

# Homographic (Homoglyph) Detector

## Objective & Deep Background

The Homoglyph attack, also known as an IDN homograph attack, exploits the visual similarity of Unicode characters to trick users into visiting spoofed domains. For instance, a user might see **example.com**, while the actual link could contain Cyrillic characters like **е** or **а**, which look identical but are distinct code points [Wikipedia](#)[Inspiroz](#). Such deceptive domains can be used in phishing, malware delivery, or impersonation of trusted brands [Mesh | Email Security Redefined for MSPs](#)[Wikipedia](#).

These attacks are subtle, often invisible to the naked eye and standard filters, and can result in stolen credentials, reputational damage, and financial loss [Sven Ruppert](#)[MDPI](#). Organizations like educational institutions are especially vulnerable given their reliance on email and digital workflows [Inspiroz](#).

## Defense Strategies & Research Insights

- **Unicode Normalization (e.g., NFKC):** Converts visually similar characters to canonical forms, reducing spoofing risk [Sven Ruppert](#)[Stack Overflow](#).
- **Homoglyph Mapping Databases:** Tools like Unicode's *confusables.txt* list mappings of visually similar characters for detection [IBM TechXchange Community](#).
- **Machine Learning & Image-Based Models:**
  - **GlyphNet** uses image rendering and CNNs to detect homoglyph domains with high accuracy (AUC  $\approx$  0.93) [arXiv](#).
  - **Siamese Neural Networks** compare visual similarity at the image level for robust detection [arXiv](#).
  - **PhishGAN** synthesizes homoglyph variations to train detection models with augmented datasets [arXiv](#).
- **Automated Scanning Tools:** Tools like *ShamFinder* can auto-generate homoglyph databases and help detect IDN homographs at scale [arXiv](#).

These advanced techniques offer powerful alternatives to simple normalization or string comparison, particularly in high-stakes environments like enterprise security.

## Tools & Libraries Used

- **Python** (Programming Language)
- `unicodedata` (Unicode normalization)
- `difflib` (String comparison)

- re (Regular expression module)
- idna (For domain name handling in Unicode)
- **Top domain whitelist** (e.g., google.com, amazon.com)

## Unicode Homoglyph Examples

Fake Character	Unicode	Looks Like	Legitimate Character
g	U+0261	g	g
o	U+03BF	o	o
c	U+0441	c	c
a	U+0430	a	a
e	U+0435	e	e

## Expanded Code & Explanation

```

1  import unicodedata
2  import difflib
3
4  # 🟩 Whitelist of known safe domains
5  whitelist = [
6      'google.com', 'amazon.com', 'facebook.com', 'microsoft.com', 'youtube.com'
7  ]
8
9  def normalize_domain(domain):
10     """
11     Normalize domain using NFKC form to standardize representations.
12     """
13     return unicodedata.normalize('NFKC', domain)
14
15  def is_suspicious(domain):
16     """
17     Check if normalized domain is suspiciously similar to any safe domain.
18     Returns (is_suspicious: bool, matched_domain: str).
19     """
20     normalized = normalize_domain(domain)
21     for safe in whitelist:
22         ratio = difflib.SequenceMatcher(None, normalized, safe).ratio()
23         if ratio > 0.8 and normalized != safe:
24             return True, safe
25     return False, None
26
27  if __name__ == "__main__":
28     user_input = input("Enter domain to check: ")
29     flag, matched = is_suspicious(user_input)
30     if flag:
31         print(f"⚠️ Domain '{user_input}' looks similar to '{matched}'")
32     else:
33         print("✅ Domain appears safe.")
34

```

Output:

```
e/Desktop/cyber_python/homo.py
Enter domain to check: www.google.com
⚠ Domain 'www.google.com' looks similar to 'google.com'
PS C:\Users\anoop\OneDrive\Desktop\cyber_python> & C:/Users/
e/Desktop/cyber_python/homo.py
Enter domain to check: www.abc.com
✅ Domain appears safe.
PS C:\Users\anoop\OneDrive\Desktop\cyber_python> █
```

## How It Works:

- First, normalize using Unicode NFKC to unify different code points.
- Then use string similarity (e.g., difflib) to compare with whitelisted domains and flag close matches.

## What You Learned

- ◆ How attackers manipulate Unicode for phishing
- ◆ Python tools for domain normalization and comparison
- ◆ Creating simple defensive tools for domain analysis
- ◆ Importance of whitelisting and string matching techniques

## Conclusion

This task introduced practical cybersecurity concerns through homoglyph-based attacks. Using Python and Unicode normalization, a basic but effective detector was implemented to flag misleading URLs. The approach can be extended with machine learning, domain reputation services, or browser plugins for more advanced threat detection.

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