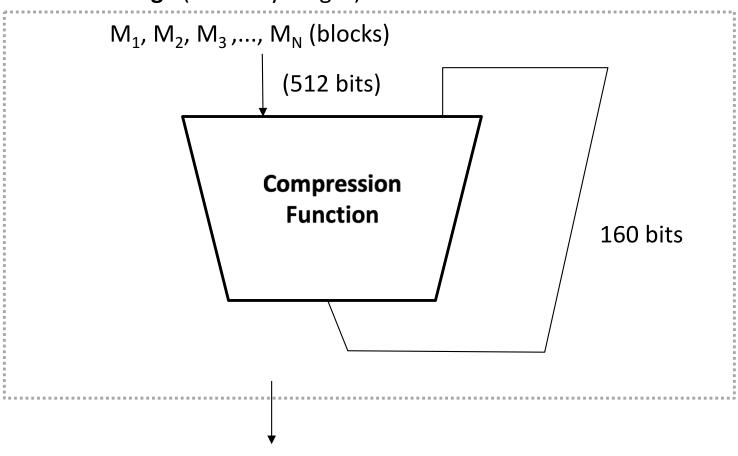


HOW TO PROTECT PASSWORDS?

Cryptographic Hash Function



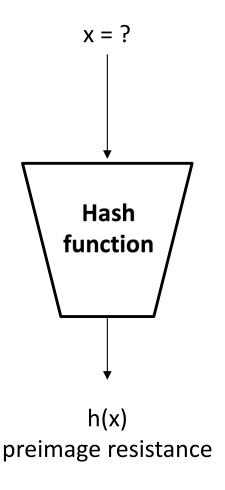
Message (arbitrary length)

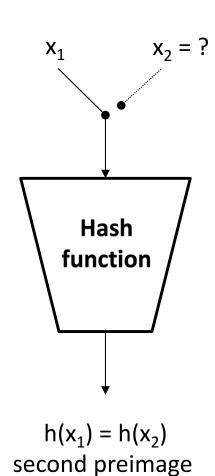


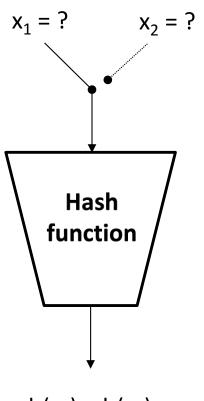
Digest (160 bits)

Basic Requirements









resistance

Threat Model



- Text-based passwords are stored in databases
- They are stored cryptographically hashed
- Databases can be leaked
- Cryptographic hashes can be cracked
- Other Interesting threat models
 - Phishing and social engineering attacks



Can we harden authentication and keep the "password" interface?

SAuth: Synergy-based Enhanced Authentication



• We propose: cooperating sites pool authentication resources

User Agent



Alice

Evernote



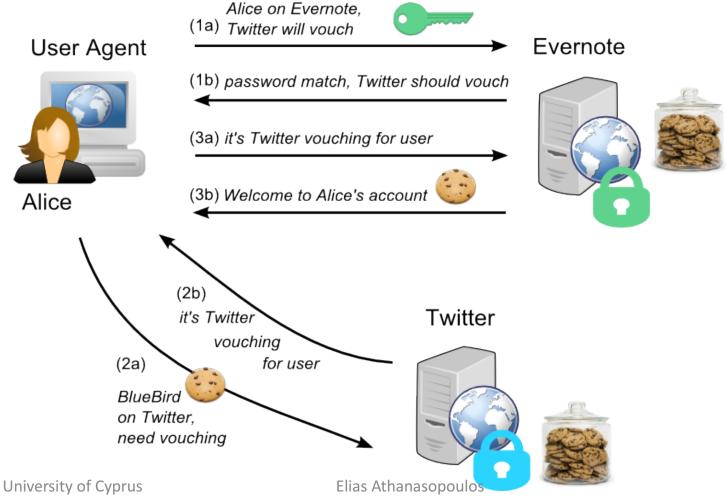
Twitter



SAuth: Synergy-based **Enhanced Authentication**



We propose: cooperating sites pool authentication resources



Why it works?



- Password reminders utilize vouching when the password is lost
- SAuth implicitly initiates such a password reminder on the background
- SAuth does this by pairing arbitrary services

Password Reuse Woes





Stolen passwords re-used to attack Best Buy accounts

Summary: Customer re-use of the same user name and password across multiple sites is being blamed for attacks on customer accounts at BestBuy.com.

Decoy Passwords



- Uncertainty about the actual password
- Store N-1 decoy passwords along
- Attack reduced to online guessing
- All decoys are valid passwords, server does not know the difference

Username	P[0]	P[1]	P[]	P[N]
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- How many decoys?
 - 16,384 for NIST L2 security when password is reused

HoneyGen



 Generate honeywords using Machine Learning (ML) techniques

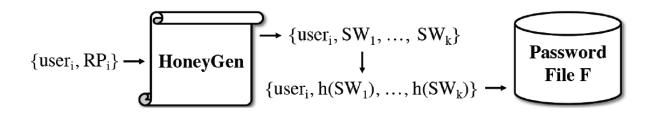


Figure 1: HoneyGen receives i_{th} user's Real Password (RP_i) and responds with an enriched with (k-1) honeywords passwords list containing k sweetwords (SWs) in total. Then, the returned SWs are hashed (according to each operator's hash function h()) and stored in the password file F.

Idea 1:

Random perturbations



- Randomly replace characters from a given password
- Problems:
 - Random generated passwords can be easily distinguished
 - Random generated passwords might not comply with the websites' policies
 - Not generic enough

Idea 2:

Probabilistic Model



 Train a model on each operator's password corpus and return as honeywords the top-k nearest neighbours of a given password

• Benefits:

- Returned passwords match each website's password policies
- This approach accurately models the password selection behaviour of each website
- Realistic looking honeywords (they are actual passwords of other users)

Problems:

- limited honeywords generation spectrum (no new honeywords can be generated apart from those that already exist in each operator's password corpus)
- The model-based approach can be reversed

Evaluation – User Study SREC



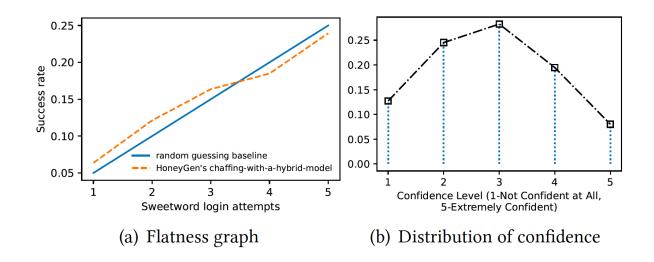


Figure 6: The flatness graph and the distribution of confidence for the user study's collected results for $T_1 = 1$ up to $T_1 = 5$, using chaffing-with-a-hybrid-model.

Take Away



- Passwords are associated with several problems
 - Not always the user's fault
- Combining services with vouching can be done not just for resetting the password
 - Raising the bar for the attackers
- Generating decoy passwords is a hard problem
 - ML can help