

# Analysis, Topology, Optimization, Machine Learning, and Computational Analysis

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June 18, 2020

## 1 Notation, Set Theory, and Logic

**Definition 1.1. Common Sets of Numbers**

$$\mathbb{N} = \{1, 2, \dots\}$$

$$\mathbb{N}_0 = \{0, 1, 2, \dots\}$$

$$\mathbb{N}_m = \{1, 2, \dots, m\} \text{ where } m \in \mathbb{N}$$

$\mathbb{R}$  is the set of Real Numbers

$$\mathbb{R}^+ = \{x \in \mathbb{R} : x > 0\}$$

$$\mathbb{R}_0^+ = \{x \in \mathbb{R} : x \geq 0\}$$

$$\mathbb{R}^- = \{x \in \mathbb{R} : x < 0\}$$

$$\mathbb{R}_0^- = \{x \in \mathbb{R} : x \leq 0\}$$

**Definition 1.2. Sets and Set Builder Notation**

*Set builder notation is a way of describing a set using mathematical, logical symbols, or words. Look at the following example:*

$$E = \{x \in \mathbb{N} : x = 2n \text{ where } n \in \mathbb{N}\}$$

*This reads:  $E$  is the set of all  $x$  in  $\mathbb{N}$  such that  $x = 2n$  where  $n$  is in  $\mathbb{N}$*

*This set is also known as the even numbers.*

*When talking about functions, another common way of describing a set is:*

$$C_X^Y = \{f : X \rightarrow Y | f \text{ is continuous}\}$$

*This reads:  $C_X^Y$  is the set of all functions  $f$  mapping from  $X$  to  $Y$  such that  $f$  is continuous.*

**Definition 1.3. Primitives**

*The logical or and not are both primitives and are written:*

*logical or :  $\vee$*

*and*

*not :  $\neg$*

Respectively.

Statements are written:  $L, M, N, O, P, Q, \dots$  etc

A statement is a sequence of words or symbols which is either true or false.

So then  $L \vee M$  is a new statement composed of  $L, \vee$ , and  $M$ .

We can then use this as the definition of a new statement:

$$N := L \vee M$$

Which is read:  $N$  is defined as  $L$  or  $M$ .

So  $N$  is true if:

$L$  is true,  $M$  is true, or  $N$  and  $M$  are both true.

#### **Definition 1.4. And**

Let  $A, B$  be sets.

$$A \wedge B := \neg((\neg A) \vee (\neg B))$$

#### **Definition 1.5. Intersection and Union**

Let  $A, B$  be sets.

$$A \cap B = \{x : x \in A \wedge x \in B\}$$

$$A \cup B = \{x : x \in A \vee x \in B\}$$

#### **Definition 1.6. Subsets**

Let  $A, B$  be sets.

$$A \subset B \Leftrightarrow (x \in A \Rightarrow x \in B)$$

#### **Theorem 1.1. The union only makes things larger**

Let  $A, B$  be sets.

Then:  $A \subset A \cup B$

#### **Theorem 1.2. Union and Intersection Distributive Properties**

$$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$$

*Proof:*

$$\text{Let } x \in A \cap (B \cup C) \Rightarrow x \in A \wedge x \in B \cup C$$

Suppose that  $x \notin C \Rightarrow x \in B$

$$\Rightarrow x \in A \wedge x \in B$$

$$\Rightarrow x \in A \cap B$$

$$\Rightarrow x \in (A \cap B) \cup (A \cap C)$$

$$[(A \cap B) \subset (A \cap B) \cup (A \cap C)]$$

Suppose that  $x \notin B \Rightarrow x \in C$

$$\Rightarrow x \in A \text{ and } x \in C$$

$$\Rightarrow x \in A \cap C$$

$$\Rightarrow x \in (A \cap B) \cup (A \cap C)$$

$$[(A \cap C) \subset (A \cap B) \cup (A \cap C)]$$

Therefore:

$$A \cap (B \cup C) \subset (A \cap B) \cup (A \cap C)$$

**Definition 1.7. Power Set**

Let  $X \neq \phi$

$$2^X := \{V : V \subseteq X\}$$

## 2 Topology

**Definition 2.1. Topology**

Let  $X \neq \phi$

Further let  $\tau \subseteq 2^X$  such that:

$$\begin{aligned}
& \phi, X \in \tau \\
& (\forall A \neq \phi) \left( \{U_\alpha\}_{\alpha \in A} \subseteq \tau \Rightarrow \bigcup_{\alpha \in A} U_\alpha \in \tau \right) \\
& (\forall m \in \mathbb{N}) \left( \{U_j\}_{j \in \mathbb{N}_m} \Rightarrow \bigcap_{j=1}^m U_j \in \tau \right)
\end{aligned}$$

**Definition 2.2. Relative Topology**

Let  $X \neq \phi$  and  $Z \subset X$

Then the relative topology on  $Z$  is written as follows:

$$\tau_Z = \{Z \cap U : U \in \tau_X\}$$

**Theorem 2.1. The Relative Topology is a Topology on  $Z$** 

Let  $E \in \tau_Z$

$$\Rightarrow E = Z \cap U \subset Z$$

$$\Rightarrow \tau_Z \subseteq 2^Z$$

And so we have met the first criteria.

Next:

$$\phi \in \tau \Rightarrow Z \cap \phi \in \tau_Z \Rightarrow Z \cap \phi = \phi \in \tau_Z$$

Next:

$$X \in \tau \Rightarrow Z \cap X \in \tau_Z \Rightarrow Z \cap X = Z \in \tau_Z$$

Next: Let  $A \neq \phi$  and  $\{U_\alpha\}_{\alpha \in A} \in \tau_Z$

$$\Rightarrow \exists \{V_\alpha\}_{\alpha \in A} \subset \tau \text{ such that: } U_\alpha = Z \cap V_\alpha$$

$$\Rightarrow \bigcup_{\alpha \in A} U_\alpha = \bigcup_{\alpha \in A} Z \cap V_\alpha$$

**Theorem 2.2. Topologies on Finite Spaces**

Let  $E \neq \phi$  and  $\text{card}(E) = \phi$  then  $\tau_E = 2^E$

### 3 Change

#### Definition 3.1. *Metric*

Let  $X$  be a non-empty set.

Let  $d : X \times X \rightarrow \mathbb{R}_0^+$  such that:

- $(\forall x, y \in X) d(x, y) = 0 \Leftrightarrow x = y$
- $(\forall x, y \in X) d(x, y) = d(y, x)$
- $(\forall x, y, z \in X) d(x, z) \leq d(x, y) + d(y, z)$

Then  $d$  is called a metric and  $(X, d)$  is called a metric space.

Reference

#### Definition 3.2. *Limit of a function*

Let  $T : X \rightarrow Y$  where  $(X, d_X)$  and  $(Y, d_Y)$  are metric spaces.

Then fix  $x_0 \in X$ .

If:

$$(\exists L \in Y)(\forall \epsilon > 0)(\exists \delta > 0)(\forall x \in X)(d(x, x_0) < \delta \Rightarrow d(f(x), L) < \epsilon)$$

Then:

$$\lim_{x \rightarrow x_0} f(x) = L$$

Reference

#### Definition 3.3. *Derivative*

Let  $\hat{f} : \mathbb{R} \rightarrow \mathbb{R}$

Further let  $f = \hat{f}|_U$  where  $U \in \tau_{\mathbb{R}}$

Then  $f$  is said to be differentiable at  $x \in U$  if there exists an  $L_x$  such that:

$$L_x = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

If  $L_x$  exists for all  $x \in U$  then we write:

$$\frac{d}{dx} f(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

Reference

#### Theorem 3.1. *Fundamental increment lemma*

Let  $f$  be described as above and be differentiable at  $x$ .

Then there exists a function  $\phi : \mathbb{R} \rightarrow \mathbb{R}$  such that:

$$f(x+h) = f(x) + \frac{d}{dx} f(x)h + \phi(x)h$$

and

$$\lim_{h \rightarrow 0} \phi(h) = 0$$

**Proof:**

Define:  $\phi(h) = \frac{f(x+h) - f(x)}{h} - \frac{d}{dx} f(x)$

Then:  $\phi(h)h = f(x+h) - f(x) - \frac{d}{dx} f(x)h$

Then:  $\phi(h)h + f(x) - \frac{d}{dx} f(x)h = f(x+h)$

And so we have property 1.

Next:

$$\begin{aligned}\lim_{h \rightarrow 0} \phi(h) &= \lim_{h \rightarrow 0} \left[ \frac{f(x+h) - f(x) - \frac{d}{dx}f(x)h}{h} \right] = \lim_{h \rightarrow 0} \left[ \frac{f(x+h) - f(x)}{h} - \frac{d}{dx}f(x) \right] \\ &= \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} - \lim_{h \rightarrow 0} \frac{d}{dx}f(x) = \frac{d}{dx}f(x) - \frac{d}{dx}f(x) = 0\end{aligned}$$

**Definition 3.4. Partial Derivative**

Let  $\hat{f} : \mathbb{R}^n \rightarrow \mathbb{R}$

Further let  $f = \hat{f}|_U$  where  $U \in \tau_{\mathbb{R}^n}$

Then  $f$  is said to be differentiable at  $x \in U$  with respect to the  $i$ 'th component of  $x$  if there exists an  $L_{x_i}$  such that:

$$L_{x_i} = \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}$$

If  $L_{x_i}$  exists for all  $x \in U$  then we write:

$$\frac{\partial}{\partial x_i} f(x) = \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}$$

Reference

**Theorem 3.2. Equivalent characterization**

Let  $\hat{f} : \mathbb{R}^n \rightarrow \mathbb{R}$

Further let  $f = \hat{f}|_U$  where  $U \in \tau_{\mathbb{R}^n}$

And let  $f$  be differentiable at  $x \in U$  with respect to the  $i$ 'th component of  $x$ , then:

$$\begin{aligned}L_{x_i} &= \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h} \\ \Leftrightarrow 0 &= \lim_{h \rightarrow 0} \left[ \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h} - L_{x_i} \right] \\ \Leftrightarrow 0 &= \lim_{h \rightarrow 0} \left[ \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h} - \frac{L_{x_i} \cdot h}{h} \right] \\ \Leftrightarrow 0 &= \lim_{h \rightarrow 0} \left[ \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n) - \langle L_{x_i}, h \rangle}{h} \right]\end{aligned}$$

**Definition 3.5. Gradient**

Let  $\hat{f} : \mathbb{R}^n \rightarrow \mathbb{R}$  and let  $f : U \rightarrow \mathbb{R}$  such that  $f = \hat{f}|_U$  where  $U \in \tau_{\mathbb{R}^n}$

$f$  is said to be differentiable at  $x \in U$  if  $\exists \nabla f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  such that:

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - \langle \nabla f(x), h \rangle|}{\|h\|} = 0$$

**Theorem 3.3. Form of the Gradient**

Let  $f$  be defined as above.

Then  $\nabla f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  where:

$$\nabla f(x) = \begin{bmatrix} \frac{\partial}{\partial x_1} f(x) \\ \vdots \\ \frac{\partial}{\partial x_n} f(x) \end{bmatrix} \quad \forall x \in \mathbb{R}^n$$

is the form of  $\nabla f$  which satisfies the above statement if  $f$  is differentiable.

*Reference*

**Proof:**

Suppose  $\nabla f$  is defined as above and all the partial derivatives exist.

Then:

$$\frac{1}{||h||} |f(x+h) - f(x) - \langle \nabla f(x), h \rangle| = \frac{1}{||h||} \left| f(x+h) - f(x) - \sum_{j=1}^n \frac{\partial}{\partial x_j} f(x) \cdot h_j \right|$$

**Definition 3.6. Matrix Functional Differentiability**

Let  $\hat{T} : \mathbb{R}^{n \times m} \rightarrow \mathbb{R}$  and let  $T : U \rightarrow \mathbb{R}$  such that  $T = \hat{T}|_U$  where  $U \in \tau_{\mathbb{R}^{n \times m}}$ .  
 $T$  is said to be differentiable at  $x \in U$  if  $\exists D : \mathbb{R}^{n \times m} \rightarrow \mathbb{R}^{n \times m}$  such that:

$$\lim_{h \rightarrow 0} \frac{|T(x+h) - T(x) - \langle DT(x), h \rangle|}{\|h\|} = 0$$

where  $\langle \cdot, \cdot \rangle$  is an inner product defined on  $\mathbb{R}^{n \times m}$

**Definition 3.7. Frobenius inner product**

The Frobenius inner product is defined as:

$$\langle \cdot, \cdot \rangle_{FB} : \mathbb{R}^{n \times m} \times \mathbb{R}^{n \times m} \rightarrow \mathbb{R} \text{ such that: } \langle A, B \rangle_{FB} = \sum_{i=1}^n \sum_{j=1}^m a_{i,j} b_{i,j} \text{ for all } A, B \in \mathbb{R}^{n \times m}$$

**Theorem 3.4. Form of Matrix Functional Derivative**

$$DT(x) = \begin{bmatrix} \frac{\partial}{\partial x_{1,1}} & \cdots & \frac{\partial}{\partial x_{1,m}} \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_{n,1}} & \cdots & \frac{\partial}{\partial x_{n,m}} \end{bmatrix}$$

**Definition 3.8. Differentiability of a multi-variable function.**

Let  $\hat{f} : \mathbb{R}^m \rightarrow \mathbb{R}^n$  such that:

$$\hat{f}(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{bmatrix} \quad \text{and } (\forall j \in \mathbb{N}_n)(f_j : \mathbb{R}^m \rightarrow \mathbb{R})$$

Further let  $f = \hat{f}|_U$  where  $U \in \tau_{\mathbb{R}^m}$

Then  $f$  is said to be differentiable at  $x \in U$  if there exists a linear operator  $J_{f(x)} : \mathbb{R}^m \rightarrow \mathbb{R}^n$  such that:

$$\lim_{h \rightarrow \vec{0}} \frac{\|f(x+h) - f(x) + J_{f(x)}(h)\|_{\mathbb{R}^n}}{\|h\|_{\mathbb{R}^m}} = 0$$

Reference

**Theorem 3.5. If a multi-variable function,  $f$ , is differentiable at  $x$  then the linear operator  $J$  is the Jacobian matrix.**

So our guess is that:

$$J_{f(x)} = \begin{bmatrix} \frac{\partial}{\partial x_1} f_1(x) & \cdots & \frac{\partial}{\partial x_n} f_1(x) \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} f_m(x) & \cdots & \frac{\partial}{\partial x_n} f_m(x) \end{bmatrix}$$

since this form is a linear operator mapping from the appropriate space to the appropriate space. It should be noted that the transpose of this matrix can not satisfy the definition of differentiability of a multi-variable function and so it is not the correct linear operator.

**Definition 3.9. Matrix operator differentiability**

Let  $T : \mathbb{R}^{n \times m} \rightarrow \mathbb{R}^n$  such that:

$$T(A) = \begin{bmatrix} T_1(A) \\ \vdots \\ T_n(A) \end{bmatrix} \quad \forall A \in \mathbb{R}^{n \times m} \quad \text{and } (\forall j \in \mathbb{N}_n)(T_j : \mathbb{R}^{n \times m} \rightarrow \mathbb{R})$$

Then  $T$  is said to be differentiable at  $A \in \mathbb{R}^{n \times m}$  if there exists a linear operator  $D : \mathbb{R}^{n \times m} \rightarrow \mathbb{R}^n$  where:

$$\lim_{h \rightarrow 0} \frac{\|T(A+h) - T(A) + D(h)\|_{\mathbb{R}^n}}{\|h\|_{\mathbb{R}^{n \times m}}} = 0$$

If  $D$  exists then it is called the Matrix operator derivative and is written:  $D_{\mathbb{R}^{n \times m}} T(A)$

**Theorem 3.6. The form of the Matrix operator derivative.**

Let  $T$  be described as above and differentiable at  $A \in \mathbb{R}^{n \times m}$

$$\frac{T(A+h) - T(A)}{\|h\|} = \begin{bmatrix} \frac{T_1(A+h) - T_1(A)}{\|h\|} \\ \vdots \\ \frac{T_n(A+h) - T_n(A)}{\|h\|} \end{bmatrix}$$

and so:

$$\lim_{h \rightarrow 0} \frac{T(A+h) - T(A)}{\|h\|} = \begin{bmatrix} \lim_{h \rightarrow 0} \frac{T_1(A+h) - T_1(A)}{\|h\|} \\ \vdots \\ \lim_{h \rightarrow 0} \frac{T_n(A+h) - T_n(A)}{\|h\|} \end{bmatrix}$$



**Definition 3.10. Subspace Differentiability**

Let  $X = \{X_j\}_{j=1}^n$  be a sequence of finite dimensional vector spaces where  $\dim(X_j) = k_j = m_j \times n_j$

Let  $T : \prod_{j=1}^n X_j \rightarrow Y$  where  $Y$  is a finite dimensional vector space with  $\dim(Y) = k_y$

Let  $x_j \in X_j$  for some  $j \in \mathbb{N}_n$

Where

$$x_j = \begin{bmatrix} x_{1,1} & \cdots & x_{1,n_j} \\ \vdots & \ddots & \vdots \\ x_{m_j,1} & \cdots & x_{m_j,n_j} \end{bmatrix}$$

$T$  is said to be differentiable at  $x \in X$  where  $x = (x_0, \dots, x_j, \dots, x_{n-1})$  with respect to  $X_j$  if there exists a linear operator  $D : X_j \rightarrow Y$ :

Given  $h \in X_j \setminus \{\vec{0}\}$  define  $\hat{h} = (0, \dots, h, \dots, 0) \in X$  where  $h$  is in the  $j$ 'th place of  $\hat{h}$ :

$$\lim_{h \rightarrow 0} \frac{\|T(x + \hat{h}) - T(x) + D(h)\|_Y}{\|h\|_{X_j}} = 0$$

Then  $D$  is called the subspace derivative of  $T$  at  $x$  with respect to  $X_j$  and is written:  $D_{x_j}T(x)$

**Definition 3.11. *Product space Derivative***

Let  $X = \{X_j\}_{j=0}^{n-1}$  be a sequence of finite dimensional vector spaces where  $\dim(X_j) = k_j$

Let  $T : \prod_{j=0}^{n-1} X_j \rightarrow Y$  where  $Y$  is a finite dimensional vector space with  $\dim(Y) = k_y$

Let  $\{x_j\}_{j=0}^{n-1}$  be a sequence of vectors such that:  $(\forall j \in \{0, \dots, n-1\})(x_j \in X_j)$

The product space derivative at the point  $z \in X$  is:

$$D_X T(z) = \begin{bmatrix} D_{x_0} T(z) \\ \vdots \\ D_{x_{n-1}} T(z) \end{bmatrix}$$

**Definition 3.12. Fréchet derivative**

Let  $V, W$  be normed vector spaces and  $U \subset V$  be an open set.

An operator  $f : U \rightarrow W$  is said to be Fréchet differentiable if there exists a bounded linear operator  $A : V \rightarrow W$  such that:

$$\lim_{\|h\| \rightarrow 0} \frac{\|f(x+h) - f(x) + Ah\|_W}{\|h\|_V} = 0$$

Reference

**Theorem 3.7. Fréchet derivative of a bounded linear operator**

Let  $V, W$  be normed vector spaces and  $U \subset V$  be an open set.

Let  $\hat{f} : V \rightarrow W$  be a bounded linear operator.

Then let's look at  $f = \hat{f}|_U$

My guess is that  $A = \hat{f}$

Let  $x \in U$  and  $h \in U$  with  $\|h\| \neq 0$  and  $x+h \in U$ , Then:

$$\frac{\|f(x+h) - f(x) + Ah\|_W}{\|h\|_V} = \frac{\|f(x) + f(h) - f(x) + \hat{f}(h)\|_W}{\|h\|_V} = \frac{\|f(x) + f(h) - f(x) + f(h)\|_W}{\|h\|_V} = 0$$

Thus let  $\epsilon > 0$  and  $\delta > 0$

Then if  $0 < \|h\| < \delta$  we know that  $\frac{\|f(x+h) - f(x) + Ah\|_W}{\|h\|_V} = 0 < \epsilon$

Therefore:

$$\lim_{\|h\| \rightarrow 0} \frac{\|f(x+h) - f(x) + Ah\|_W}{\|h\|_V} = 0$$

Thus  $A = \hat{f}$  is the Fréchet derivative of  $f$ .

**3.1 Finite Composition Operator****Definition 3.13. Finite Composition Operator**

Let the collection  $X = \{X_j\}_{j=0}^n$  be a finite sequence of sets.

Further let  $\{T_j\}_{j=0}^{n-1}$  be a finite sequence of operators such that  $(\forall j \in \mathbb{N}_{n-1})(T_j : X_j \rightarrow X_{j+1})$

Then  $T^n : X_0 \rightarrow X_n$  defined by:

$$T^n := \bigcirc_{j=0}^{n-1} T_j$$

is called the **Finite Composition Operator defined on  $X$** .

**Definition 3.14. Multi-variable Finite Composition Iteration**

Let the collection  $X = \{X_j\}_{j=0}^n$  and  $Y = \{Y_j\}_{j=0}^{n-1}$  be finite sequences of sets.

Further let  $\{T_j\}_{j=0}^{n-1}$  be a finite sequence of operators such that:  $(\forall j \in \mathbb{N}_{n-1})(T_j : X_j \times Y_j \rightarrow X_{j+1})$

Let  $T^n : X_0 \times \prod_{j=0}^{n-1} Y_j \rightarrow X_n$  where:

$$T^n(x, y) = z_n \text{ where } z_{j+1} = T_j(z_j, \pi_j(y)) \text{ or } z_{j+1} = T_j(z_j) \text{ and } z_0 = x \in X_0$$

**Definition 3.15. Gradient Descent**

Let  $E : \mathbb{R}^n \rightarrow \mathbb{R}$  be a differentiable operator.

The method of Gradient Descent says that a local minimum of  $E$  can be found using the following iteration:

$$a_{n+1} = a_n - \gamma \nabla E(a_n)$$

Where  $\gamma > 0$

**Example 3.1. Objective Operator for Data Set Defined Operator Approximation**  
Let  $X \subset \mathbb{R}^n, Y \subset \mathbb{R}^m$  such that  $X \times Y$  defines an operator  $T$ .

$$E(a) = \sum_{x \in X} ||T(x) - T^n(x, a)||$$

## 4 Surjective Continuous Non-decreasing Bounded Functionals

Let  $B = \{f : \mathbb{R} \rightarrow [0, 1] \mid f \text{ is surjective, continuous, and non-decreasing.}\}$

**Theorem 4.1.  $B$  is convex.**

Let  $f, g \in B$  and  $h(x) := \lambda f(x) + (1 - \lambda)g(x)$  where  $\lambda \in [0, 1]$

Then  $h$  is still continuous since the linear combination of continuous functions is continuous.

Since both  $f$  and  $g$  are surjective and non-decreasing, then there exists  $x_0, y_0, x_1, y_1$  in  $\mathbb{R}$  such that:

$f(x_0) = 0 = g(y_0)$  and  $f(x_1) = 1 = g(y_1)$

Suppose WLOG that  $x_0 \leq y_0$  and  $x_1 \leq y_1$

Then we know that:

$$h(x_0) = \lambda f(x_0) + (1 - \lambda)g(x_0) = \lambda 0 + (1 - \lambda)0 = 0$$

and

$$h(y_1) = \lambda f(y_1) + (1 - \lambda)g(y_1) = \lambda 1 + (1 - \lambda)1 = 1$$

Now if we pick  $\alpha \in [0, 1]$  by the intermediate value theorem, we know that there exists an  $x_\alpha \in [x_0, y_1]$  such that:

$$h(x_\alpha) = \alpha$$

Since  $\alpha$  was arbitrary element, I have shown that  $h$  is surjective.

Finally, let  $x_0 < x_1$  be elements in  $\mathbb{R}$

Then we know that  $f(x_0) \leq f(x_1)$  and  $g(x_0) \leq g(x_1)$

$\Rightarrow \lambda f(x_0) \leq \lambda f(x_1)$  and  $(1 - \lambda)g(x_0) \leq (1 - \lambda)g(x_1)$

$\Rightarrow \lambda f(x_0) + (1 - \lambda)g(x_0) \leq \lambda f(x_1) + (1 - \lambda)g(x_1)$

$\Rightarrow h(x_0) \leq h(x_1)$

Thus  $h$  is non-decreasing.

Since  $h$  is surjective, continuous, and non-decreasing, then  $h \in B$

Thus  $B$  is convex.

**Theorem 4.2.  $B$  is translation invariant.**

Let  $f \in B$  and  $g(x) := f(x + c)$  where  $c \in \mathbb{R}$

$f$  is continuous and so is the addition operator so  $g$  is continuous.

Let  $\alpha \in [0, 1]$  since  $f$  is surjective then  $\exists x \in \mathbb{R} \cap f(x) = \alpha$

Then  $g(x - c) = f(x + c - c) = f(x) = \alpha$  and so  $g$  is surjective.

Let  $x < y$  be elements in  $\mathbb{R}$

Then  $f(x) \leq f(y) \Rightarrow f(x + c) \leq f(y + c)$

$\Rightarrow g(x) \leq g(y)$  and so  $g$  is non-decreasing.

Thus  $g \in B$  and  $B$  is therefore translation invariant.

**Theorem 4.3.  $B$  is not complete.**

**Theorem 4.4. Every element in  $B$  can be decomposed as a finite non-trivial convex combination from  $B$**

## 5 A topological description of finite normed spaces.

Finite spaces are of interest because they describe the world of computers. This being the case, we would still like to do analysis on these spaces and analysis starts with topological descriptions.

**Theorem 5.1.** *The first rule of induced topologies on finite subsets is:*

*Let  $V \subset X$  where  $0 < \text{card}(V) =: N < \text{card}(\mathbb{N})$  and  $X$  is a Normed space.*

*The associated topological space on  $V$  is:*

$$\tau_V = \{V \cap U : U \in \tau_X\}$$

*Since  $V$  is of finite cardinality, we can uniquely number each element.*

*Thus:*

$$V = \bigcup_{i=1}^N \{v_i\}$$

*Further:*

$$V_{\min} := \min\{\|x - y\| : x, y \in V\}$$

*Let  $v \in V$  and  $\epsilon_V = \frac{V_{\min}}{2}$*

*Then  $B(v, \epsilon_V) \in \tau_X \Rightarrow V \cap B(v, \epsilon_V) \in \tau_V$*

*However  $V \cap B(v, \epsilon_V) = \{v\}$  and since  $v$  was arbitrary, we thus know that:  $(\forall v \in V)(\{v\} \in \tau_V)$*

*We can now prove that  $\tau_V = 2^V$*

*By definition we know that  $\tau_V \subset 2^V$*

*Let  $E \in 2^V$  then  $E = \bigcup_{j=1}^M \{v_j\}$  where  $M \leq N$*

*Since we know ever  $\{v_j\}$  is open we know that  $E$  is open and thus:  $E \in \tau_V$*

***Thus the induced topology on a finite subset of a Normed space is the discrete topology.***

So if we want to have some form of meaningful topological description of continuity for these spaces, we are going to need a "stronger" form of continuity.