

Theory Problems:

Problem 1.1:

Norm Requirements:

1) Positivity: $\|x\|_p \geq 0$, $\|x\|_p = 0$ iff $x = \emptyset$

2) Homogeneity: $\|\alpha x\|_p = |\alpha| \|x\|_p$

3) Sub-additivity: $\|x + y\|_p \leq \|x\|_p + \|y\|_p$

l_1 :

✓ 1) $\|x\|_1 = 0 \Leftrightarrow \sum_{i=1}^n |x_i| = 0 \Leftrightarrow x_i = 0, \forall i$

✓ 2) $\|\alpha x\|_1 = \sum_{i=1}^n |\alpha x_i| = \sum_{i=1}^n |\alpha| \cdot |x_i| = |\alpha| \cdot \underbrace{\sum_{i=1}^n |x_i|}_{\|x\|_1}$
 $\Rightarrow |\alpha| \|x\|_1$

✓ 3) Triangle Inequality: $|a + b| \leq |a| + |b|$

✓ $\|x + y\|_1 = \sum_{i=1}^n |x_i + y_i| \leq \sum_{i=1}^n (|x_i| + |y_i|) = \|x\|_1 + \|y\|_1$

l_∞ :

✓ 1) holds based on definition of " l_∞ "

✓ 2) $\|\alpha x\|_\infty = \max_i |\alpha x_i| = \max_i |\alpha| \cdot |x_i|$

$$\checkmark = |\alpha| \cdot \underbrace{\max_i |x_i|}_{\|x\|_\infty} = |\alpha| \cdot \|x\|_\infty$$

$$\checkmark 3) \|x+y\|_\infty = \max_i |x_i + y_i| = x_j + y_j$$

$$\Rightarrow x_j + y_j \leq \max_i |x_i| + \max_i |y_i|$$

$$\Rightarrow x_j + y_j \leq \|x\|_\infty + \|y\|_\infty$$

↳ jth entry
is the max
entry.

of the vector "x+y"

Problem 1.2:

Exercise 2.6 (Norm inequalities)

1. Show that the following inequalities hold for any vector x :

$$\frac{1}{\sqrt{n}} \|x\|_2 \stackrel{(1)}{\leq} \|x\|_\infty \stackrel{(2)}{\leq} \|x\|_2 \stackrel{(3)}{\leq} \|x\|_1 \stackrel{(4)}{\leq} \sqrt{n} \|x\|_2 \stackrel{(5)}{\leq} n \|x\|_\infty.$$

Hint: use the Cauchy-Schwartz inequality.

2. Show that for any nonzero vector x ,

$$\text{card}(x) \geq \frac{\|x\|_1^2}{\|x\|_2^2},$$

where $\text{card}(x)$ is the *cardinality* of the vector x , defined as the number of nonzero elements in x . Find vectors x for which the lower bound is attained.

$$1) \frac{\|x\|_2}{\sqrt{n}} \leq \|x\|_\infty \Rightarrow \|x\|_2^2 \leq n \cdot \|x\|_\infty^2$$

$$\Rightarrow \sum_{i=1}^n x_i^2 \leq n \cdot \max |x_i|^2 \quad \checkmark$$

$$\checkmark 2) \|x\|_\infty \leq \|x\|_2 \Rightarrow \|x\|_\infty \leq \sqrt{x_1^2 + x_2^2 + \dots} \quad \text{Set } |x_j| = \max_i |x_i|$$

$$\Rightarrow |x_j| \leq \sqrt{|x_j|^2 + \sum \dots}$$

$$\checkmark 3) \|x\|_2 \leq \|x\|_1 \Rightarrow \|x\|_2^2 \leq \|x\|_1^2 \Rightarrow \sum_i x_i^2 \leq \left(\sum_i x_i \right)^2$$

$$4) \|x\|_1 \leq \sqrt{n} \|x\|_2 \Rightarrow \text{did not complete.}$$

4) Continued: Based on Cauchy-Schwarz inequality ($|z^T y| \leq \|y\|_2 \cdot \|z\|_2$)

we can derive $y = (\underbrace{1, \dots, 1}_{n \text{ times}})$ & $z = |x| = (|x_1|, \dots, |x_n|)$

$$\begin{aligned} \Rightarrow z^T y &= \sum z_i y_i = \sum |x_i| = \|x\|_1 \\ \Rightarrow \|y\|_2 &= \sqrt{n} \quad \& \quad \|z\|_2 = \sqrt{\sum |x_i|^2} = \|x\|_2 \end{aligned} \quad \left. \begin{array}{l} |z^T y| \leq \|y\|_2 \cdot \|z\|_2 \\ \therefore \|x\|_1 \leq \sqrt{n} \cdot \|x\|_2 \end{array} \right\}$$

5) Not covered by solutions. However, the proof follows 1)

b) Again based on Cauchy-Schwarz we can assume:

$$z = |x| \quad \& \quad y = \begin{cases} y_i = 0, & x_i = 0 \\ y_i = 1, & x_i \neq 0 \end{cases}$$

$$\Rightarrow \|y\|_2 = \sqrt{k} \quad \text{with "k" being equal to } \text{card}(x).$$

$$\Rightarrow |z^T y| \leq \|y\|_2 \cdot \|z\|_2$$

$$\|x\|_1 \leq \sqrt{k} \|x\|_2 \quad \therefore \quad \frac{\|x\|_1^2}{\|x\|_2^2} \leq k = \text{card}(x) \quad \text{This True for } \forall x$$

with "k" non zero elements of equal magnitude.

Did not Complete Problems 1.3-1.6

Problem 1.3:

a) Based on the linear independence of a_1, \dots, a_k we know that no linear combination of " c_i " vectors could be equal to zero without the multipliers being zero.

b) No conclusion can be made regarding the linear dependence of c_i as no information is provided for b_i and therefore the state of the c_i dependence is unknown.

Problem 1.4:

Since x, y are unit norm. we can derive:

$$(x-y)^T(x+y) = x^T x - y^T y - y^T x + x^T y = x^T x - y^T y = 0$$

For any $z = \lambda x + \mu y$, $\lambda, \mu \in \mathbb{R}$ it can be expressed as:

$$z = \alpha u + \beta v \text{ where } \alpha = \frac{\lambda + \mu}{2}, \beta = \frac{\lambda - \mu}{2} \\ \text{and } u = x + y, v = x - y$$

Since $(x-y)^T(x+y) = 0$ & orthogonal then u & v are orthogonal.

& Since $\text{span}(\{u, v\}) = \text{span}(\{x, y\})$

$$\text{Finally normalize } u, v \Rightarrow \left\{ \frac{u}{\|u\|_2}, \frac{v}{\|v\|_2} \right\} = \left\{ \frac{(x+y)}{\|(x+y)\|_2}, \frac{(x-y)}{\|(x-y)\|_2} \right\}$$

Problem 1.5:

By definition $f(x, y)$ meets all requirements for an inner product except $f(x, x) \geq 0$ & $f(x, x) = 0$ for $x = \phi$. Since if any x_k is equal or smaller than zero any vector x that is a unit vector in \mathbb{R}^n would not pass the condition.

\therefore for any x s.t. $\text{card}(x) = 0$ the function is an inner product

Problem 1.6:

a) using $n=1$, $z_1 = 1$, $z_2 = 10$, $x = -1$. we have:

$$\begin{aligned}\|x - z_1\| &= 2 & \|x - z_2\| &= 9 \\ \angle(x, z_1) &= \frac{\pi}{2} & \angle(x, z_2) &= 0\end{aligned}$$

$$b) \|x - z_j\| \leq \|x - z_i\| \Leftrightarrow \|x - z_j\|^2 \leq \|x - z_i\|^2$$

$$\Leftrightarrow \|x\|^2 - 2x^T z_j + \|z_j\|^2 \leq \|x\|^2 - 2x^T z_i + \|z_i\|^2$$

$$\Leftrightarrow -2x^T z_j \leq -2x^T z_i$$

$$\Leftrightarrow x^T z_j \geq x^T z_i$$

$$\Leftrightarrow \frac{x^T z_j}{\|x\| \|z_j\|} \geq \frac{x^T z_i}{\|x\| \|z_i\|}$$

$$\Leftrightarrow \arccos\left(\frac{x^T z_j}{\|x\| \|z_j\|}\right) \leq \arccos\left(\frac{x^T z_i}{\|x\| \|z_i\|}\right)$$

$$\Leftrightarrow \angle(x, z_j) \leq \angle(x, z_i)$$

% PROBLEM 1.7 PART a

```
[d,i]=pdist2(V',V',"euclidean","Smallest",2);  
[min_distance,index] = mink(d(2,:),1);  
pair = i(2,index);  
display([min_distance,index,pair])
```

24.7184 7.0000 8.0000 ✓

```
[angles,i]=pdist2(V',V',"cosine","Smallest",2);  
[min_angle,index] = mink(angles(2,:),1);  
pair = i(2,index);  
display([min_angle,index,pair])
```

0.1386 9.0000 10.0000 ✓

% PROBLEM 1.7 PART b

```
S = sum(V); % find the sum of all word frequencies for each document  
Vnorm = V./S; % normalize the dataset
```

```
[d,i]=pdist2(Vnorm',Vnorm',"euclidean","Smallest",2);  
[min_distance,index] = mink(d(2,:),1);  
pair = i(2,index);  
display([min_distance,index,pair])
```

0.0643 9.0000 10.0000 ✓

```
[angles,i]=pdist2(Vnorm',Vnorm',"cosine","Smallest",2);  
[min_angle,index] = mink(angles(2,:),1);  
pair = i(2,index);  
display([min_angle,index,pair])
```

0.1386 9.0000 10.0000 ✓

% PROBLEM 1.7 PART c

```
VTF = zeros(size(V));  
for i = 1:10  
    for j = 1:1651  
        VTF(j,i) = tf_idf(j,i,V);  
    end  
end  
[d,i]=pdist2(VTF',VTF',"euclidean","Smallest",2);  
[min_distance,index] = mink(d(2,:),1);  
pair = i(2,index);  
display([min_distance,index,pair])
```

0.0821 9.0000 10.0000 ✓

% PROBLEM 1.7 PART d

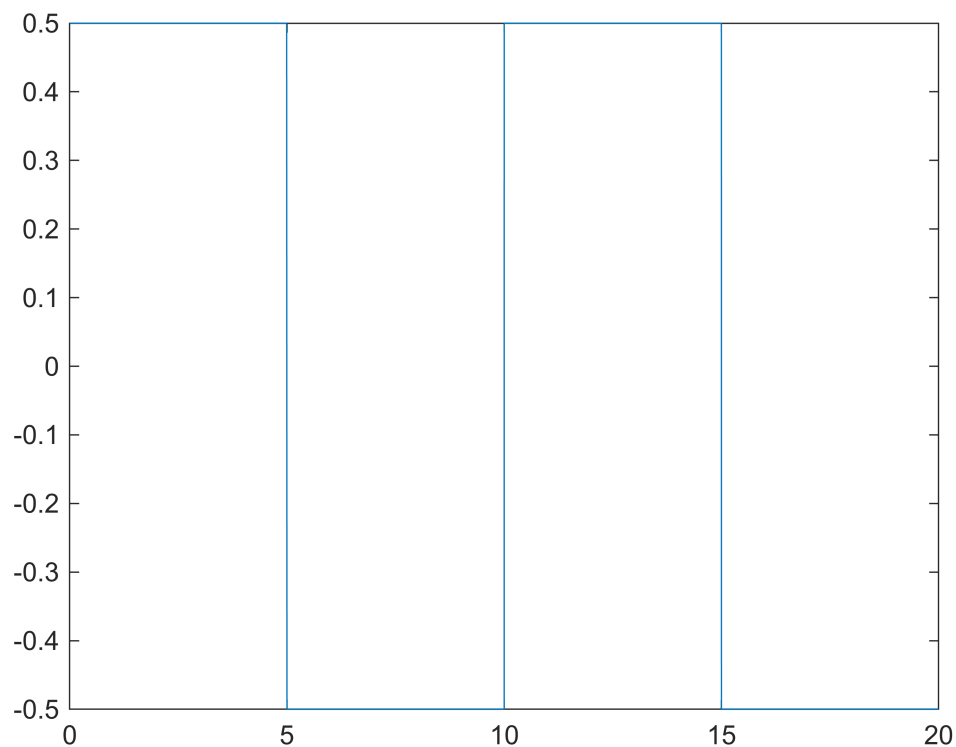
% the adjustment is multiplying the rarity of the a word in a document by
 % the rarity of the word in all the document. This will make all words
 % weighting equal so that words that are very frequent like "the" aren't
 % skewing results due to how often they appear.

% PROBLEM 1.8 PART a

[time_pos , sq_wave , B_unnorm] = generate_data

```
time_pos = 1x200001
0    0.0001    0.0002    0.0003    0.0004    0.0005    0.0006    0.0007 ...
sq_wave = 1x200001
0.5000    0.5000    0.5000    0.5000    0.5000    0.5000    0.5000    0.5000 ...
B_unnorm = 30x200001
0    0.0000    0.0001    0.0001    0.0001    0.0002    0.0002    0.0002 ...
0    0.0001    0.0002    0.0003    0.0004    0.0005    0.0006    0.0007
0    0.0002    0.0003    0.0005    0.0006    0.0008    0.0009    0.0011
0    0.0002    0.0004    0.0007    0.0009    0.0011    0.0013    0.0015
0    0.0003    0.0006    0.0008    0.0011    0.0014    0.0017    0.0020
0    0.0003    0.0007    0.0010    0.0014    0.0017    0.0021    0.0024
0    0.0004    0.0008    0.0012    0.0016    0.0020    0.0025    0.0029
0    0.0005    0.0009    0.0014    0.0019    0.0024    0.0028    0.0033
0    0.0005    0.0011    0.0016    0.0021    0.0027    0.0032    0.0037
0    0.0006    0.0012    0.0018    0.0024    0.0030    0.0036    0.0042
:
:
```

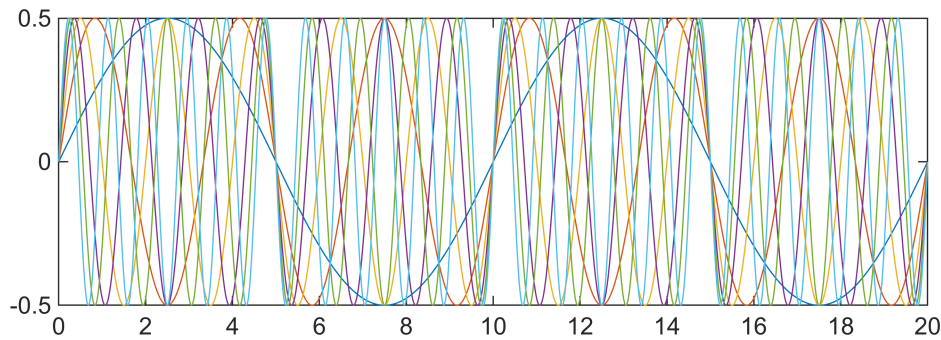
```
subplot(1,1,1);
plot(time_pos,sq_wave)
```



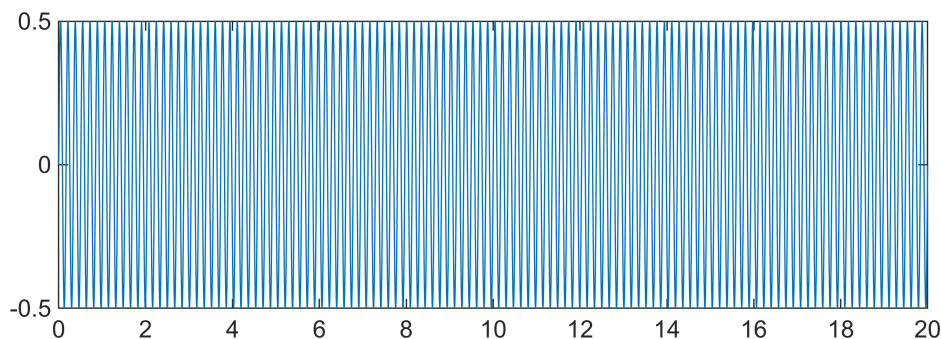

```

% PROBLEM 1.8 PART b
% I would use the gram schmidt procedure to turn the basis into an
% orthonormal set of basis vectors
subplot(2,1,1);
plot(time_pos, B_unnorm(1:6,:))
subplot(2,1,2);
plot(time_pos,B_unnorm(30,:))

```



X wrong display



```

% PROBLEM 1.8 PART c
[q,r]=qr(B_unnorm);

```

X Projection not done.

```

function wd = tf_idf(t,d,V)
    wd = V(t,d) / sum( V(:,d) ) * sqrt( log( 10 / length( nonzeros( V(t,:) ) ) ) );
end
function [ time_pos , sq_wave , B_unnorm ] = generate_data
    n_comps = 30; period = 10; fundFreq = 1/ period;
    time_pos = 0:0.0001:2* period ; harmonics = 2*(1: n_comps ) -1;
    sq_wave = floor (0.9* sin (2* pi * fundFreq * time_pos ) ) +.5; %% generate the signal
    B_unnorm = sin (2* pi * fundFreq *( harmonics' * time_pos ) ) /2; %% generate the basis
end

```