

Paper

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1 Introduction

The increasing use of public key cryptography by instant messaging and secure email means key fingerprint verification is an ever more important task. One of the biggest risks to the security of the communication channel is a Man-in-the-middle (MiTM) attack . A successful MiTH attack can circumvent the encryption as it allows an attacker to read all the encrypted data. A counter-measure for this is the verification of each parties' fingerprint. [...]

2 Literature Review

Fingerprints are used to quickly compare the similarity of two keys. The comparison occurs between the digest of a strong one-way hash function. Historically these have been represented as a hexadecimal string and users were tasked with manually comparing fingerprints between two substrates, for example, a monitor screen and a business card. We will refer to this process throughout the paper as the *“authentication ceremony”*.

It has been shown that the average human can only hold 7-digits worth of data in our working memory[1]. This rules out the possibility of comparing full digests. For examples SHA-1 is 40 hex digits (160-bit) and, therefore, will make it difficult for effective comparison. Therefore , if the process is to involve humans interaction, there is a need for schemes that work effectively with our abilities and senses.

Research in this area has taken various encoding schemes and compared their fallibility to impersonated attacks. Impersonation of a key would involve brute force of the encoding schemes through the mass creation of keys. The main elements of comparison have been accuracy of attack detection and time to complete the comparison. In this context we will consider the metrics a measure of security and usability respectively.

Definition of terms

Throughout the paper various terms will be used to define elements of the authentication ceremony relevant to fingerprints verification. In this section well defined definitions will be made. This is to remove ambiguity and underlying associations readers may have for the words chosen.

Encoding scheme - This is the physical method of encoding used to represent the fingerprint. For example “Hexadecimal” or “Words” are a unique way to represent a fingerprint and thus, are encoding schemes.

Method of comparison - This is the way in which the encoding scheme is assessed by the user. The most common example and the one used as default is “Compare-and-Confirm”. This as the name suggests is the *comparison* of two fingerprints on different devices or mediums that are then *compared*. Further examples of these method of comparisons will be explained in later sections.

2.1 Fingerprint representation comparison

Encoding scheme performance

Results from the literature consistently show the effectiveness of language based encodings (Words or Sentences). This was shown in [2], [3] and [4] with the accuracy of these schemes all being above 94.00%. In all cases these were the best schemes in terms of accuracy. The exception to this is the work performed by **Hsiao, et al.**[5] in 2009 with Words achieving an abnormal accuracy of 63.00% (Comparisons of methodology will be discussed in later sections).

Aside from textual representations were graphical schemes. Examples of schemes assessed were: Random Art[6], Flag[7], T-Flag[8], Vash¹, OpenSSH visual host key and Unicorns² to name a few. Accuracy from these schemes were mixed with positive results obtained by **Hsiao, et al.**[5] where their best accuracy was achieved by Random Art (94.00%) and Flag Ext.³(88.00%). On the other hand, the worst accuracy was obtained by Flag(50%). The only other paper assessing graphical representations was the work of **Tan et al.**[3] where they also achieved mixed results with accuracies ranging from 46% to 90%.

In terms of usability of graphical schemes the literature concurred on their ease-of-use. In terms of speed of comparison all schemes consistently obtained some of the quickest comparisons times. This is also supported indirectly from the the comparison results of alternative literature [4][2] where with these in consideration graphical schemes still remain the quickest overall. In terms of research into the performance of graphical schemes, the literature does not contain an

¹<https://github.com/thevash/vash>

²<https://unicornify.pictures/>

³The authors’ own improvement on Flag

extensive review with only two papers including graphical schemes in their comparison. This is, therefore, a candidate for further research.

One unique paper in this research area was the work by **M. Shirvanian *et al.***[9] produced in the context of secure messaging pairing. This paper was unique for a number of reasons. First was to consideration for “remote-vs-proximity” pairing where, this is the first consideration of this aspect found in the literature. Second, was their studies aim into investigating fingerprint security and usability in a end-to-end encryption context. The study compared Numerical, Visual and Auditory representations. This is a standard setup for similar studies in the literature. QR code verification was also added to the proximity tests.

The findings from the paper showed a high false negative rate for all the schemes, this is a consideration also missing from the literature. Alongside this, results for usability were lower in a remote setting for all the schemes where the author comments on the expected nature of this result.

Images were assessed as being the most secure method of authentication in the remote setting, but voted as the method with the worst usability. This is highly inconsistent with all other work in this area. However, this could be due to the unique setting of remote verification resulting in distinct results from user studies. Without further work in this area it is difficult to conclusively validate these results.

To provide a visual comparison the accuracy results of the papers have been provided in Table 2 & 3. Each paper used a unique metric for measuring effectiveness, and, therefore, have been translated into “overall accuracy”.

Methods of comparison performance

As explained in the definition section “methods of comparison” are the ways used to compare fingerprints. Aside from “Compare-and-Confirm” (CaC) there is: “Compare-and-Select” (CaS) and “Compare-and-Enter” (CaE). CaS is the method where one device displays the fingerprint and the other user is provided with a number of options, the user then has to choose the correct value from the list of candidates. If there is no match the user must deny the connection attempt. This was due to concerns that CaC would be “too easy” for users leading to complacency and errors[10]. CaE is designed for scenarios where both devices might not have a display, i.e. pairing between a phone and a keyboard. One device displays the checksum and the other device enters the string. The second device then compares the entered string and checks for a match.

Research has been performed assessing the performance of these schemes and how they affect the ultimate security of the authentication ceremony. The literature agrees on CaC being the best overall scheme to compare fingerprints

[3][10] with CaS being highlighted for its poor security and usability. CaE has had contrasting results. [10] discarded it after one round due to “poor usability”, however, [3] considered it the as the best method overall if usability and security is considered. This, therefore, shows how polarised the results are regarding CaE. This, however, could be due to the different overall use-cases of the studies. Validation of results from either study would, therefore, be an area of extra study.

One aspect of note from the work by **Ersin Uzun *et al.***[10] is their studies context of “Secure device pairing” and thus some comparison schemes are tested that are not relevant in a fingerprint comparison context, for example “Choose-and-Enter” (A participant chooses a passphrase from an available list and enters it into the other device). Thus, only relevant comparison schemes and their results have been extracted and compared.

Experimentation methodology comparison

To further look into the validity of previously discussed results it is necessary to asses how the respective studies reached these conclusions. Areas for consideration are scheme entropy, attacker strength and participant factors among others.

The starkest limitations of the literature are the range of participant and encoding entropy, where at its worst only 22-bits of entropy were tested. This makes it difficult to directly compare results and needs to be taken into consideration when comparing. One of the papers this most affects is the early work by **Hsu-Chun Hsiao, *et al.*** where their highest entropy is 28-bits. This was small even at the time of publication. There is an attempt by the authors’ to address this issue in the later stages of the paper where they write “*[...] increasing entropy is not a solution because it sacrifices usability and accuracy. With more entropy, representations will contain longer sequences of characters or more minute details which will lead to increased time and errors during comparisons.*”. Although the assumptions in this statement could be seen as following common sense they fail to consider the implications of of a small entropy on the security of the system.

This leads into the assessment of attacker strength considerations made by each paper. Some papers failed to directly address their attacker strength consideration, but the overall strength has been inferred from the changes made to their schemes. For example, if they decided to change a single character in a 40-digit (160-bit) SHA-1 hex digest they are indirectly stating that the attacker can control 39-digits (156-bit). To achieve this the attacker would have to compute 2^{156} SHA-1 compressions to find a key match. In the literature this element ranged from 2^{28} to around 2^{242} , however, this has some relation to the size of the encoding schemes used.

In terms of the studies performed they all considered demographical data when

presenting their results. Their average ages were all around ~ 35 years old with the majority of participants being educated with at least a bachelor degree. This was alongside the equal split between male and female.

One consideration of note is that made by Dechand *et al.* [2] where they briefly consider medical conditions such as ADHD and reading or visual disorders and they way they affect the comparison’s effectiveness. They highlight on a slight reduction in accuracy but due to their small sample size they cannot conclusively validate these results. This is an aspect unique to all other literature, further work would be required to produce conclusive results, thus showing a gap in the literature.

One glaring issue with the demographical health of **Ersin Uzun *et al.***[10] study is the use of two entirely different groups of participants. Not only were their demographics different but they were from different countries (America and Finland). This issue is of a concern due to the dual study design of this paper. Changes were made pragmatically regarding the results from the first round of 40 participants. These were then re-assessed and the results were directly compared. This puts huge doubts on the validity of the results with no consideration made by the authors to control external factors that may effect the methods performance.

TODO: Do an overall comparison of the sizes of the studies and link to the table produced

| | Hsu-Chun[5] | Kainda[4] | Dechand[2] | Tan[3] | M. Shirvanian[9] |
|--------------------------------|---------------|---------------|------------|----------|------------------|
| Attacker Strength ⁴ | $\sim 2^{28}$ | $\sim 2^{40}$ | 2^{80} | 2^{60} | $\sim 2^{242}$ |
| Entropy Range | 22-28 bits | 20-40 bits | 122 bits | 128 bits | 160-256 bits |
| No° Participants | 436 | 30 | 1001 | 661 | 25 |

Table 1: Paper attribute comparison

2.2 Fingerprint representation schemes

Another area of research is investigations into the actual physical encodings of the hash digest. This section will briefly discuss the current research available on the creation and security of actual encoding schemes. The actual details of the operation of the schemes are outside the scope of this literature review, therefore, minimal attention will be allocated to these details.

Some of the oldest preliminary work into visual encoding schemes was performed by **Adrian Perrig *et al.***[6]. in the creation of their scheme “Random Art” in 1999. The motivation for creating such a scheme was the perceived flaws in the ways humans verify and compare written information. As mentioned in previous sections visual encoding schemes have been shown to have mixed success, with low security being one of their most alarming flaws. This research

laid the foundation for further work in analysing the security of visual encoding schemes.

Further research into the creation of unique visual hash schemes have been performed by **C Ellison et al.** [7] (Flag), **Yue-Hsun Lin et al.**[8] (T-Flag) and work by **M. Olembo et al.**[11]. Each publication has provided a new way to visually represent a key fingerprint. Alongside the academic literature, there are more informally presented methods of visual fingerprints such as Unicorns⁵ and Robots⁶. This list is by no means exhaustive but is used to depict the amount of research and work invested into graphical hash representations.

One paper of note is the preliminary work performed by **D Loss et al.**[12] in their “*An analysis of the OpenSSH fingerprint visualization algorithm*” where their aim was to spur on further research with their initial findings into the security of the OpenSSH scheme. The authors claim that the use of the algorithm in OpenSSH is only heuristically defined and there is a need for a formal proof of its security.

The paper proposed a number of ways to generate similar fingerprints. The methods proposed were: Naive brute force, Graph Theory, and brute force of a full visual set. They were only able to produce only very basic results and have proposed a large amount of potential further work. Since the paper’s publication in 2009, there seems to have been no research building on the work of the authors. This highlights a current gap in the available literature.

Minimal research has also focused on basic textual fingerprint representations and their respective security. Work by **A. Karole** and **N. Saxena**[13] looked into ways to improve the security of a textual representation. This research aim was to improve the secure device pairing process of comparing two numerical values. The devices used (Nokia 6030b; Mid-range devices at the time of publication) and the SAS compared results in findings that are not directly applicable in a fingerprint comparison context.

A more specific subsection of textual fingerprints is the use of words and sentences to encode hash digests. Some of the first work in this area was produced by **Juola** and **Zimmermann** [14] and their work in generating a word list where phonetic distinctiveness was prioritised. Each word is mapped to a single byte. The unique aspect of the word list is the separation of “even” and “odd” words where “even” byte positions are sample from the even-list and “odd” from the odd-list. This effectively creates two sub-word lists. The maximisation of linguistic distinctiveness of these word lists were maximised through the use of a Genetic algorithm. The paper also includes a study on effective measures of “linguistic distances” of words and provided an in-depth discussion into these areas.

Overall the paper provides a foundation for formalising the creation of effective

⁵<https://unicornify.pictures/>

⁶<https://github.com/e1ven/Robohash>

wordlists. A limitation is the lack of empirical data gathered on the performance. However, this was later evaluated in work by Dechand *et al.* [2] and shown to be an effective encoding scheme.

Other research of note is work by **M. Goodrich *et al.***[15] called *Loud and Clear: Human-Verifiable Authentication Based on Audio*. As the name suggests the authors were researching ways to improve current methods of secure device pairing. The unique aspect of this work is the use of a Text-to-Speech system reading out syntactically correct English sentences. The sentences are based on a MadLibs⁷ where static placeholders were replaced with potential words. The work into a potential wordlist can be seen as an extension to the work performed by Juola and Zimmermann[14] as they aimed to emulate the techniques used in PGPfone. The paper’s finding are limited by the lack of empirical data backing up claims made by the author as the systems performance and security are only theoretically assessed.

Aside from this research, there have been further informal implementations of fingerprint encodings. The first being by **Michael Rogers**⁸. Rogers’ implementation is a program designed to map fingerprints to pseudo-random poems. This implementation was again, empirically evaluated by Dechand *et al.*[2]. Older work by **N. Haller** with the S/KEY[16] shows the implementation of a system designed to represent a hash as a series of six short words. However, this system is designed for a one-time-password purpose and only provides word mappings for basic human usability of the password and not within a fingerprint verification context. Therefore, the wordlist has not been designed with pronounceability in mind.

A very recent implementation of a word list can be found in Pretty Easy Privacy (pEp) implementation of TrustWords⁹. The unique aspect of TrustWords is its mapping of a single word to 16-bits. In comparison to other literature, this is the highest number of bits-per-word seen. Full mappings (no duplication of words) would, therefore, require 2^{16} words in the dictionary and arguably is higher than most users vocabulary. this deviation from the norm has not been currently backed up by research. Moreover the main RFC documentation still remains in a draft stage and states “*It is for further study, what minimal number of words (or entropy) should be required.*”. These aspects clearly highlight on a gap in the current literature.

2.2.1 Topic conclusion

In conclusion to this topic, the current research has primarily focused on the research and creation of visual representations. Research for textual fingerprints is fragmented and incomplete with work Juola and Zimmermann [14] and M.

⁷https://en.wikipedia.org/wiki/Mad_Libs

⁸<https://github.com/akwizgran/basic-english>

⁹<https://tools.ietf.org/html/draft-birk-pep-trustwords-03>

Goodrich *et al.*[15] providing meaningful research to build upon in terms of word a sentence based encodings. The fragmentation of this research leaves room for further work into this topic area. Alongside this, findings from the previous sections research shows that human language based encodings provided the best usability and, therefore, should be a target for further research looking to improve upon their security and usability.

2.3 Fingerprint representation attacks

This area of research studies ways to physically execute attacks on fingerprint encoding schemes. This differs from previously examined work due to papers discussing the performance and fallibility of encoding schemes simulated the attack without consideration for how the attack would be performed. Research in this area is scant, with lots of research attention being directed towards the security of Man-in-the-Middle (MITM) attacks and not the encoding schemes themselves.

Research in 2002 by **Konrad Rieck**[17] is the first formalisation of attacks on fingerprint representations. The paper titled *“Fuzzy Fingerprints Attacking Vulnerabilities in the Human Brain”* aimed to look into ways users check hexadecimal encoded OpenSSH fingerprint representations. The author created an elegant way to ‘weight’ more important chunks of the digest. The bytes furthest to the right and left of the digests provided the highest weight. The weight was the smallest in the centre of the digest. This provides a way to score digests and determine the best partial collisions found. For example with the target fingerprint: **9F23** a partial match **9313** is given a score of 45% even though only two characters were matching. This is due to the weightings.

The paper contains an implementation with a “1.2GHz CPU” being able to obtain 130,000 H/s (With MD5). In comparison to this, a mid-range Intel i5-3320M CPU can today obtain 111,700,000 MD5 H/s. This shows that the results obtained from the paper are significantly outdated. However, even with the low hash rate, the author was able to obtain some promising results. Figure 1 contains the best example used.

```
TARGET: d6:b7:df:31:aa:55:d2:56:9b:32:71:61:24:08:44:87
MATCH:  d6:b7:8f:a6:fa:21:0c:0d:7d:0a:fb:9d:30:90:4a:87
```

Figure 1: Best match obtained after a few minutes of hashing

Overall the paper shows an interesting way to create partial fingerprint matches but is not quantified by any empirical evidence gathered on real world users. This, therefore, highlights on gaps in the coverage of this literature.

The only other relevant research on this topic is the work by **M Shirvanian et al.**[18] and their paper *“Wiretapping via Mimicry: Short Voice Imitation*

Man-in-the-Middle Attacks on Crypto Phones”. Further research in the area of “human voice impersonation” has received lots of attention [19][20][21]. This paper was chosen over other alternatives due to its specific use of encoding schemes in its evaluation.

In this paper, the authors develop a way to impersonate users when authenticating Short-authentication-Strings (SAS) in pairing of Crypto-phones. To achieve this impersonating they propose two methods: “Short voice reordering attack” where an arbitrary SAS string is recreated by re-ordering snippets obtained from eavesdropping a previous connection and “Short voice morphing attacks” whereby the use of previously eavesdropped audio snippets the attacker can morph their own voice to match that of the victim. With these methods, they aimed to attack encodings of Numbers, PGP word list (previously discussed work by Juola and Zimmermann [14]) and MadLib (M. Goodrich *et al.*[15] work also previously discussed). The effectiveness of these attacks were evaluated with a study involving 30 participants.

Results from the paper show the effectiveness of these methods. Compared to the baseline of the attacker’s voice replacing the victim where this performed with a $\sim 18\%$ success rate. Morphing gained an overall success rate of 50.58% and Reordering a very impressive 78.23% success rate. Showing that these attacks provide an improvement on top of the naive implementation.

One of the biggest limitations addressed by the authors was the reduction in success rates as the size of the authentication string grew. The morphing and reordering attacks become increasingly ineffective as the user has more time to detect imperfections. This is not quantified by the author and the extent of this degradation is never empirically discussed. Therefore, the results from this study are only effective and applicable in a SAS context.

2.3.1 Topic Conclusion

Overall the literature for this subtopic remains sparse and incomplete. Further suggested work could look into the feasibility of generating partial collisions for all textual representations alongside quantified effectiveness on users. With the possibility to concentrate on a few selected implementations. The work would aim to focus on the various physical methods used and their feasibility. This is one area the previous literature has failed to cover and has only theoretically quantified attacker strength without consideration for the actual real-world cost of these attacks.

3 Overall Summary

TODO: Create an overall summary of all the gaps identified

4 Research Questions

TODO: - Backup choice of questions up using my previous discussion.

Appendices

A Encoding Scheme Results

| Scheme | Hsu-Chun[5] | Kainda[4] | Dechand[2] | Tan[3] | M. Shirvanian[9] |
|------------------|-------------|-----------|------------|--------|------------------|
| Hexadecimal | | | 89.56% | 79.00% | |
| Numerical | | 100.00% | 93.66% | 65.00% | 97.33% |
| Base32 | 86.00% | 90.00% | 91.50% | | |
| Words | 63.00% | 100.00% | 94.25% | 94.00% | |
| Scentences | | 100.00% | 97.01% | 94.00% | |
| Chinese Symbols | 59.00% | | | | |
| Japanese Symbols | 57.00% | | | | |
| Korean Symbols | 54.00% | | | | |
| Random Art | 94.00% | | | | |
| Flag | 50.00% | | | | |
| T-Flag | 85.00% | | | | |
| Flag Ext. | 88.00% | | | | |
| OpenSSH' | | | | 90.00% | |
| Unicorns | | | | 46.00% | |
| Vash | | | | 88.00% | |

Table 2: Accuracy of correct comparison for the encoding schemes assessed

| Scheme | Hsu-Chun[5] | Kainda[4] | Dechand[2] | Tan[3] |
|------------------|-------------|-----------|------------|--------|
| Hexadecimal | | | 11.20 | 9.00 |
| Numerical | | 6.00 | 10.60 | 9.00 |
| Base32 | 3.51 | 6.00 | 10.20 | |
| Words | 4.63 | 7.00 | 8.70 | 7.00 |
| Scentences | | 11.00 | 12.30 | 8.00 |
| Chinese Symbols | 5.01 | | | |
| Japanese Symbols | 5.07 | | | |
| Korean Symbols | 4.92 | | | |
| Random Art | 3.21 | | | |
| Flag | 4.28 | | | |
| T-Flag | 4.00 | | | |
| Flag Ext. | 4.02 | | | |
| OpenSSH' | | | | 5.00 |
| Unicorns | | | | 3.00 |
| Vash | | | | 3.00 |

Table 3: Timing results in seconds for the related schemes

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