## Advanced Engineering Mathematics

Session 2

## Linear Space

- A linear space over the field F, is a nonempty set S of vectors with a binary operation addition + : SxS->S and a scalar multiplication . : FxS->S such that:
  - (S,+) is an Abeliab group.
  - $(\alpha + \beta).x = \alpha.x + \beta.x, \alpha.(x+y) = \alpha.x + \alpha.y, \\ \alpha.(\beta.x) = (\alpha\beta).x, 1.x = x$

(x and y are vectors whereas  $\alpha$ , $\beta$  are scalars)

## Linear Space

- The set F<sup>n</sup> of n-tuples of scalars form a linear space over F.
- The set F<sup>x</sup> of functions f: X->F is a linear space, where x is a non-empty set and

$$-(f_1+f_2)(x)=f_1(x)+f_2(x),$$

 $-(\alpha f)(x)=\alpha f(x)$ 

⇒Let G be an open subset in R<sup>n</sup> then the set C(G,R) of continuous functions f: X->R is a linear space.

## Linear Independence

- Why?
- What?
  - The vectors are *linearly dependent* if there exists a *linear combination* of the vectors- with at least one nonzero coefficients  $\alpha_k$ , that renders the zero vector.
  - Sinusoidal functions in C(0,1) are linearly independent. Why?

## Dimension

- A measure of how many and what sort of vectors describe the space.
- A vector space S has dimension *n*, if:
  - It possesses a set of *n* independent vectors,
  - Every set of (n+1) vectors is dependent.
- If for every positive integer *k*, one can find *k* independent vectors in S, then S has infinite dimension.

## Basis

- A certain set inside the vector space S, forms a basis provided that:
  - The vectors in the set be linearly independent.
  - Every arbitrary vector in S can be written as a linear combination of basis elements.
- Representation of each vector within S is unique in terms of the given basis.

#### **Inner Product Space**

 A linear space S is a complex inner product space, if for every ordered pair (x,y) of vectors; i.e. in SxS, there exists a unique scalar <x,y> such that:

$$- < x + y, z > = < x, z > + < y, z >$$

 $- \langle x, x \rangle \geq 0$ , with equality iff x=0

## **Inner Product Space**

- A real inner product space can be similarly defined.
- Cauchy-Schwarz-Bunjakowsky inequality reads as:

 $|<x,y>| <= (<x,x>)^{1/2} (<y,y>)^{1/2}$ Why?

# Orthogonality

- Two vectors are orthogonal if their inner product is zero.
- A set of vectors is orthogonal if all members are mutually orthogonal.
- *Proper* orthogonal set is composed of linearly independent vectors.
- Orthonormal set.

## **Normed Linear Space**

- A linear space S is a normed linear space if for every vector x in S, there exists a unique real number ||x|| such that:
  - $||x|| \ge 0$ , with equality iff x = 0
  - $||\alpha x || = |\alpha| ||x||$
  - $||x+y|| \le ||x|| + ||y||$  (triangle inequality)

# **Normed Linear Space**

- A norm can be induced by the inner product.
- Why?
- Norm can be used to measure 'closedness', how? Why? What is it good for?

## **Normed Linear Space**

- Convergence
- Cauchy sequence
- Completeness
- $C(\alpha,\beta)$  is incomplete, why?