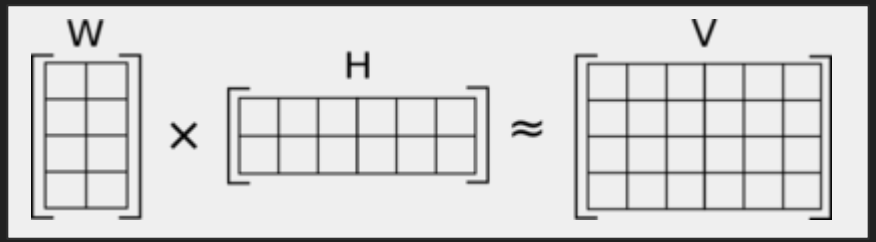
Non-Negative Matrix Factorization

NNMF

What is NMF?

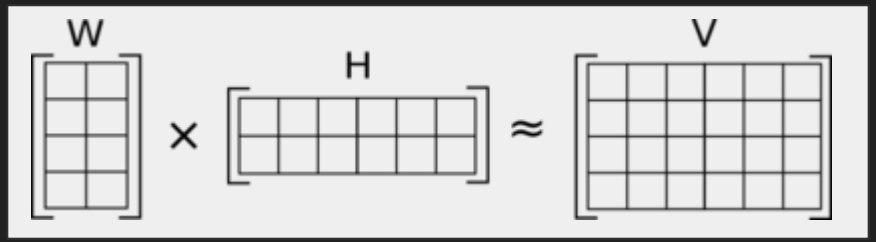
- Non-Negative Matrix Factorization (NMF) - For a matrix V of dimension *m x n* where each element $V_{ij} \geq 0$, the matrix is decomposed into two matrices W and H of dimensions *m x k* and *k x n*, respectively, where each element $w_{ij} \geq 0$ and $h_{ij} \geq 0$ and $k < \min(m, n)$ such that:

$$W \times H = V$$



What is NMF cont...

- V is decomposed into a long matrix W and wide matrix H.
- *k* is specified by the user as long as $k < \min(m, n)$. This specifies the number of clusters the algorithm will create
- Each column of V, or v_i, can be calculated as: $= W \times h_i$



What is NMF cont...

- Lee and Seung's multiplicative update rule

$$H_{cj} \leftarrow H_{cj} \frac{(W^T X)_{cj}}{(WHH^T)_{cj} + eps}$$

$$W_{cj} \leftarrow W_{cj} \frac{(XH^T)_{jc}}{(WHH^T)_{jc} + eps}$$

Why we use NMF?

- The data output we receive is a correlation between Terms and Documents
- NMF breaks down the multivariate data by creating a user-defined number of features. Each one of these features is a combination of the original attribute set. It is also key to remember these coefficients of these linear combinations are non-negative.

Process Overview

Step 1: Select terms

Step 2: Select web documents

Step 3: Use web scraping to count the amount of terms in each document

Step 4: Place web scraped matrix into Matlab NMF algorithm

Step 5: State the amount of iterations and clusters wanted

Step 6: Run NMF algorithm and evaluate data

Web Scrap Implementation

- Written in Python
- Word_count function used to automatically go to each website and count the number of times each term appeared
- The output was formatted to be placed directly into matlab

```
urls = [
"https://en.wikipedia.org/wiki/Bulimia_nervosa"
,"https://en.wikipedia.org/wiki/Narcolepsy"
]
terms = [
"anger"
,"cure"
]
def word_count(url,term):
    html = urllib.request.urlopen(url).read()
    soup = BeautifulSoup(html, "lxml")

    for script in soup(["script", "style"]):
        script.extract()

    # get text
    text = soup.get_text()
    # break into lines and remove leading and trailing space on each
    lines = (line.strip() for line in text.splitlines())
    # break multi-headlines into a line each
    chunks = (phrase.strip() for line in lines for phrase in line.split("  "))
    # drop blank lines
    text = '\n'.join(chunk for chunk in chunks if chunk)

    #print (text)

    count = 0

    text_list = re.sub("[^\w]", " ",  text).split()

    flag = False
    for word in text_list:
        if word == term:
            count+=1
            flag = True
        term.capitalize()
        if word == term and flag == False:
            count+=1
    return count
```

- Our original V matrix created from the web scrape code

```
V = []
for url in urls:
    temp = []
    for term in terms:
        temp.append(word_count(url,term))
        iteration += 1
    V.append(temp)

V = str(V)

V = V.replace(",","")
V = V.replace("]","; \n")
V = V.replace("[","")

print("V =[",V[:-6],"];")
```

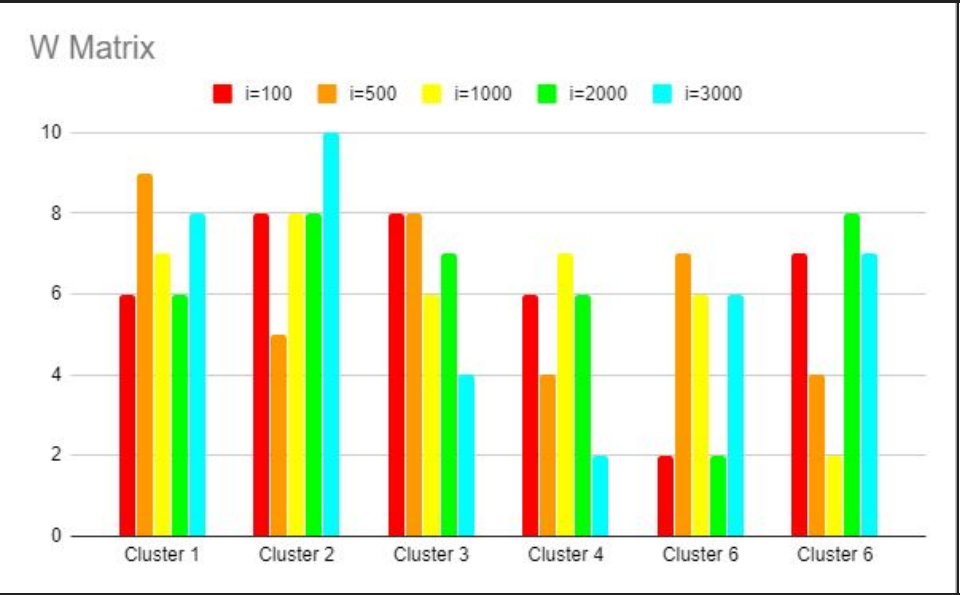
```
V =[ 4 53 1 31 1 2 41 11 11 10 4 182 0 7 1 0 7 77 10 0 1 8 3 7;
 9 29 0 19 0 1 8 28 6 5 0 24 0 1 0 3 4 3 4 2 0 0 0 10;
 13 32 0 13 0 1 16 29 11 5 0 39 0 1 0 0 1 8 2 2 0 0 0 6;
 0 21 0 6 1 0 3 0 9 0 2 5 0 39 0 0 0 0 0 0 0 0 0 3;
 0 9 1 13 1 1 2 2 1 0 1 38 0 2 0 1 0 14 0 0 0 0 0 2;
 0 18 0 9 1 1 15 2 6 0 1 61 0 6 1 3 8 16 1 0 0 0 1 2;
 3 12 1 3 1 0 52 5 1 0 0 33 1 1 2 2 9 41 1 0 9 2 0 16;
 1 1 0 11 3 0 18 0 0 1 0 12 0 3 0 2 12 23 0 0 15 0 2 2;
 2 8 1 1 0 0 57 1 0 0 0 23 0 5 1 1 11 73 1 0 18 3 0 28;
 0 3 1 2 0 0 8 2 0 0 0 11 0 0 1 0 3 19 1 0 1 0 2 3;
 5 33 0 41 0 6 20 4 5 5 0 22 4 13 1 2 4 21 8 1 11 3 1 15;
 1 10 0 2 0 3 62 0 1 0 4 24 3 0 1 3 4 34 4 1 10 0 3 4;
 0 13 1 8 2 0 40 0 1 2 1 27 0 2 0 4 2 26 1 1 1 0 0 3;
 1 1 0 0 0 0 10 1 0 0 1 19 1 1 0 0 1 19 0 0 1 0 2 0;
 3 6 3 24 3 2 9 2 2 4 1 15 1 0 0 0 12 6 1 1 4 0 3 0;
 0 14 1 3 0 0 1 0 1 0 0 2 1 0 0 0 8 1 0 0 2 0 0 1;
 4 12 0 162 0 7 2 4 1 0 0 12 1 0 0 8 2 13 1 1 1 0 7 3;
 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 1 5 0 0 5 0 0 2;
 1 6 0 3 0 0 68 0 4 0 1 34 0 2 0 2 21 50 3 0 13 2 3 17;
 0 7 0 0 0 1 2 0 0 0 0 0 1 0 0 0 1 1 2 0 3 0 0 0 0;
 0 5 0 9 0 0 0 0 0 1 0 2 2 0 0 1 0 0 3 0 0 0 0 2 0;
 1 10 0 52 2 1 8 4 1 4 0 33 0 0 2 3 6 28 7 0 1 8 2 5;
 3 3 0 0 2 0 7 0 0 2 0 28 3 0 0 1 5 35 0 0 0 0 1 2;
 2 5 2 8 0 0 5 3 0 3 2 64 3 0 0 0 5 65 1 0 0 1 0 7;
 0 1 0 21 0 1 3 2 0 1 0 1 0 0 0 0 0 1 2 0 0 4 0 0 1;
 0 1 0 1 1 0 7 0 1 0 0 9 0 0 0 0 0 0 7 0 0 0 0 0 0 1;
 2 8 1 1 0 0 57 1 0 0 0 23 0 5 1 1 11 73 1 0 18 3 0 28;
 0 0 0 0 0 0 54 0 1 0 2 5 0 0 0 0 2 11 1 0 0 1 0 2;
 1 3 0 4 3 0 1 0 1 0 0 6 0 3 0 1 0 7 0 0 0 0 0 0 1;
 0 2 0 1 0 0 8 0 0 0 2 7 0 0 0 0 0 5 15 1 0 5 0 0 4;
 3 16 0 44 0 3 26 7 5 2 36 36 4 12 0 0 0 8 3 0 0 1 1 3;
 0 7 0 10 0 0 6 1 4 0 2 8 0 8 0 0 1 1 0 0 0 0 1 1 0;
 7 54 0 101 2 16 15 20 2 1 4 57 0 40 0 3 4 9 0 0 0 0 0 0 8;
 3 17 0 31 3 1 2 8 1 2 0 26 0 0 0 0 4 4 0 0 1 0 1 15;
 3 10 2 16 2 0 2 19 0 0 1 25 0 0 0 1 14 5 2 2 0 0 0 17;
 1 9 1 1 0 0 21 0 3 1 3 4 0 0 0 2 1 7 0 2 0 0 4 16;
 11 33 0 10 0 2 5 7 2 1 0 9 0 0 0 5 3 3 2 1 0 0 2 25 ];
```

NNMF Implementation

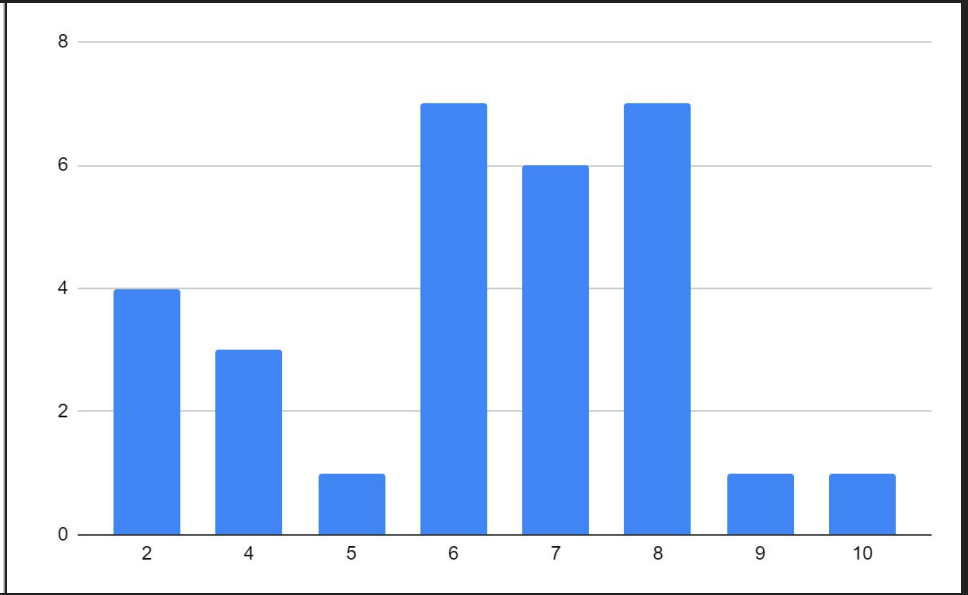
Matlab code

- Rank = clusters
- V = Original matix

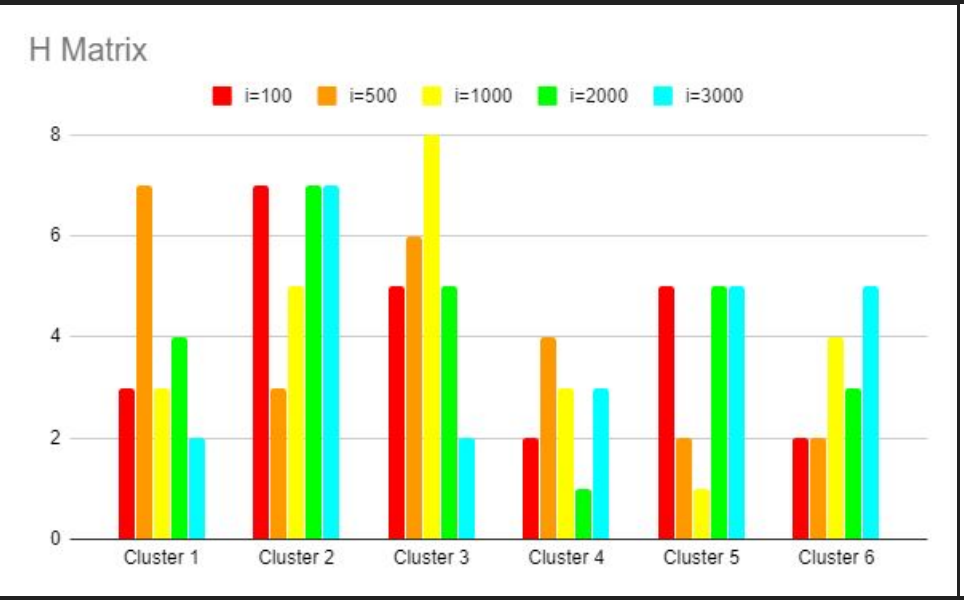
```
V = V + eps;
rank = 6;
iteration = 2000;
for i = 1:iteration
    W = W .* ((V*H') ./ (W*(H*H') + eps));
    H = H .* ((W'*V) ./ ((W'*W)*H) + eps);
end

disp(W)
disp(H)
disp(H')
```



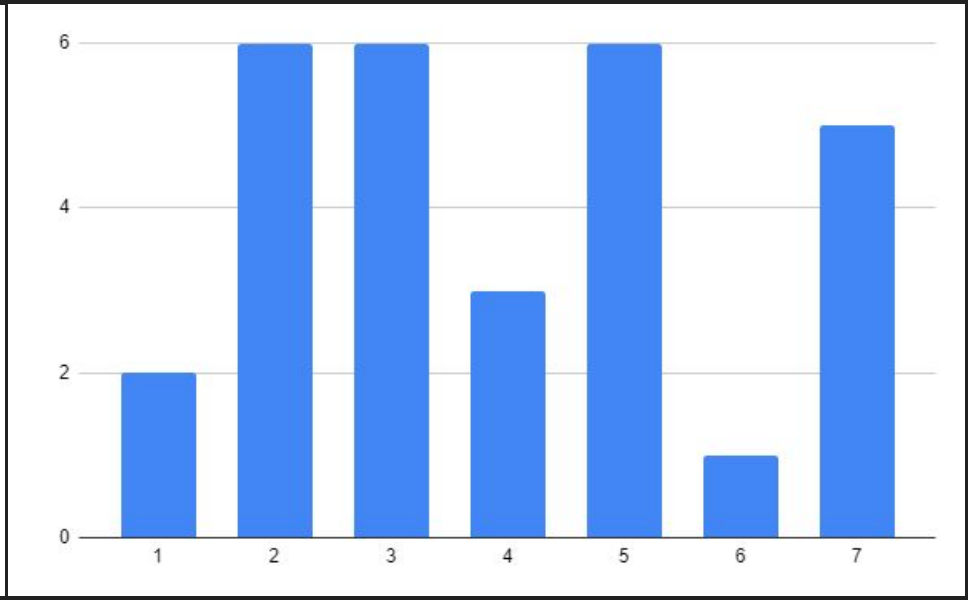
Size of each cluster through different iterations sizes



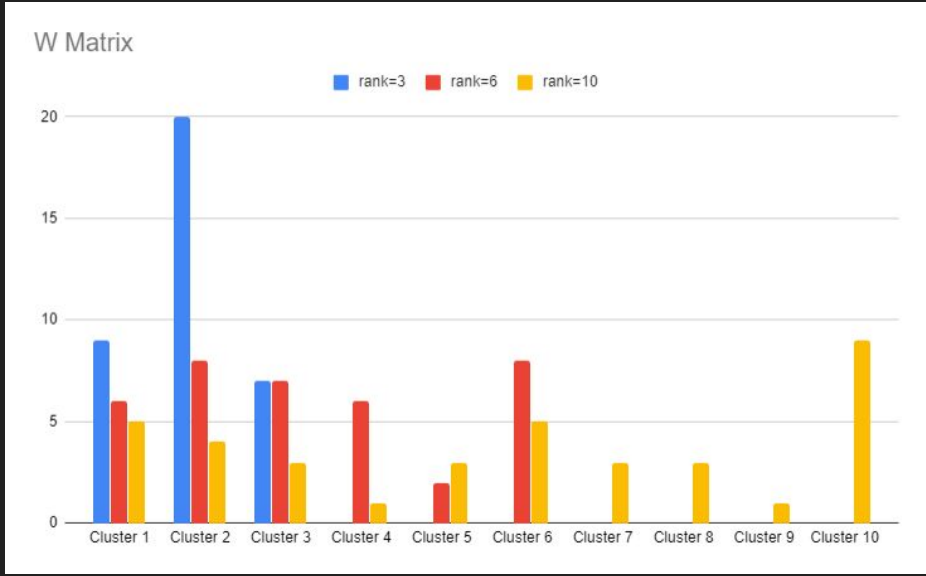
Number of times each cluster size appears



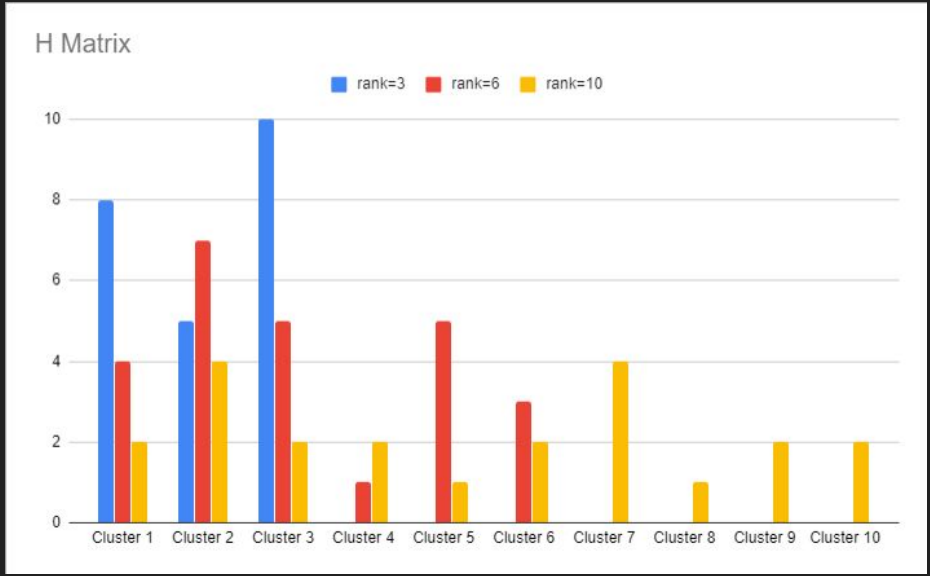
Size of each cluster through different iterations sizes



Number of times each cluster size appears



Size of each cluster through different rank sizes for W matrix



Size of each cluster through different rank sizes for H matrix

W	Cluster 1	Cluster 2	Cluster 3	Cluster 4	Cluster 5	Cluster 6
Computer security	12.3307	0.9584	1.2577	1.7957	0.7825	1.3874
Spetre	0.8178	0	4.0528	0.2786	0	0.8318
Meltdown	1.8938	0	4.0079	0.6997	0.0239	0.4803
Encryption	0	0.0092	0	0	3.4414	0
Password	2.5828	0.0877	0.0666	0	0.1727	0.6603
Internet security	4.0213	0	0.3516	0.707	0.5856	0.3708
Malware	1.3484	2.6026	0.937	2.9164	0	0.0898
Botnet	0.2683	1.7296	0	1.038	0.1105	0.5972
Computer virus	0	5.5234	0.4863	3.138	0.3247	0
Computer worm	0.5145	1.14	0.0771	0.4004	0	0.103
Ransomware	0.468	1.4599	1.9955	0.9914	1.4597	2.0359
Spyware	1.0757	2.003	0	3.6108	0.1513	0.1016
Keystroke logging	1.5108	1.2017	0.112	2.2568	0.3145	0.3899
Trojan horse	1.1273	0.874	0	0.4985	0	0
Phishing	0.7956	0.2497	0.3479	0.5247	0	1.3115
Web-scraping-attack	0	0.0726	0.9182	0.0425	0.2834	0.0983
DDOS Attack	0	0.8989	0	0.0606	0	8.9909
Email spoofing	0	0.4106	0	0.0496	0	0
Antivirus software	1.2621	3.4479	0.1768	3.9187	0.0135	0.1346
Layered Server Provider	0.0001	0.0871	0.3976	0.1	0.1558	0
Doxing	0.0672	0.1426	0.168	0	0.1315	0.4823
Cyberattack	1.8207	1.3584	0.177	0.3335	0	2.8492
Hacker	1.6096	1.818	0	0.2479	0	0
Security Hacker	3.834	3.135	0	0	0	0.3614
Watering Hole Attack	0	0.1836	0.0004	0.1747	0	1.1583
Honeypot	0.5471	0.2868	0	0.3753	0	0.0414
Computer virus	0	5.5234	0.4863	3.138	0.3247	0
Adware	0	0.7135	0	3.1968	0	0
Session hijacking	0.3033	0.3709	0	0.0093	0.2715	0.1924
Redirect	0.2032	1.078	0.0511	0.4404	0.0293	0.0495
Tor	2.0091	0	0.2188	1.46	1.3699	2.2925
HTTPS	0.3827	0	0.051	0.3076	0.7509	0.4743
TLS	2.5097	0	2.4133	0.4773	3.6788	5.1173
SQL Injection	1.2338	0.0421	1.9461	0	0	1.5834
Cross-site scripting	0.9821	0.3002	2.375	0	0	0.7313
Cheating in online games	0	0.6259	1.1102	1.1849	0.0308	0
Buffer overflow	0	0.3263	3.8103	0.1232	0.2349	0.3281

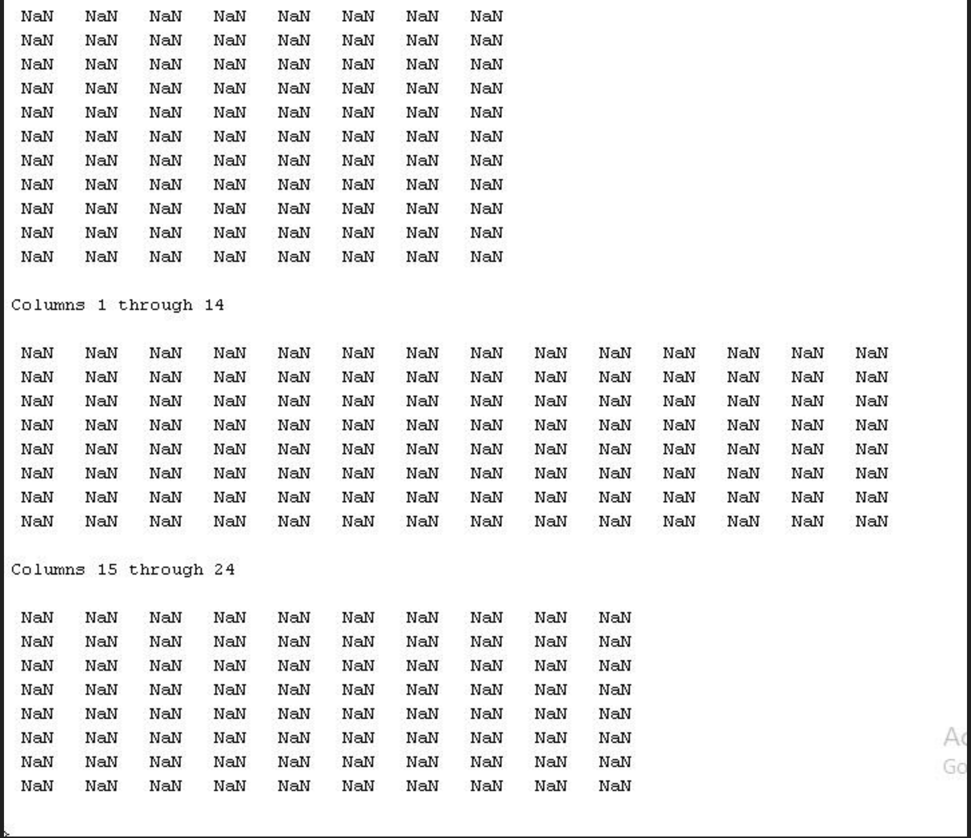
H(Transposed)	Cluster 1	Cluster 2	Cluster 3	Cluster 4	Cluster 5	Cluster 6
exploit	0.0537	0.2263	2.3178	0	0	0.3461
data	2.6251	0	6.7644	0.8686	6.7383	1.2385
trick	0.087	0.1825	0.0644	0	0	0
attack	0.2815	0	0.9546	0	1.6348	18.0063
steal	0.1204	0.0867	0.0939	0	0.1809	0.1168
block	0	0	0.2988	0	1.4109	0.9639
software	0.7147	0.7661	0.6656	16.5448	0.8539	0.0101
vulnerability	0.4855	0	4.9701	0	0	0.6096
protect	0.6243	0	0.9584	0.2307	1.1576	0
hack	0.6525	0.0035	0.6525	0	0	0.0698
privacy	0.2728	0	0	0.6939	1.7246	0.3165
security	13.9645	3.2266	2.7225	0.8723	1.9502	1.1317
illegal	0.0463	0.1853	0	0.0603	0.1675	0.128
encryption	0	0.1641	0	0	10.6539	0
harm	0.0612	0.1439	0.0021	0.0699	0	0.0216
remote	0	0.1147	0.3249	0.324	0	0.6815
malicious	0.3235	1.9517	1.0127	0.8378	0	0.1957
computer	5.0555	13.3991	0	0	0	0
threat	0.5522	0.0146	0.4496	0.4559	0	0.1856
bypass	0	0	0.3998	0.0506	0	0.0282
infected	0	2.4828	0	0.8675	0.1378	0
damage	0.4588	0.4447	0	0	0	0.076
internet	0.0704	0.0055	0	0.358	0	0.5729
code	0	4.0744	3.9624	0.0288	0.2765	0
	4	7	5	1	5	3

Final Output

Cluster 1		Cluster 2		Cluster 3		Cluster 4		Cluster 5		Cluster 6	
W	H	W	H	W	H	W	H	W	H	W	H
Computer Security	Hack	Botnet	Trick	Spectre	Exploit	Malware	Software	Encryption	Steal	Ransomware	Attack
Password	Security	Computer Virus	Illegal	Meltdown	Data	Spyware		HTTPS	Block	Phising	Remote
Internet Security	Threat	Computer Worm	Harm	Web-scaping-attack	Vulnerability	Keystroke Logging			Protect	DDOS Attack	Internet
Trojan Horse	Damage	Email spoofing	Malicious	Layered Server Provider	Hack	Antivirus Software			Privacy	Doxing	
Security Hacker		Hacker	Computer	SQL Injection	Bypass	Adware			Encryption	Cyberattack	
Honeypot		Session hijacking	Infected	Cross-site scripting		Cheating in online games				Watering Hole Attack	
		Redirect	Code	Buffer overflow						Tor	
										TLS	

Pitfalls

- We had many 0's in our original matrix making for NaN output for our W and H matrices
  - Solution: we needed to add eps (epsilon) to our V matrix



```
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN

Columns 1 through 14

  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
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Columns 15 through 24

  NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN   NaN
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```

Works cited

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