

# Application Problems

## Problem 3.7 (Latent Semantic Indexing)

```
load("HW03\HW1_DataSet\wordVecV.mat")
M = V;
M( M < 1 ) = 0;
M( M > 0 ) = 1;
S = vecnorm(M);
Mnorm = V./S;
```

```
[U,S,V] = svd(Mnorm)
```

```
U = 1651x1651
-0.0023 -0.0106 -0.0019 -0.0201 -0.0031 0.0314 0.0015 -0.0186 ...
-0.0047 0.0246 -0.0028 -0.0035 0.0119 0.0016 0.0091 -0.0077
-0.0017 0.0009 -0.0011 0.0026 -0.0114 0.0069 -0.0014 0.0350
-0.0114 -0.0040 -0.0391 0.0053 -0.0562 0.0207 -0.1346 -0.0638
-0.0113 -0.0141 0.0042 0.0453 -0.0448 -0.0235 -0.0399 -0.0542
-0.0023 -0.0106 -0.0019 -0.0201 -0.0031 0.0314 0.0015 -0.0186
-0.0023 -0.0106 -0.0019 -0.0201 -0.0031 0.0314 0.0015 -0.0186
-0.0029 -0.0084 0.0074 0.0042 0.0209 -0.0024 -0.0089 0.0075
-0.0017 0.0009 -0.0011 0.0026 -0.0114 0.0069 -0.0014 0.0350
-0.0045 0.0033 -0.0186 0.0127 -0.0265 -0.0053 -0.0680 -0.0226
:
S = 1651x10
8.8275 0 0 0 0 0 0 0 ...
0 2.4708 0 0 0 0 0 0
0 0 2.0516 0 0 0 0 0
0 0 0 1.9314 0 0 0 0
0 0 0 0 1.7457 0 0 0
0 0 0 0 0 1.6522 0 0
0 0 0 0 0 0 1.5659 0
0 0 0 0 0 0 0 1.4216
0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0
:
V = 10x10
-0.2384 0.0341 -0.0363 0.0792 -0.3099 0.1779 -0.0335 0.7719 ...
-0.2943 0.0604 -0.2810 0.1806 -0.3412 -0.0649 -0.7848 -0.2367
-0.3180 -0.2908 -0.6076 -0.3131 -0.0798 -0.4527 0.2980 0.1123
-0.3067 0.4894 -0.4545 0.3962 0.3190 0.3047 0.2805 -0.1483
-0.2536 -0.3277 -0.0498 -0.4841 -0.0672 0.6483 0.0286 -0.3297
-0.4357 0.5884 0.3470 -0.5081 0.0535 -0.2535 -0.0251 -0.0468
-0.2380 0.0588 0.2250 0.0935 -0.3794 0.3095 0.1982 0.1532
-0.2034 -0.0907 0.2573 0.3079 -0.4041 -0.2829 0.3213 -0.3700
-0.4261 -0.3430 0.2496 0.1325 0.6033 -0.0643 -0.2315 0.1772
-0.3574 -0.2970 0.2047 0.3043 0.0307 -0.0431 0.1343 -0.1042
```

### Part A)

If since matrix  $M$  is the term-by-document matrix and we know that  $U$  and  $V$  are produced from  $MM'$  (document-to-document) and  $M'M$  (term-to-term) matrices we can imagine the vectors of  $U$  to represent how close the documents are together and  $V$  to represent how close the terms are together.

## Part B)

```
k= 3;  
U_k = U(:, 1:k)
```

```
U_k = 1651x3  
-0.0023 -0.0106 -0.0019  
-0.0047 0.0246 -0.0028  
-0.0017 0.0009 -0.0011  
-0.0114 -0.0040 -0.0391  
-0.0113 -0.0141 0.0042  
-0.0023 -0.0106 -0.0019  
-0.0023 -0.0106 -0.0019  
-0.0029 -0.0084 0.0074  
-0.0017 0.0009 -0.0011  
-0.0045 0.0033 -0.0186  
⋮
```

```
% S_k = S(1:k, 1:k);  
% V_k = V(:, 1:k);  
% Mnorm_k = U_k * S_k * V_k';
```

## Part C)

```
largest_elements = maxk(S,:), 10)
```

```
largest_elements = 10x1  
8.8275  
2.4708  
2.0516  
1.9314  
1.7457  
1.6522  
1.5659  
1.4216  
1.3984  
1.2598
```

## Part d)

```
k= 9;  
U_k = U(:, 1:k);  
S_k = S(1:k, 1:k);  
V_k = V(:, 1:k);  
Mnorm_k = U_k * S_k * V_k';  
[angles,i]=pdist2(Mnorm_k',Mnorm_k',"cosine","Smallest",2);  
[min_angle,index] = mink(angles(2,:),1);  
pair = i(2,index);  
display([min_angle,index,pair])
```

```
0.0717 9.0000 10.0000
```

09: Barack Obama

10: George W. Bush

## Part e)

```
for k= 1:8
    U_k = U(:, 1:k);
    S_k = S(1:k, 1:k);
    V_k = V(:, 1:k);
    Mnorm_k = U_k * S_k * V_k';
    [angles,i]=pdist2(Mnorm_k',Mnorm_k',"cosine","Smallest",2);
    [min_angle,index] = mink(angles(2,:),1);
    pair = i(2,index);
    display([k,min_angle,index,pair])
end
```

1	0	1	2
2.0000	0.0000	9.0000	10.0000
3.0000	0.0000	9.0000	10.0000
4.0000	0.0063	9.0000	10.0000
5.0000	0.0178	1.0000	2.0000
6.0000	0.0325	1.0000	2.0000
7.0000	0.0453	1.0000	7.0000
8.0000	0.0536	9.0000	10.0000

The lowest value of k that still gives the same answer is 2.

The pair of similar documents for k-1 is:

01: B. J. Cole

02: Mary J. Blige

## Problem 3.8 (Eigenfaces and l2 projection)

```
load("HW03\HW3_DataSet\yalefaces.mat")
figure;
imshow(M(:, :, 1)/255)
```



$N$  = number of images = 2414

$d$  = number of pixels in image = 1024 = 32\*32

$x^{(i)}$  = a flat vector of an image =  $1 \times d = 1 \times 1024$

$\bar{x}$  = average of all  $x^{(i)}$ s = average face

$\bar{x^{(i)}} = x^{(i)} - \bar{x}$  = centered data

```
Mflat = reshape(M,1024,[])
```

```
Mflat = 1024x2414
```

82	86	76	15	70	53	101	94	118	90	117	120	104	...
81	86	88	16	69	51	101	93	120	93	117	121	105	
72	74	80	22	48	37	79	69	90	84	104	110	97	
72	67	62	11	64	41	82	83	99	87	95	104	96	
40	47	60	23	46	50	54	58	69	55	68	77	62	
93	66	20	8	109	105	98	121	134	126	93	116	130	
119	87	22	7	138	138	116	145	152	161	117	144	159	
129	101	40	7	147	141	126	151	158	166	125	144	162	
135	118	75	6	136	122	138	152	168	160	142	153	163	
121	132	137	4	102	78	146	133	163	125	158	158	141	
:													
:													

```
muX = sum(Mflat')' ./ 2414
```

```
muX = 1024x1
```

62.3637
62.4606
57.5282
53.5638
49.6885
52.2655
59.1877
63.2142
67.4818
73.6794
:
:

```
X = Mflat - muX
```

```
X = 1024x2414
```

19.6363	23.6363	13.6363	-47.3637	7.6363	-9.3637	38.6363	31.6363	...
18.5394	23.5394	25.5394	-46.4606	6.5394	-11.4606	38.5394	30.5394	
14.4718	16.4718	22.4718	-35.5282	-9.5282	-20.5282	21.4718	11.4718	
18.4362	13.4362	8.4362	-42.5638	10.4362	-12.5638	28.4362	29.4362	
-9.6885	-2.6885	10.3115	-26.6885	-3.6885	0.3115	4.3115	8.3115	
40.7345	13.7345	-32.2655	-44.2655	56.7345	52.7345	45.7345	68.7345	
59.8123	27.8123	-37.1877	-52.1877	78.8123	78.8123	56.8123	85.8123	
65.7858	37.7858	-23.2142	-56.2142	83.7858	77.7858	62.7858	87.7858	
67.5182	50.5182	7.5182	-61.4818	68.5182	54.5182	70.5182	84.5182	
47.3206	58.3206	63.3206	-69.6794	28.3206	4.3206	72.3206	59.3206	
:								
:								

```
C = X*X'
```

```
C = 1024x1024
```

```
107 x
```

1.0834	1.0591	0.9640	0.8609	0.7183	0.6881	0.7272	0.7295	...
1.0591	1.0845	0.9946	0.8738	0.7132	0.6738	0.7165	0.7171	
0.9640	0.9946	1.0051	0.8850	0.6793	0.6136	0.6478	0.6394	
0.8609	0.8738	0.8850	0.9035	0.7185	0.6195	0.6352	0.6253	
0.7183	0.7132	0.6793	0.7185	0.7696	0.6859	0.6475	0.6129	
0.6881	0.6738	0.6136	0.6195	0.6859	0.7842	0.7666	0.7132	

0.7272	0.7165	0.6478	0.6352	0.6475	0.7666	0.8676	0.8421
0.7295	0.7171	0.6394	0.6253	0.6129	0.7132	0.8421	0.8858
0.7635	0.7496	0.6727	0.6343	0.5940	0.6802	0.8030	0.8596
0.8256	0.8158	0.7453	0.6828	0.6074	0.6730	0.7616	0.8027
⋮							

## Part a)

Looking at  $X = U\Sigma V'$  and  $C = X' * X$  we can see that

The eigenvalues of  $C$  are the squares of the singular values of  $X$ . This means that if  $\sigma$  is a singular value of  $X$ , then  $\sigma^2$  is an eigenvalue of  $C$ .

Additionally, the left-singular vectors of  $X$  (the columns of  $U$ ) are the same as the eigenvectors of  $C$ .

## Part b)

```
[V, D] = eig(C);
eigenvalues = flip(diag(D))
```

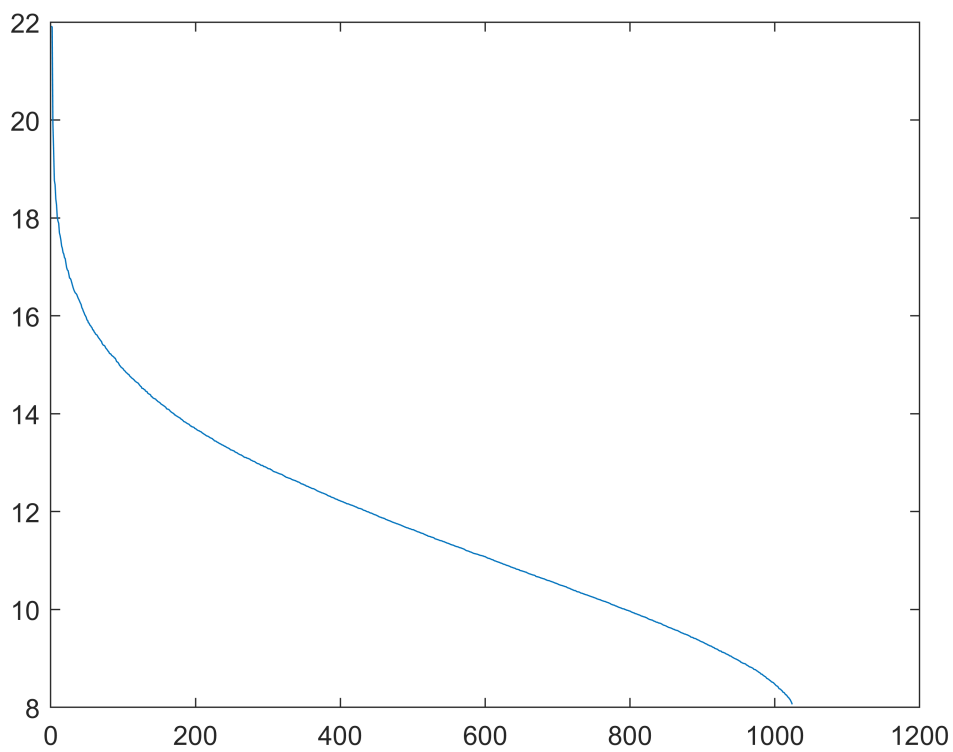
```
eigenvalues = 1024x1
109 ×
    3.3037
    3.2362
    0.4473
    0.2597
    0.1409
    0.1297
    0.0966
    0.0858
    0.0667
    0.0615
    ⋮
```

```
eigenvectors = flip(V)'
```

```
eigenvectors = 1024x1024
    0.0130    0.0432    0.0129   -0.0753   -0.0396   -0.0243   -0.0796   -0.0188 ...
    0.0118    0.0434    0.0098   -0.0746   -0.0409   -0.0270   -0.0892   -0.0162
    0.0091    0.0410    0.0067   -0.0747   -0.0442   -0.0344   -0.0941   -0.0276
    0.0078    0.0384    0.0104   -0.0676   -0.0404   -0.0288   -0.0762   -0.0641
    0.0085    0.0344    0.0227   -0.0480   -0.0341    0.0052   -0.0428   -0.0770
    0.0134    0.0344    0.0347   -0.0369   -0.0391    0.0266   -0.0177   -0.0575
    0.0188    0.0370    0.0440   -0.0313   -0.0463    0.0237   -0.0048   -0.0204
    0.0226    0.0370    0.0424   -0.0264   -0.0375    0.0219    0.0033   -0.0113
    0.0260    0.0381    0.0288   -0.0331   -0.0184    0.0250   -0.0029   -0.0091
    0.0276    0.0410    0.0052   -0.0489    0.0117    0.0221   -0.0305   -0.0145
    ⋮
```

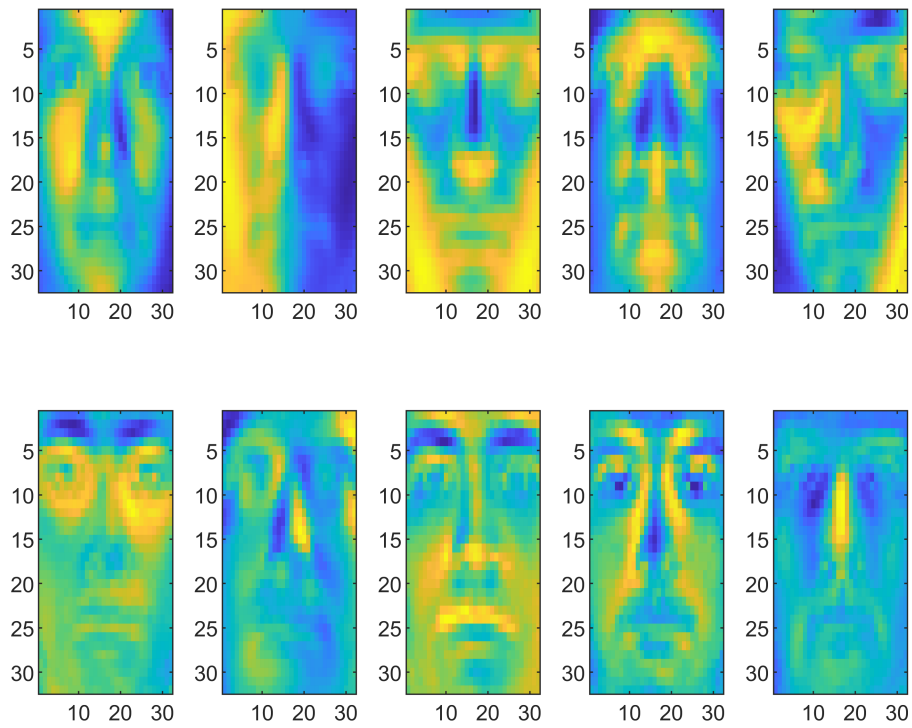
The eigen values are all real.

```
plot(log(eigenvalues))
```



### Part c)

```
L = reshape(eigenvectors, 32,32,[]);  
figure;  
tiledlayout(2,5)  
for i = 1:10  
    nexttile;  
    imagesc(L(:,:,i))  
end
```



```
figure;
tiledlayout(2,5)
for i = 1024-9:1024
    nexttile;
    imagesc(L(:,:,i))
end
```

