

# Linear Algebra Review Session Day 1 (part 1)

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# Outline

- Motivation for Linear Algebra

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- Why is one not *easy*?
- Why are these both equally *easy*?



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- This is the essence of math... Math is hard, nearly impossible without the right tools.
- Linear Algebra helps use create tools to answer tricky problems!

# Why Linear Algebra?

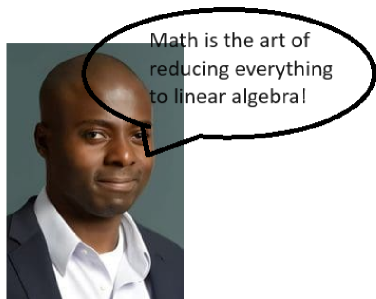


Figure: Dr. Edinah Grang

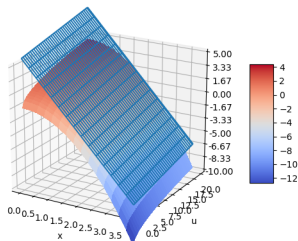


Figure: Linearization

# Fields

In order to do linear algebra, we need numbers. Or at the very least, we need stuff to work with—something that looks familiar (to the real numbers) but is general enough to handle more *complex* settings

As we'll see, the *linearity* of *linear* algebra, needs us to be able to add, scale (multiply by a constant), and distribute the two operations. Thus, we need an ambient space to do these calculations.



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⑥ Distributivity:  $\forall a, b, c \in F, a \cdot (b + c) = a \cdot b + a \cdot c$

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- What if we want to handle multiple values at once? Say we want to keep track of position with an  $x$ ,  $y$ , and  $z$  coordinate.

# Lists

## Definition

**List:** For a fixed natural number  $n \in \mathbb{N}$ , define an  $n$ -dimensional list over field  $\mathbb{F}$  to be

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## Remark

We call  $x_j$  then  $j^{\text{th}}$ -coordinate

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- **Scalar multiplication:**  $(\cdot : \mathbb{F} \times \mathbb{F}^n \rightarrow \mathbb{F}^n)$  is component wise too:

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- **(additive) Inverse:** need for  $\mathbf{x} + -\mathbf{x} = \mathbf{0}$ , so we define  $-\mathbf{x} = (-x_1, \dots, -x_n)$

# Operations in $\mathbb{F}^n$

## Remark

*We immediately get associativity, commutativity, and distributivity (of scalar multiplication) from the properties of the underlying field. We also get uniqueness of inverses, identities, and the properties that  $\mathbf{x} = 1 \cdot \mathbf{x}$  and  $-\mathbf{x} = (-1) \cdot \mathbf{x}$*

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## Proof.

$$\begin{aligned}\mathbf{x} + \mathbf{y} &= (x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n) = \\ &= (y_1 + x_1, \dots, y_n + x_n) = (y_1, \dots, y_n) + (x_1, \dots, x_n) = \mathbf{y} + \mathbf{x}\end{aligned}$$



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- 6  $\begin{cases} a(u + v) = au + av & \forall a \in \mathbb{F}, u, v \in V \\ (a + b)v = av + bv & \forall a, b \in \mathbb{F}, v \in V \end{cases}$

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## Remark (notation)

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## Remark

$\mathbb{F}^n$  and  $\mathbb{F}^\infty$  are of the latter form, for  $S = \{1, 2, \dots, n\}$  with  $n \in \mathbb{N} \cup \{\infty\}$  as  $(x_1, \dots, x_n)$  can be seen as the output of the function  $f : \{1, \dots, n\} \rightarrow \mathbb{F}$  with  $x_j = f(j)$

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- We can do the same for  $\mathbb{F} = \mathbb{C}$ , or really any field as well.
- In summary, there are many types of vector spaces, but the easiest to understand and utilize are *finite dimensional* vector spaces (will see soon).

# Some Properties of Vector Spaces

$0v = v \quad \forall v \in V$	<i>Proof</i> : $0v = (0 + 0)v = 0v + 0v \implies 0 = 0v$
$a0 = 0 \quad \forall a \in \mathbb{F}$	<i>Proof</i> : $a0 = a(0 + 0) = a0 + a0 \implies a0 = 0$
$(-1)v = -v \quad \forall v \in V$	<i>Proof</i> : $(-1)v = 0 + (-1)v = (-v + v) + (-1)v$ $= (-v) + ((1 + -1)v) = (-v)$
Unique Identity	<i>Proof</i> : $0' = 0' + 0 = 0$
Unique Inverses	<i>Proof</i> : $w = w + 0 = w + (v + w')$ $= (w + v) + w' = 0 + w' = w'$
$\sum_{i=1}^n v_i \in V, v_i \in V$	<i>Proof</i> : By induction... $\underbrace{(v_1 + \dots + v_{n-1})}_{\in V} + v_n \in V$

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## Remark

*We actually just have to check the first because the other two are direct corollaries! Suppose 1) holds, then:*

$$0 = 0 * v \in V \quad -v = (-1)v \in V$$



## Subspace (examples)

- $U_1 = \{(x_1, x_2, 0) : x_1, x_2 \in \mathbb{F}\}$  is a subspace of  $\mathbb{F}^3$

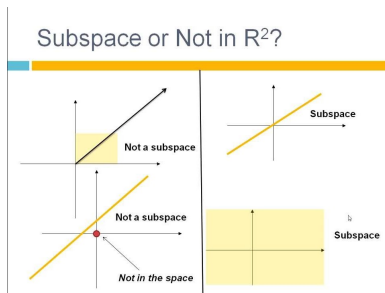


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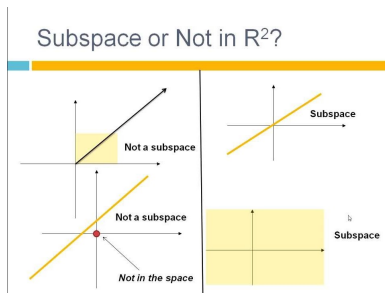


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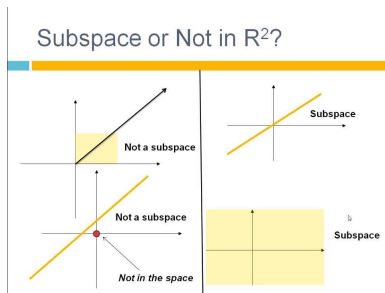


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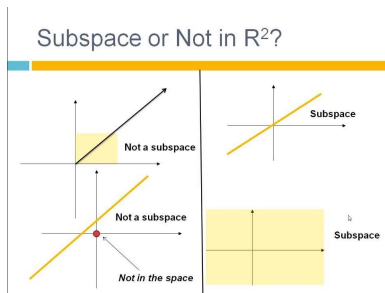


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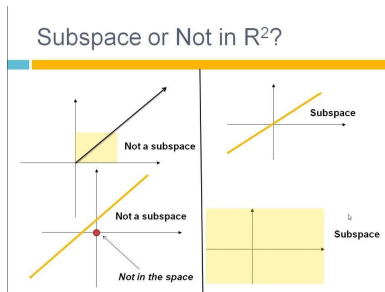


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# Direct Sum

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We say a sum of subspace  $U_1 + \dots + U_m$  is a **direct sum** iff each element  $v = u_1 + \dots + u_m$ ,  $u_j \in U_j$  has *unique* representation

We write this as  $U_1 \oplus \dots \oplus U_m$

That is to say, if

$$v = u_1 + \dots + u_m$$

$$v = \hat{u}_1 + \dots + \hat{u}_m$$

Then  $u_j = \hat{u}_j$  for all  $j = 1, \dots, m$

Equivalently, by subtracting both equations, we get as necessary and sufficient condition for a sum of subspaces being a **direct sum**:

$$0 = u_1 + \dots + u_m \iff u_1 = u_2 = \dots = u_m = 0$$

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- Direct sums are the "disjoint subset" analog, and in the case of adding two subspaces, we have direct sum iff  $U_1 \cap U_2 = \{0\}$