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Unit 1 - Sem 4 - 22MAT230

## Mathematics for Computing

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If you find any mistakes or have any comments to share,

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Definition of Norm to be added

$L_0$  L0 norm

Number of non zero components in  $\mathbf{x}$

```
n = 10;  
x = rand(n,1);  
x(x<0.5) = 0;  
lo_x = sum(x~=0)
```

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```
lo_x =  
4
```

(not strictly a norm function)

### Understanding the $L_0$ Norm: Counting the Non-Zero

The  $L_0$  norm of a vector, denoted as  $\|x\|_0$  is defined as the number of non-zero elements in that vector.

$$\begin{bmatrix} 0 \\ 5 \\ 0 \\ -2 \\ 3 \end{bmatrix}$$

For instance, for the vector  $\mathbf{x} = \begin{bmatrix} 0 \\ 5 \\ 0 \\ -2 \\ 3 \end{bmatrix}$ , the  $L_0$  norm  $\|x\|_0$  is 3, as there are three non-zero elements 5, -2 and 3.

### Mathematical Definition:

For a vector  $\mathbf{x} \in R^n$ , the  $L_0$  norm is given by:

$$\|x\|_0 = \sum_{i=1}^n |x_i|^0$$

where  $|x_i|^0 = 1$  if  $x_i \neq 0$  and 0 if  $x_i = 0$ .

## ***L<sub>1</sub>* L1 norm**

Sum of absolute value of the components of  $\mathbf{x}$

```
n = 5;  
x = randi([-1,4],n,1)
```

```
x = 5x1  
2  
-1  
1  
-1  
4
```

```
l1_x = sum(abs(x))
```

```
l1_x =  
9
```

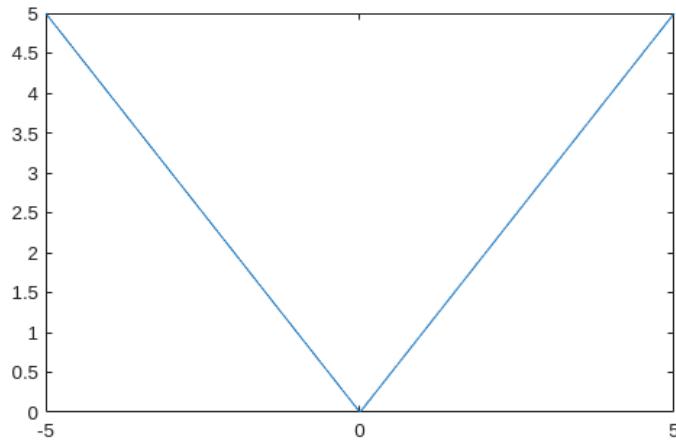
is the sum of the absolute values of each elements of  $\mathbf{x}$ .

$$\|\mathbf{x}\|_1 = \sum_{i=1}^n |x_i|$$

Visualizing  $|x|$

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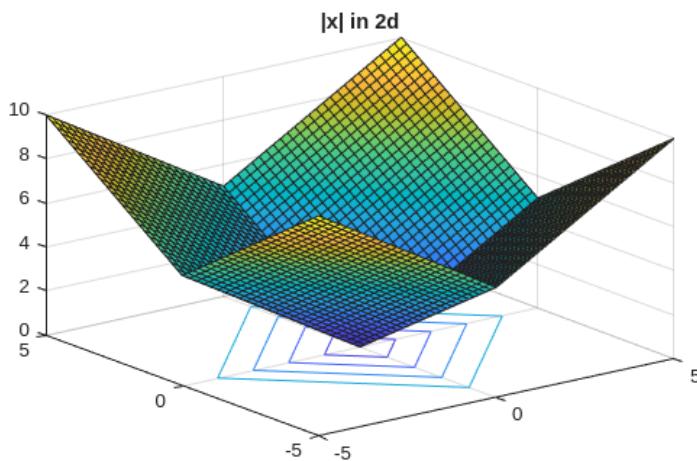
```
x1 = single(-5:0.2:5);
y = abs(x1);
plot(x1,y)
```



```
clearvars
```

Visualizing  $|\mathbf{x}|$ ,  $\mathbf{x} \in R^2$

```
x1 = single(-5:0.2:5);
x2 = x1;
[X1,X2] = meshgrid(x1,x2);
Z1 = abs(X1)+abs(X2);
surf(X1,X2,Z1);hold on
contour(X1,X2,Z1,1:4); hold off
title("|\mathbf{x}| in 2d")
```

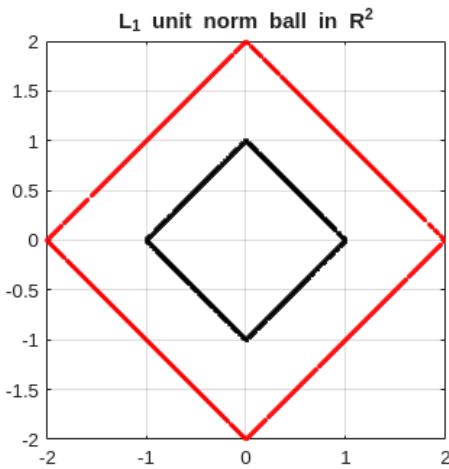


## $L_1$ Unit norm ball in 2 dimension

$$\mathbf{x} \in R^2$$

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```
clearvars
x = rand(2,2500)-0.5;
x = x./sum(abs(x));
x2 = 2*x;
plot(x(1,:),x(2,:),'k.');//hold on
plot(x2(1,:),x2(2,:),'r.');//hold off
axis equal
title("L_1 unit norm ball in R^2")
grid on
xlim([-2 2])
ylim([-2 2])
```

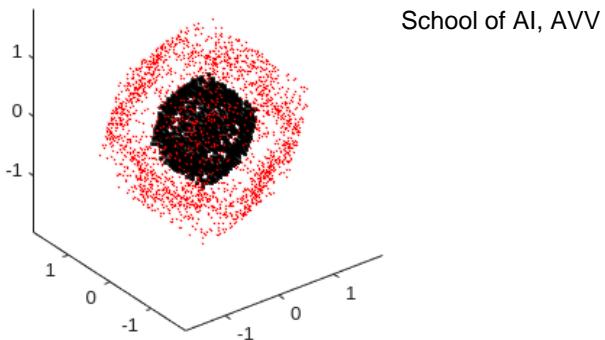


## $L_1$ Unit norm ball in 3 dimension

$$\mathbf{x} \in R^3$$

```
clearvars
x = rand(3,2500)-0.5;
x = x./sum(abs(x));
plot3(x(1,:),x(2,:),x(3,:),'k.');//hold on
axis equal
title("L_1 unit norm ball in R^3")

x2 = 2*x;
plot3(x2(1,:),x2(2,:),x2(3,:),'r.',MarkerSize=0.5);
hold off
```



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## Weighted $L_1$ L1 norm

```
n = 5;
x = randi([-1,4],n,1)
```

```
x = 5x1
 0
 -1
 3
 1
 2
```

```
w = rand(n,1);
w = w/sum(w)
```

```
w =
 0.3359
 0.0987
 0.2185
 0.2706
 0.0763
```

```
wl1_x = w' *abs(x)
```

```
wl1_x =
 1.1774
```

## $L_2$ L2 norm

Sqrt of sum of squares of the components of  $\mathbf{x}$

```
n = 5;
x = randi([-1,4],n,1)
```

```
x = 5x1
```

3  
-1  
1  
0  
3

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```
l2_x = sqrt(sum(x.*x))
```

```
l2_x =  
4.4721
```

```
l2_x = sqrt(sum(x.^2))
```

```
l2_x =  
4.4721
```

```
l2_x = sqrt(dot(x,x))
```

```
l2_x =  
4.4721
```

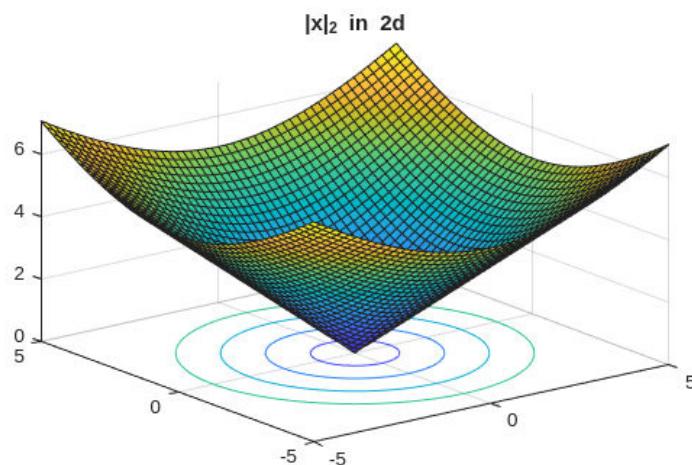
is the square root of the sum of the squares of each values of  $\mathbf{x}$ .

$\|\mathbf{x}\|_2^2 = \sum_{i=1}^n x_i^2$  square of the  $L_2$  norm

$L_2$  norm =  $\sqrt{\sum_{i=1}^n x_i^2}$

Visualizing  $|\mathbf{x}|_2, \mathbf{x} \in R^2$

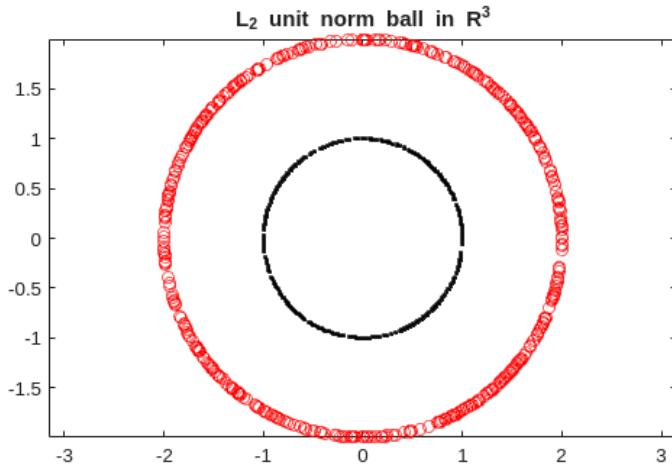
```
x1 = single(-5:0.2:5);  
x2 = x1;  
[X1,X2] = meshgrid(x1,x2);  
Z1 = sqrt(X1.^2 + X2.^2);  
surf(X1,X2,Z1);hold on  
contour(X1,X2,Z1,1:4); hold off  
title("|\mathbf{x}|_2 in 2d")
```



## $L_2$ Unit norm ball in 2 dimension

```
x = rand(2,500)-0.5;
x = x./sqrt(sum(x.^2));
plot(x(1,:),x(2,:),'k.');?>
axis equal
title("L_2 unit norm ball in R^3")
x = 2*x./sqrt(sum(x.^2));
plot(x(1,:),x(2,:),'ro');hold off
```

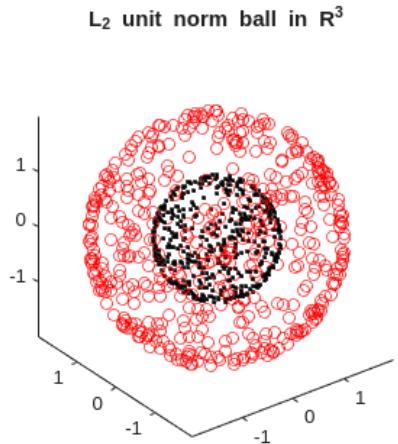
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## $L_2$ Unit norm ball in 3 dimension

```
x = rand(3,500)-0.5;
x = x./sqrt(sum(x.^2));

plot3(x(1,:),x(2,:),x(3,:),'k.');?>
axis equal
title("L_2 unit norm ball in R^3")
x = 2*x./sqrt(sum(x.^2));
plot3(x(1,:),x(2,:),x(3,:),'ro');hold off
```



## $L_\infty$ norm

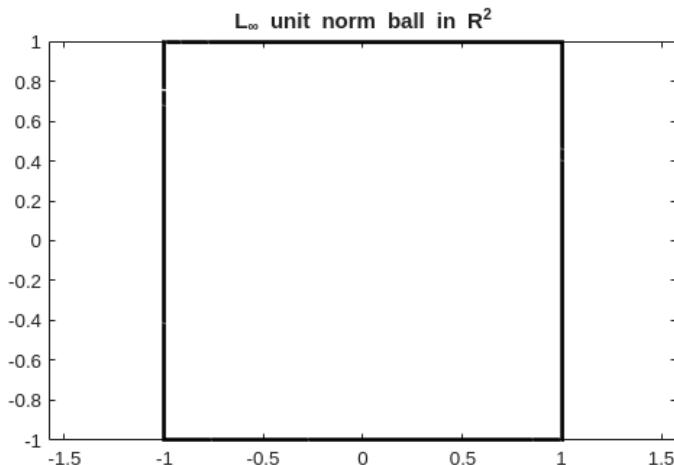
is the maximum of the absolute values of  $x$ .

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$$\|x\|_\infty = \max \{|x_i|\}$$

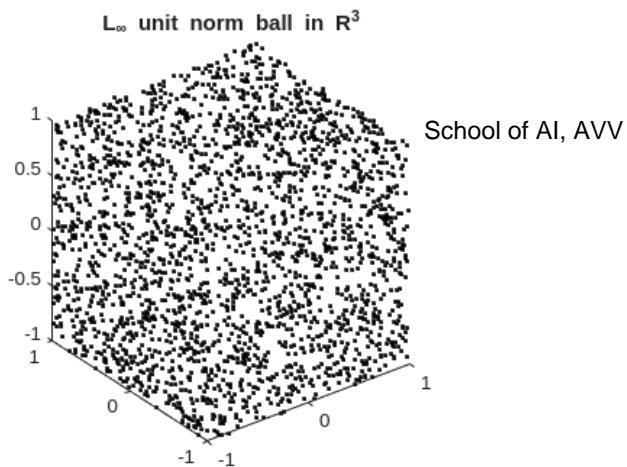
$$x \in R^2$$

```
x = rand(2,2500)-0.5;
x = x./ (max(abs(x)));
plot(x(1,:),x(2,:),'k.');
axis equal
title("L_\infty unit norm ball in R^2")
```



$$x \in R^3$$

```
x = rand(3,2500)-0.5;
x = x./ (max(abs(x)));
plot3(x(1,:),x(2,:),x(3,:),'k.')
axis equal
title("L_\infty unit norm ball in R^3")
```



## L<sub>p</sub> norm

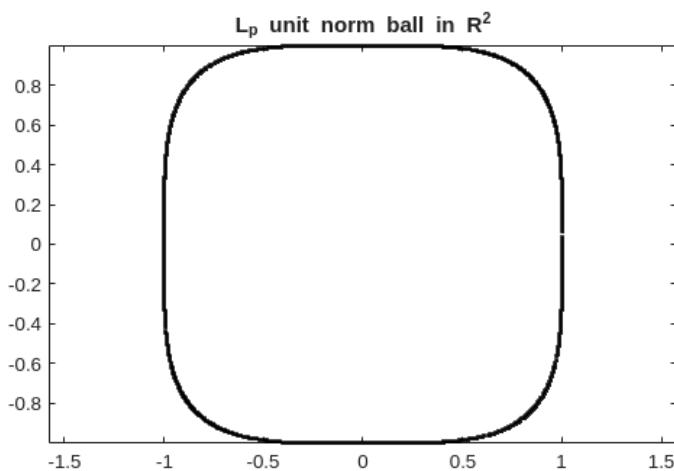
is defined as

$$\|\mathbf{x}\|_p = \left( \sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}}$$

with  $1 \leq p < \infty$

$\mathbf{x} \in R^2$

```
x = rand(2,2500)-0.5;
p = 4;
x = x./power(sum(abs(x).^p),1/p);
plot(x(1,:),x(2,:),'k.')
axis equal
title("L_p unit norm ball in R^2")
```

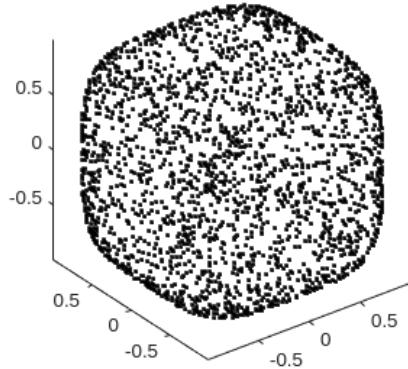


$\mathbf{x} \in R^3$

```
x = rand(3,2500)-0.5;
p = 4;
x = x./power(sum(abs(x).^p),1/p);
plot3(x(1,:),x(2,:),x(3,:),'k.');
axis equal
title("L_p unit norm ball in R^3")
```

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L<sub>p</sub> unit norm ball in R<sup>3</sup>



## L\* Trace norm, Schatten 1-norm, Nuclear norm

$\ell_1$  norm of the vector of singular values of a matrix A

sum of singular values

```
rng(100)
X = randi([-1,4],3,4);

Ln_X = sum(svd(x))
```

```
Ln_X =
102.7801
```

## Frobenius norm

$\ell_2$  norm of the vector of singular values of a matrix A

```
rng(100)

A = randi([-1,3],3,5)
```

```
A = 3x5
 1     3     2     1    -1
 0    -1     3     3    -1
 1    -1    -1     0     0
```

```
norm(A, "fro")
```

```
ans =  
6.2450
```

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```
sqrt(sum(A.^2, "all"))
```

```
ans =  
6.2450
```

```
sqrt(dot(A(:, ), A(:, )) )
```

```
ans =  
6.2450
```

```
sqrt(sum(svd(A).^2))
```

```
ans =  
6.2450
```

```
sqrt(trace(A' *A))
```

```
ans =  
6.2450
```

```
sqrt(trace(A*A'))
```

```
ans =  
6.2450
```

## L<sub>2,1</sub> L21norm

Sum of the  $L_2$  norm of the column vectors of  $A$

```
rng(100)
```

```
A = randi([-1,3], 3, 5)
```

```
A = 3x5  
1 3 2 1 -1  
0 -1 3 3 -1  
1 -1 -1 0 0
```

```
sum(sqrt(sum(A.^2)))
```

```
ans =  
13.0490
```

## Matrix inner product

The inner product of 2  $m \times n$  matrices  $A$  and  $B$  is defined as

$\text{trace}(A^T B)$ . It is equal to the dot product of the vectorized form of  $A$  and  $B$ .

```
rng(100)
```

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```
A = randi([-1,3],4,2)
```

```
A = 4x2
```

1	-1
0	-1
1	2
3	3

```
B = randi([-1,3],4,2)
```

```
B = 4x2
```

-1	-1
1	-1
3	0
0	3

```
trace(A' *B)
```

```
ans =
13
```

```
trace(B' *A)
```

```
ans =
13
```

```
trace(A*B')
```

```
ans =
13
```

```
trace(B*A')
```

```
ans =
13
```

```
dot(A(:),B(:))
```

```
ans =
13
```

## Operator Norm

Maximum value of the singular value of a matrix  $A$

## Weighted $L_{2,1}$ norm

```
A = repmat(1:5,3,1)
```

```
A = 3x5
    1     2     3     4     5
    1     2     3     4     5
    1     2     3     4     5
```

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```
w = rand(5,1)
```

```
w = 5x1
0.8117
0.1719
0.8162
0.2741
0.4317
```

```
L21 = sum(sqrt(sum(A.^2)))
```

```
L21 =
25.9808
```

```
x2 = sqrt(sum(A.^2))'
```

```
x2 = 5x1
1.7321
3.4641
5.1962
6.9282
8.6603
```

```
WL21 = w' *x2
```

```
WL21 =
11.8802
```

```
sum(w.*x2)
```

```
ans =
11.8802
```

*Definition 1 (Weighted L<sub>1</sub>-Norm):* For matrix  $X \in \mathbb{R}^{m \times n}$ , the weighted  $L_1$ -norm of  $X$  is defined as follows:

$$\|X\|_{w_n,1} = \sum_{i=1}^m \sum_{j=1}^n w_{n,i,j} |X_{i,j}|, \quad \text{School of AI, AVV}$$

where  $w_n \in \mathbb{R}^{m \times n}$  denotes the norm weight. The weighted  $L_1$ -norm is calculated by multiplying the absolute value of  $X$  by its corresponding norm weight and then summing all the results.

*Definition 2 (Weighted L<sub>2,1</sub>-Norm):* For matrix  $X \in \mathbb{R}^{m \times n}$ , the weighted  $L_{2,1}$ -norm of  $X$  is defined as follows:

$$\|X\|_{w_g,1} = \sum_{j=1}^n w_{g_j} \|X^{[j]}\|_2, \quad (2)$$

where  $w_n \in \mathbb{R}^{m \times n}$  is the group norm weight, and  $g_j$  denotes the group corresponding to the  $j$ th column. The weighted  $L_{2,1}$ -norm is calculated by multiplying the  $L_2$ -norm of each group by its corresponding group norm weight, and then summing these values.

Team C14

```
cd( "/media/user/DATA4LINUX/new1/Repos/Mine/MFC4_22MAT230/" )
mlxfile = matlab.desktop.editor.getActive().Filename;
outfile = mlxfile + ".pdf";
export(matlab.desktop.editor.getActive().Filename, outfile);
```