Math Camp

Module #1 Vector spaces and linear maps Part I: enacting peace

"The introduction of numbers as coordinates is an act of violence." H. Wevl

Why vector spaces?

Economics is a multivariate affair...

- trade offs only occur when there are two or more outcomes
- linear (or vector) spaces are the simplest environment capable of handling multi-dimensional problems
- start abstract: no reference yet to bases, dimension or coordinates

Vector spaces and direct sums

Definition A *vector space* is a set V that is closed under addition, and under scalar multiplication by elements of \mathbb{R}

they are related by the distributed property:

$$\alpha \in \mathbb{R}, v, w \in V \Rightarrow \alpha(v+w) = \alpha v + \alpha w$$

V contains a special element:

$$v \in V \implies v + 0 = v$$

More generally, could be any field.

Vector spaces and direct sums

• If V and W are vector spaces then $V \oplus W$ is the Cartesian product of V and W such that

$$∘ x ∈ V ⊕ W implies x = (v_x, w_x), with v_x ∈ V, w_x ∈ W$$

$$∘ x + y = (v_x + w_x, v_y + w_y)$$

• Q: what is $\mathbb{R} \oplus \mathbb{R}$?

Spans and subsets

Definition. A subspace W of V is a subset of V that is, itself, a vector space.

Definition. If $A \subset V$ then the *span* of A is the set of all finite linear combinations of elements of A:

$$\operatorname{\mathsf{span}}(A) = \left\{ \sum_{k=1}^m \alpha_k a_k \text{ such that } m \in \mathbb{N}, \alpha_k \in \mathbb{R}, a_k \in A \right\}$$

span(A) is the smallest subspace of V containing A.

Linear Independence and bases

Definition. $A \subset V$ is *linearly independent* if no element of A can be written as a finite linear combination of other elements of A.

More formally, A is linearly independent if whenever

$$\{a_1,...,a_n\}\subset A \text{ and } \sum_{k=1}^n lpha_k a_k=0$$

it follows that $\alpha_k = 0$ for $k = 1, \dots n$.

Definition. A *basis* of V is a linearly independent set $B \subset V$ that spans V.

Linear Independence and bases

Theorem (Fundamental theorem of linear algebra)

if $A = \{a_1,...,a_n\} \subset V$ is a basis for V and $B = \{b_1,...,b_m\} \subset V$ is linearly independent then $m \leq n$

Dimension

If $B \subset V$ is a basis then $\dim V = |B|$.

Let $\dim V = n$. Then

- If B ⊂ V is linearly independent then B is a basis for span(B) and dim span(B) = |B|.
- If $B \subset V$ is linearly independent and |B| = m < n then there exists $C \subset V$ so that $B \cup C$ is a basis for V.
- If W is a subspace of V then $\dim(W) \leq \dim(V)$.

Linear functions

Category Theory: everything is objects and arrows

In the category of linear spaces,

- objects are vector spaces
- arrows are linear maps

Definition. If V and W are vector spaces then a map $f:V\to W$ is *linear* provided

$$f(\alpha v_1 + \beta v_2) = \alpha f(v_1) + \beta f(v_2)$$

Linear functions

Definition. A linear map is an *isomorphism* if it is invertible.

If there is an isomorphism between the vector spaces V and W then we say that V and W are *isomorphic*, written $V \cong W$.

Linear functionals

Definition. A *linear functional* α on V is a linear map $\alpha: V \to \mathbb{R}$.

- The set of linear functionals, V^* , called the *dual space*
- V* is a vector space
- If $\dim V < \infty$ the $V^* \cong V$.

Linear functions

Let $f: V \to W$ be a linear map.

Definition. The *kernel (or nullspace)* of f is the collection of vectors that f sends to zero:

$$ker(f) = \{v \in V : f(v) = 0\} \subset W$$

- ker(f) is a subspace of V
- The *nullity* of *f* is the dimension of its kernel

Linear functions

Definition. The *range* of f is the image of V in W under f:

$$f(V) = \{ w \in W : \exists v \in V \text{ with } f(v) = w \} \subset W$$

- The range of *f* is a subspace of *W*
- The rank of f is the dimension of the range

Linear Functions

Rank Nullity Theorem: if V is finite dimensional and $f: V \to W$ is a linear map then the rank of f plus the nullity of f equals the dimension of V

$$\dim ker(f) + \dim f(V) = \dim V$$

Information and nullity

Let $f: V \to W$ b a linear map. Define the following relation on V:

$$v_1 \sim v_2 \iff f(v_1) = f(v_2) \text{ i.e. } v_1 - v_2 \in ker(f)$$

- ullet \sim is an equivalence relation
- ullet \sim measures the information lost by f

Linear extensions

Let V and W be vector spaces and $B \subset V$ a basis. Let $\phi : B \to W$ be *any* function. Then there exists unique linear map $\Phi : V \to W$ such that the following diagram commutes:



Let $B = \{b_1, \dots, b_n\}$ be a basis for V and $\{w_1, \dots, w_n\} \subset W$. Define a linear map Φ from V to W by sending b_i to $\phi(b_i) = w_i$, and *extending linearly:*

$$v = \sum_{k=1}^{n} \beta_k b_k \to \sum_{k=1}^{n} \beta_k \phi(b_k) = \sum_{k=1}^{n} \beta_k w_k \equiv \Phi(v).$$

Linear extensions

Let V and W be vector spaces and $B \subset V$ a basis. Let $\phi : B \to W$ be *any* function. Then there exists unique linear map $\Phi : V \to W$ such that the following diagram commutes:



Note that the behavior of linear map is entirely characterized by its behavior on a basis

Characterization of finite dimensions vector spaces

Theorem. $\dim(V) = n \implies V \cong \mathbb{R}^n$

Dynamics and Decompositions

Let dim V = n and $f : V \to V$ be linear. Given $v_0 \in V$, define $v_{t+1} = f(v_t)$.

- The *dynamic f* traces a path/orbit in the vector space *V*.
- The orbits of *f* partition *V*.
- The subspace W ⊂ V is invariant (under the action of f) provide f(W) ⊂ W.

Dynamics and decompositions

Theorem (Schur Decomposition) Let $\dim V = n$ and $f: V \to V$ be linear. Then there is a collection of invariant subspaces $\{V_k\}_{k=1}^n$ such that

$$\dim(V_k) = k$$
 and $V_k \subset V_{k+1}$

Dynamics and decompositions

Definition. The scalar $\lambda \in \mathbb{R}$ of a linear map f is an eigenvalue provided there exists $v \in V$ such that $f(v) = \lambda v$

- v is called an associated eigenvector.
- if v is an eigenvector associated to λ then f scales v by λ .
- the collection of all eigenvectors associated with an eigenvalue is called the eigenspace.

Eigenspace decomposition (greatest thing ever!)

The set up:

- $\dim V = m$ and $f: V \to V$ linear
- $V(\lambda)$ is the eigenspace associated with λ
- Assume the eigenvalues are distinct.

Then there is an isomorphism $\phi: V \to \bigoplus_{i=1}^m V(\lambda_i)$ such that

