Econ 204 2020

Lecture 8

Outline

1. Bases
2. Linear Transformations
3. Isomorphisms

Linear Combinations and Spans

Definition 1. *Let X be a vector space over a field F. A* linear combination *of $x_1, \ldots, x_n \in X$ is a vector of the form*

$$y = \sum_{i=1}^{n} \alpha_i x_i \text{ where } \alpha_1, \ldots, \alpha_n \in F$$

α_i is the coefficient *of x_i in the linear combination.*

If $V \subseteq X$, the span *of V, denoted* $\operatorname{span} V$*, is the set of all linear combinations of elements of V.*

A set $V \subseteq X$ *spans* X if $\operatorname{span} V = X$.

Linear Dependence and Independence

Definition 2. *A set* $V \subseteq X$ *is* linearly dependent *if there exist* $v_1, \ldots, v_n \in V$ *and* $\alpha_1, \ldots, \alpha_n \in F$ *not all zero such that*

$$\sum_{i=1}^{n} \alpha_i v_i = 0$$

A set $V \subseteq X$ *is* linearly independent *if it is not linearly dependent.*

Thus $V \subseteq X$ is linearly independent if and only if

$$\sum_{i=1}^{n} \alpha_i v_i = 0, \quad v_i \in V \ \forall i \Rightarrow \alpha_i = 0 \ \forall i$$

Bases

Definition 3. *A* Hamel basis *(often just called a* basis*) of a vector space X is a linearly independent set of vectors in X that spans X.*

Example: $\{(1,0),(0,1)\}$ is a basis for $\mathbf{R}^2$ (this is the standard basis).

$\alpha, \beta \in \mathbb{R}$:

$\alpha(1,0) + \beta(0,1) = (\alpha, \beta)$

Example, cont: $\{(1,1),(-1,1)\}$ is another basis for $\mathbf{R}^2$:

$$\begin{aligned}
\text{Suppose } (x,y) &= \alpha(1,1)+\beta(-1,1) \text{ for some } \alpha,\beta \in \mathbf{R} \\
x &= \alpha-\beta \qquad = (\alpha-\beta, \alpha+\beta) \\
y &= \alpha+\beta \\
x+y &= 2\alpha \\
\Rightarrow \alpha &= \frac{x+y}{2} \\
y-x &= 2\beta \\
\Rightarrow \beta &= \frac{y-x}{2} \\
(x,y) &= \frac{x+y}{2}(1,1)+\frac{y-x}{2}(-1,1)
\end{aligned}$$

Since (x,y) is an arbitrary element of $\mathbf{R}^2$, $\{(1,1),(-1,1)\}$ spans $\mathbf{R}^2$. If $(x,y)=(0,0)$,

$$\alpha = \frac{0+0}{2} = 0, \quad \beta = \frac{0-0}{2} = 0$$

so the coefficients are all zero, so $\{(1,1),(-1,1)\}$ is linearly independent. Since it is linearly independent and spans $\mathbf{R}^2$, it is a basis.

Example: $\{(1,0,0),(0,1,0)\}$ is not a basis of $\mathbf{R}^3$, because it does not span $\mathbf{R}^3$. (x,y,z) with $z \neq 0$ not in span

Example: $\{(1,0),(0,1),(1,1)\}$ is not a basis for $\mathbf{R}^2$.

$$1(1,0)+1(0,1)+(-1)(1,1)=(0,0)$$

so the set is not linearly independent.

Bases

Theorem 1 (Thm. 1.2'). *Let V be a Hamel basis for X. Then every vector $x \in X$ has a* unique *representation as a linear combination of a finite number of elements of V (with* all *coefficients nonzero).**

Proof. Let $x \in X$. Since V spans X, we can write

$$x = \sum_{s \in S_1} \alpha_s v_s$$

where S_1 is finite, $\alpha_s \in F$, $\alpha_s \neq 0$, and $v_s \in V$ for each $s \in S_1$. Now, suppose

$$x = \sum_{s \in S_1} \alpha_s v_s = \sum_{s \in S_2} \beta_s v_s$$

*The unique representation of 0 is $0 = \sum_{i \in \emptyset} \alpha_i b_i$.

where S_2 is finite, $\beta_s \in F$, $\beta_s \neq 0$, and $v_s \in V$ for each $s \in S_2$. Let $S = S_1 \cup S_2$, and define

$$\begin{aligned} \alpha_s = 0 \quad &\text{for} \quad s \in S_2 \setminus S_1 \\ \beta_s = 0 \quad &\text{for} \quad s \in S_1 \setminus S_2 \end{aligned}$$

Then

$$\begin{aligned} 0 &= x - x \\ &= \sum_{s \in S_1} \alpha_s v_s - \sum_{s \in S_2} \beta_s v_s \\ &= \sum_{s \in S} \alpha_s v_s - \sum_{s \in S} \beta_s v_s \\ &= \sum_{s \in S} (\alpha_s - \beta_s) v_s \end{aligned}$$

Since V is linearly independent, we must have $\alpha_s - \beta_s = 0$, so $\alpha_s = \beta_s$, for all $s \in S$.

$$s \in S_1 \Leftrightarrow \alpha_s \neq 0 \Leftrightarrow \beta_s \neq 0 \Leftrightarrow s \in S_2$$

so $S_1 = S_2$ and $\alpha_s = \beta_s$ for $s \in S_1 = S_2$, so the representation is unique. $\square$

Bases

Theorem 2. *Every vector space has a Hamel basis.*

Proof. The proof uses the Axiom of Choice. Indeed, the theorem is equivalent to the Axiom of Choice. □

Bases

A closely related result, from which you can derive the previous result, shows that any linearly independent set V in a vector space X can be extended to a basis of X.

Theorem 3. *If X is a vector space and $V \subseteq X$ is linearly independent, then there exists a linearly independent set $W \subseteq X$ such that*

$$V \subseteq W \subseteq \operatorname{span} W = X$$

Bases

Theorem 4. *Any two Hamel bases of a vector space X have the same cardinality (are numerically equivalent).*

Proof. The proof depends on the so-called Exchange Lemma, whose idea we sketch. Suppose that $V = \{v_\lambda : \lambda \in \Lambda\}$ and $W = \{w_\gamma : \gamma \in \Gamma\}$ are Hamel bases of X. Remove one vector v_{λ_0} from V, so that it no longer spans (if it did still span, then v_{λ_0} would be a linear combination of other elements of V, and V would not be linearly independent). If $w_\gamma \in \operatorname{span}(V \setminus \{v_{\lambda_0}\})$ for every $\gamma \in \Gamma$, then since W spans, $V \setminus \{v_{\lambda_0}\}$ would also span, contradiction. Thus, we can choose $\gamma_0 \in \Gamma$ such that

$$w_{\gamma_0} \notin \operatorname{span}\left(V \setminus \{v_{\lambda_0}\}\right)$$

Because $w_{\gamma_0} \in \overset{\overset{X}{\shortparallel}}{\operatorname{span} V}$, we can write

$$w_{\gamma_0} = \sum_{i=0}^{n} \alpha_i v_{\lambda_i}$$

where α_0, the coefficient of v_{λ_0}, is not zero (if it were, then we would have $w_{\gamma_0} \in \operatorname{span}\left(V \setminus \{v_{\lambda_0}\}\right)$). Since $\alpha_0 \neq 0$, we can solve for v_{λ_0} as a linear combination of w_{γ_0} and $v_{\lambda_1}, \ldots, v_{\lambda_n}$, so

$$\begin{aligned} &\operatorname{span}\left(\left(V \setminus \{v_{\lambda_0}\}\right) \cup \{w_{\gamma_0}\}\right) \quad \ni v_{\lambda_0} \\ &\quad \supseteq \operatorname{span} V \quad = \operatorname{span}\left(\left(V \setminus \{v_{\lambda_0}\}\right) \cup \{v_{\lambda_0}\}\right) \\ &\quad = X \end{aligned}$$

so

$$\left(\left(V \setminus \{v_{\lambda_0}\}\right) \cup \{w_{\gamma_0}\}\right)$$

spans X. From the fact that $w_{\gamma_0} \notin \operatorname{span}\left(V \setminus \{v_{\lambda_0}\}\right)$ one can

V W

$n < m$

$\exists\, w_{\gamma_1} \in W \setminus \{w_{\gamma_0}\}$

$W \setminus \{w_{\gamma_0}, w_{\gamma_1}\}$

show that

$$\Big(\big(V \setminus \{v_{\lambda_0}\}\big) \cup \{w_{\gamma_0}\}\Big)$$

is linearly independent, so it is a basis of X. Repeat this process to exchange every element of V with an element of W (when V is uncountable, this is done by a process called transfinite induction). At the end, we obtain a bijection from V to W, so that V and W are numerically equivalent. $\square$

Dimension

Definition 4. *The* dimension *of a vector space* X*, denoted* dim X*, is the cardinality of any basis of* X*.*

For $V \subseteq X$, $|V|$ denotes the cardinality of the set V.

- If $\dim X = n$ for some $n \in \mathbb{N}$, X is <u>finite-dimensional</u>.

Otherwise, X is <u>infinite-dimensional</u>.

Dimension

Example: The set of all $m \times n$ real-valued matrices is a vector space over $\mathbf{R}$. A basis is given by

$$\{E_{ij} : 1 \leq i \leq m, 1 \leq j \leq n\}$$ E_{ij} $m \times n$ matrix

where

$k\ell^{\text{th}}$ entry in matrix E_{ij}

$$\left(E_{ij}\right)_{k\ell} = \begin{cases} 1 & \text{if } k = i \text{ and } \ell = j \\ 0 & \text{otherwise.} \end{cases}$$

The dimension of the vector space of $m \times n$ matrices is mn.

$$E_{ij} = \begin{array}{c c} & \ell: \; 1 \; \cdots \; j \; \cdots \; n \\ k: \begin{array}{c} 1 \\ \vdots \\ i \\ \vdots \\ m \end{array} & \begin{pmatrix} 0 & \cdots & 0 & \cdots & 0 \\ \vdots & & \vdots & \ddots & \vdots \\ 0 & \cdots & \vdots & \cdots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & \cdots & 0 \end{pmatrix} \end{array}$$

Dimension and Dependence

Theorem 5 (Thm. 1.4)**.** *Suppose* $\dim X = n \in \mathbf{N}$*. If* $V \subseteq X$ *and* $|V| > n$*, then* V *is linearly dependent.*

If not, V is linearly independent, so V can be extended to a basis W of X, and

$$V \subseteq W \Rightarrow n < |V| \leq |W|$$

contradiction.

Dimension and Dependence

Theorem 6 (Thm. 1.5'). *Suppose* dim $X = n \in \mathbf{N}$, $V \subseteq X$, *and* $|V| = n$.

- *If V is linearly independent, then V spans X, so V is a Hamel basis.*

- *If V spans X, then V is linearly independent, so V is a Hamel basis.*

① otherwise, extend V to a basis W of X, with $V \subsetneq W$,
so $|W| > |V| = n$
contradiction.

② otherwise, choose $V' \subsetneq V$ a basis for X, and
$|V'| < |V| = n$

contradiction

Linear Transformations

Definition 5. *Let X and Y be two vector spaces over the field F. We say $T : X \to Y$ is a* linear transformation *if*

$$T(\underbrace{\alpha_1 x_1 + \alpha_2 x_2}_{x \in X}) = \alpha_1 T(x_1) + \alpha_2 T(x_2) \quad \forall x_1, x_2 \in X, \alpha_1, \alpha_2 \in F$$

$$\underbrace{\alpha_1 y_1 + \alpha_2 y_2}_{y \in Y}$$

$y_1 = T(x_1)$

$y_2 = T(x_2)$

Let $L(X, Y)$ denote the set of all linear transformations from X to Y.

Equivalently:

- $T(\alpha x) = \alpha T(x) \quad \forall \alpha \in F, \forall x \in X$
- $T(x_1 + x_2) = T(x_1) + T(x_2) \quad \forall x_1, x_2 \in X$

$T \in L(X,Y)$ $T(x) \in Y$

Linear Transformations

Theorem 7. *$L(X, Y)$ is a vector space over F.*

Proof. First, define linear combinations in $L(X, Y)$ as follows. For $T_1, T_2 \in L(X, Y)$ and $\alpha, \beta \in F$, define $\alpha T_1 + \beta T_2$ by

$$(\alpha T_1 + \beta T_2)(x) = \alpha T_1(x) + \beta T_2(x)$$

We need to show that $\alpha T_1 + \beta T_2 \in L(X, Y)$.

$$\begin{aligned}
&(\alpha T_1 + \beta T_2)(\gamma x_1 + \delta x_2) \\
&= \alpha T_1(\gamma x_1 + \delta x_2) + \beta T_2(\gamma x_1 + \delta x_2) && \text{(definition)} \\
&= \alpha \left(\gamma T_1(x_1) + \delta T_1(x_2)\right) + \beta \left(\gamma T_2(x_1) + \delta T_2(x_2)\right) && (T_1, T_2 \text{ linear}) \\
&= \gamma \left(\alpha T_1(x_1) + \beta T_2(x_1)\right) + \delta \left(\alpha T_1(x_2) + \beta T_2(x_2)\right) && \text{(collect terms)} \\
&= \gamma \left(\alpha T_1 + \beta T_2\right)(x_1) + \delta \left(\alpha T_1 + \beta T_2\right)(x_2) && \text{(definition again)}
\end{aligned}$$

$+ : L(X,Y) \times L(X,Y) \to L(X,Y)$

$\cdot : F \times L(X,Y) \to L(X,Y)$

so $\alpha T_1 + \beta T_2 \in L(X,Y)$.

The rest of the proof involves straightforward checking of the vector space axioms. $\square$

Compositions of Linear Transformations

X, Y, Z vector spaces over same field F

Given $R \in L(X,Y)$ and $S \in L(Y,Z)$, $S \circ R : X \to Z$. We will show that $S \circ R \in L(X,Z)$, that is, the composition of two linear transformations is linear.

$$
\begin{aligned}
(S \circ R)(\alpha x_1 + \beta x_2) &= S(R(\alpha x_1 + \beta x_2)) && \text{(defn of } S \circ R\text{)} \\
&= S(\alpha R(x_1) + \beta R(x_2)) && \text{(R linear)} \\
&= \alpha S(R(x_1)) + \beta S(R(x_2)) && \text{(S linear)} \\
&= \alpha (S \circ R)(x_1) + \beta (S \circ R)(x_2) && \text{(defn of } S \circ R\text{)}
\end{aligned}
$$

so $S \circ R \in L(X,Z)$.

Kernel and Rank

Definition 6. *Let $T \in L(X, Y)$.*

- *The* image *of T is* $\text{Im}\, T = T(X)$ $\subseteq Y$
  - can show $\text{Im}\, T$ is a vector subspace of Y
- *The* kernel *of T is* $\ker T = \{x \in X : T(x) = 0\}$ (null space of T)
- *The* rank *of T is* $\text{Rank}\, T = \dim(\text{Im}\, T)$

Recall:
- $W \subseteq X$ is a vector subspace if it is a vector space over F under $+, \cdot$ from X
- $W \subseteq X$, $W \neq \emptyset$ is a vector subspace $\iff \forall w_1, w_2 \in W, \forall \alpha, \beta \in F$ $\alpha w_1 + \beta w_2 \in W$

Rank-Nullity Theorem

Theorem 8 (Thms. 2.9, 2.7, 2.6: The Rank-Nullity Theorem)**.** *Let X be a finite-dimensional vector space, $T \in L(X, Y)$. Then* $\operatorname{Im} T$ *and* $\ker T$ *are vector subspaces of Y and X respectively, and*

$$\dim X = \underbrace{\dim \ker T}_{\text{nullity of } T} + \operatorname{Rank} T$$

sketch:

- Show $\operatorname{Im} T$, $\ker T$ are vector subspaces
- take $\{v_1, \dots, v_k\}$ a basis for $\ker T$
- extend to $\{v_1, \dots, v_k, w_1, \dots, w_r\}$ a basis for X
- show $\{T(w_1), \dots, T(w_r)\}$ is a basis for $\operatorname{Im} T$

Kernel and Rank

Theorem 9 (Thm. 2.13). $T \in L(X, Y)$ *is one-to-one if and only if* $\ker T = \{0\}$.

$\Rightarrow$: *Proof.* Suppose T is one-to-one. Suppose $x \in \ker T$. Then $T(x) = 0$. But since T is linear, $T(0) = T(0 \cdot 0) = 0 \cdot T(0) = 0$. Since T is one-to-one, $x = 0$, so $\ker T = \{0\}$.

$\Leftarrow$: Conversely, suppose that $\ker T = \{0\}$. Suppose $T(x_1) = T(x_2)$. Then

$$\begin{aligned} T(x_1 - x_2) &= T(x_1) - T(x_2) \\ &= 0 \end{aligned}$$

which says $x_1 - x_2 \in \ker T$, so $x_1 - x_2 = 0$, so $x_1 = x_2$. Thus, T is one-to-one. □

Invertible Linear Transformations

Definition 7. $T \in L(X, Y)$ *is* invertible *if there exists a function* $S : Y \to X$ *such that*

$$\begin{aligned} S(T(x)) &= x \quad \forall x \in X \\ T(S(y)) &= y \quad \forall y \in Y \end{aligned}$$

$S \circ T = id_X$

$T \circ S = id_Y$

Denote S *by* T^{-1}.

Note that T is invertible if and only if it is one-to-one and onto. This is just the condition for the existence of an inverse *function.* The linearity of the inverse follows from the linearity of T.

(we will show this)

Invertible Linear Transformations

Theorem 10 (Thm. 2.11). *If $T \in L(X, Y)$ is invertible, then $T^{-1} \in L(Y, X)$, i.e. T^{-1} is linear.*

Proof. Suppose $\alpha, \beta \in F$ and $v, w \in Y$. Since T is invertible, there exist unique $v', w' \in X$ such that

$$\begin{array}{rclcrcl} T(v') & = & v & \quad & T^{-1}(v) & = & v' \\ T(w') & = & w & & T^{-1}(w) & = & w' \end{array}.$$

Then

$$\begin{aligned} T^{-1}(\alpha v + \beta w) &= T^{-1}\left(\alpha \overset{v}{\overset{\shortparallel}{T(v')}} + \beta \overset{w}{\overset{\shortparallel}{T(w')}}\right) && \text{(definition)} \\ &= T^{-1}\left(T(\alpha v' + \beta w')\right) && \text{(T linear)} \\ &= \alpha v' + \beta w' && \text{(defn of } T^{-1}\text{)} \\ &= \alpha T^{-1}(v) + \beta T^{-1}(w) && \text{(defn of } v', w'\text{)} \end{aligned}$$

so $T^{-1} \in L(Y, X)$. $\square$

Linear Transformations and Bases

Theorem 11 (Thm. 3.2)**.** *Let X and Y be two vector spaces over the same field F, and let $V = \{v_\lambda : \lambda \in \Lambda\}$ be a basis for X. Then a linear transformation $T \in L(X, Y)$ is completely determined by its values on V, that is:*

1. *Given any set $\{y_\lambda : \lambda \in \Lambda\} \subseteq Y$, $\exists T \in L(X, Y)$ s.t.*

$$T(v_\lambda) = y_\lambda \quad \forall \lambda \in \Lambda$$

2. *If $S, T \in L(X, Y)$ and $S(v_\lambda) = T(v_\lambda)$ for all $\lambda \in \Lambda$, then $S = T$.*

Proof. 1. If $x \in X$, x has a unique representation of the form

$$x = \sum_{i=1}^{n} \alpha_i v_{\lambda_i} \quad \alpha_i \neq 0 \;\; i = 1, \ldots, n$$

(Recall that if $x = 0$, then $n = 0$.) Define

$$T(x) = \sum_{i=1}^{n} \alpha_i \overset{= T(v_{\lambda_i})}{y_{\lambda_i}}$$

(so $T(v_\lambda) = y_\lambda \ \forall \lambda$ by defn)

Then $T(x) \in Y$. The verification that T is linear is left as an exercise.

2. Suppose $S(v_\lambda) = T(v_\lambda)$ for all $\lambda \in \Lambda$. Given $x \in X$,

$$\begin{aligned}
S(x) &= S\left(\sum_{i=1}^{n} \alpha_i v_{\lambda_i}\right) \\
&= \sum_{i=1}^{n} \alpha_i S\left(v_{\lambda_i}\right) && \text{(S linear)} \\
&= \sum_{i=1}^{n} \alpha_i T\left(v_{\lambda_i}\right) && \left(\begin{array}{l}\text{S and T agree} \\ \text{on } \{v_\lambda : \lambda \in \Lambda\}\end{array}\right) \\
&= T\left(\sum_{i=1}^{n} \alpha_i v_{\lambda_i}\right) && \text{(T linear)} \\
&= T(x)
\end{aligned}$$

so $S = T$.

$\square$

Isomorphisms

Definition 8. *Two vector spaces X and Y over a field F are* isomorphic *if there is an invertible $T \in L(X, Y)$.*

$T \in L(X, Y)$ *is an* isomorphism *if it is invertible (one-to-one and onto).*

Isomorphic vector spaces are essentially indistinguishable as vector spaces.

Isomorphisms

Theorem 12 (Thm. 3.3)**.** *Two vector spaces X and Y over the same field are isomorphic if and only if* $\dim X = \dim Y$*.*

$\Rightarrow$: *Proof.* Suppose X and Y are isomorphic, and let $T \in L(X, Y)$ be an isomorphism. Let

$$U = \{u_\lambda : \lambda \in \Lambda\}$$

be a basis of X, and let $v_\lambda = T(u_\lambda)$ for each $\lambda \in \Lambda$. Set

$$V = \{v_\lambda : \lambda \in \Lambda\}$$

Since T is one-to-one, U and V have the same cardinality. If

$\shortparallel$

$T(U)$

$y \in Y$, then there exists $x \in X$ such that

$$\begin{aligned} y &= T(x) && \text{(T is onto)} \\ &= T\left(\sum_{i=1}^{n} \alpha_{\lambda_i} u_{\lambda_i}\right) \\ &= \sum_{i=1}^{n} \alpha_{\lambda_i} T\left(u_{\lambda_i}\right) && \text{(linearity of T)} \\ &= \sum_{i=1}^{n} \alpha_{\lambda_i} v_{\lambda_i} && \text{(defn of } v_{\lambda_i}\text{)} \end{aligned}$$

which shows that V spans Y. To see that V is linearly independent, suppose

$$
\begin{aligned}
0 &= \sum_{i=1}^{m} \beta_i v_{\lambda_i} \\
&= \sum_{i=1}^{m} \beta_i T\left(u_{\lambda_i}\right) && \text{(defn of } v_{\lambda_i}\text{)} \\
&= T\left(\sum_{i=1}^{m} \beta_i u_{\lambda_i}\right) && \text{(} T \text{ linear)}
\end{aligned}
$$

Since T is one-to-one, $\ker T = \{0\}$, so

$$\sum_{i=1}^{m} \beta_i u_{\lambda_i} = 0$$

Since U is a basis, we have $\beta_1 = \cdots = \beta_m = 0$, so V is linearly independent. Thus, V is a basis of Y; since U and V are numerically equivalent, $\dim X = \dim Y$.

$$|U| = |V|$$

$\Leftarrow$: Now suppose $\dim X = \dim Y$. Let

$$U = \{u_\lambda : \lambda \in \Lambda\} \text{ and } V = \{v_\lambda : \lambda \in \Lambda\}$$

be bases of X and Y; note we can use the same index set Λ for both because $\dim X = \dim Y$. By Theorem 3.2, there is a unique

previous result

$T \in L(X, Y)$ such that $T(u_\lambda) = v_\lambda$ for all $\lambda \in \Lambda$. If $T(x) = 0$, then

T is 1-1 :

$$\begin{aligned}
0 &= T(x) \\
&= T\left(\sum_{i=1}^{n} \alpha_i u_{\lambda_i}\right) \\
&= \sum_{\text{i}=1}^{n} \alpha_i T\left(u_{\lambda_i}\right) && (T \text{ linear}) \\
&= \sum_{\text{i}=1}^{n} \alpha_i v_{\lambda_i} && (T(u_{\lambda_i}) = v_{\lambda_i} \ \forall i) \\
\Rightarrow&\ \alpha_1 = \cdots = \alpha_n = 0 \text{ since } V \text{ is a basis} \\
\Rightarrow&\ x = 0 && = \sum_{i=1}^{n} \alpha_i u_{\lambda_i} \\
\Rightarrow&\ \ker T = \{0\} \\
\Rightarrow&\ T \text{ is one-to-one}
\end{aligned}$$

T is onto:

If $y \in Y$, write $y = \sum_{i=1}^{m} \beta_i v_{\lambda_i}$. Let

$$x = \sum_{i=1}^{m} \beta_i u_{\lambda_i}$$

Then

$$\begin{aligned} T(x) &= T\left(\sum_{i=1}^{m} \beta_i u_{\lambda_i}\right) \\ &= \sum_{i=1}^{m} \beta_i T(u_{\lambda_i}) \qquad \text{(T linear)} \\ &= \sum_{i=1}^{m} \beta_i v_{\lambda_i} \\ &= y \end{aligned}$$

so T is onto, so T is an isomorphism and X, Y are isomorphic. $\square$

$$f(x) = \begin{cases} 0 & x = 0 \\ \frac{1}{x} & x \in (0, \infty) \end{cases}$$

$$\lim_{x \to 0} f(x) = \infty$$

$$J(f) = \int_0^1 e^{-x} \frac{f(x)}{1 + f'(x)^2} \, dx$$

$$|J(f) - J(g)| = \left| \int_0^1 e^{-x} \left(\frac{f(x)}{1 + f'(x)^2} - \frac{g(x)}{1 + g'(x)^2} \right) dx \right|$$

$$\leq \int_0^1 e^{-x} \left| \frac{f(x)(1 + g'(x)^2) - g(x)(1 + f'(x)^2)}{(1 + f'(x)^2)(1 + g'(x)^2)} \right| dx$$

$$\le \int_0^1 e^{-x} \left| f(x)(1+g'(x)^2) - g(x)(1+f'(x)^2) \right| dx$$

$$\le \int_0^1 e^{-x} \left| f(x)-g(x) + (f(x)g'(x)^2 - g(x)f'(x)^2) \right| dx$$

$$\rho(x, A) = 0 \Rightarrow x \in \bar{A}$$

$$\rho(x, A) = \inf_{a \in A} d(x, a)$$

$A = \{\bar{a}\}$

$a_n = \bar{a} \quad \forall n$

$\forall n \quad \exists \, a_n \in A \quad$ s.t.

$$0 \le d(x, a_n) \le \frac{1}{n}$$

$$\Rightarrow a_n \to x$$

$$\{a_n\} \subseteq A \Rightarrow x \in \bar{A}$$

$\bar{A} = A \cup$ limit points of A

$x \in \bar{A} \Rightarrow \quad d(x, A) = 0$

$Y \subseteq A \cup B$, $Y \cap A \neq \emptyset$, $Y \cap B \neq 0$

$\bigcup_{x \neq y} [x]$ vs. $[y]$

$[x]$ $\quad y \notin [x]$

$\exists \varepsilon > 0$ s.t. $B_\varepsilon(y) \cap [x] = \emptyset$

$[x]$ open

$\Rightarrow \exists \varepsilon_x > 0$ s.t. $B_{\varepsilon_x}(x) \subseteq O_x$

$\overline{\forall x}$ $\exists$ open O_x, $x \in O_x$, s.t. $O_x \subset [x]$

$z \in [x] \Rightarrow [z] = [x]$

$\Rightarrow \exists$ open O_z, $z \in O_z$, $O_z \subseteq [x] = [z]$

Suppose $\exists\, x, y$ s.t. $y \notin [x]$

$$Y = X = \bigcup_{x \in X} [x] = \left(\bigcup_{x \neq y} [x] \right) \cup [y]$$

$$[y] \cap \left(\bigcup_{x \notin [y]} [x] \right) = \emptyset$$