

What is meant by a "natural mapping"?

$$V \rightarrow V^{**} \quad v \mapsto (f \mapsto f(v))$$

$$V \rightarrow V/U \quad v \mapsto v + U$$

this has both an informal and a categorical meaning.

Polynomials over a field F

For our purposes now, you can just think of a polynomial over F as a function $p: F \rightarrow F$ which can be written

$$p(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n$$

for $a_0, a_1, \dots, a_n \in F$, $a_n \neq 0$. n is called the degree of p .

Fact: If p has a root λ , $p(x) = (x - \lambda)q(x)$ where q is an $n-1$ st degree polynomial.

Proof: Let $p(\lambda) = 0$. $d(x) \equiv p(\lambda + x) = a_0 + a_1(\lambda + x) + \dots + a_n(\lambda + x)^n$

$$\Rightarrow d(x) = b_0 + b_1 x + b_2 x^2 + \dots + b_n x^n \text{ with } b_n \neq 0 \text{ and } b_0 = 0 \text{ (since } d(0) = p(\lambda) = 0\text{).}$$

$$\Rightarrow d(x) = x(b_1 + b_2 x + \dots + b_n x^{n-1}), \quad b_n \neq 0$$

$$\Rightarrow p(x) = d(x - \lambda) = (x - \lambda) \cdot (b_1 + b_2 x + \dots + b_n x^{n-1}).$$

Fact: An n th degree polynomial can have at most n roots.

Proof: (homework).

Theorem: Every nonconstant complex polynomial has a root.

Proof: This is the "fundamental theorem of algebra". We'll prove it only once we get to manifolds.

Let's let " $\text{Poly}_{\leq n}$ " denote the set of polynomials with degree $\leq n$.

Defining addition and scalar multiplication in the obvious way, we have

Fact: $\text{Poly}_{\leq n}$ is an $n+1$ -dimensional vector space with basis $\{1, x, x^2, \dots, x^n\}$.

Proof. $\{1, x, x^2, \dots, x^n\}$ obviously spans $\text{Poly}_{\leq n}$. If

$$a_0 \cdot 1 + a_1 \cdot x + a_2 x^2 + \dots + a_n x^n = 0$$

then $a_1 = a_2 = \dots = a_n = 0$ since, otherwise the polynomial would have an infinite number of roots and we just proved that it can have at most n roots. $\Rightarrow a_0 = 0$ also \Rightarrow basis.

Application: Solve $\frac{d^3 f}{dx^3} + \frac{d^2 f}{dx^2} - f(x) = 0$ $f: \mathbb{R} \rightarrow \mathbb{R}$

We can actually do better. Let

$$a_m \frac{df}{dx^n} + a_{m-1} \frac{d^{n-1}f}{dx^{n-1}} + \dots + a_0 f = 0$$

$V = \{ \text{continuous functions from } \mathbb{R} \rightarrow \mathbb{R} \text{ with continuous derivatives of all orders} \}$

Think of $\frac{d}{dx} \equiv L$ as an operator $V \xrightarrow{L} V$.

$$(a_m L^m + a_{m-1} L^{m-1} + \dots + a_0 L^0) f = 0$$

This can just be factored like a polynomial

$$a (L - \lambda_1)^{m_1} (L - \lambda_2)^{m_2} \dots (L - \lambda_k)^{m_k} f = 0$$

\Rightarrow The space of solutions is exactly the kernel of

$$\text{Ker}[(L - \lambda_1)^{m_1} (L - \lambda_2)^{m_2} \dots (L - \lambda_k)^{m_k}]$$

$$= \text{Ker}(L - \lambda_1)^{m_1} \oplus \text{Ker}(L - \lambda_2)^{m_2} \oplus \dots \oplus \text{Ker}(L - \lambda_k)^{m_k}$$

You can easily guess that the solution space $\text{Ker}(\frac{d}{dx} - \lambda \cdot 1)^{\oplus m}$ is spanned by $\{e^{\lambda x}, x e^{\lambda x}, x^2 e^{\lambda x}, \dots, x^{m-1} e^{\lambda x}\}$.

\Rightarrow Thus, we have found all solutions to

$$a_m \frac{df}{dx^n} + a_{m-1} \frac{d^{n-1}f}{dx^{n-1}} + \dots + a_0 f = 0 !$$

\Rightarrow Space of solutions is $m_1 + m_2 + \dots + m_k$ dimensional.

Since Ker
 $\text{Ker } A \circ B \circ C$
 $= \text{Ker } A \oplus \text{Ker } B \oplus \text{Ker } C$
 if A, B, C
 commute.

More tensors

We now have quite a few ways to construct vector spaces:

$S \rightarrow V$ free construction

$V \xrightarrow{\varphi} W$ Ker φ and Im φ

$V \oplus W$ Direct product / sum

V/U Quotient with a subspace

V^* Duals

$\text{Mor}(V, W)$ Maps with pointwise addition

Last time, we added the tensor product $V \otimes W$ to this list:

$$\begin{array}{ccccc} V \times W & \xrightarrow{\alpha} & F & \longrightarrow & F/A \equiv V \otimes W \\ & \searrow \mu & \downarrow \bar{\mu} & \nearrow \bar{\mu} & \\ & & Z & & "vw" \equiv \alpha(v, w) \end{array}$$

$A \equiv$ The vector space generated by

$$\left\{ \begin{array}{l} (v+v')w - vw - v'w, \\ v(w+w') - vw - v w', \\ (av)w - a(vw) \\ v(aw) - a(vw) \end{array} \right. \quad \begin{array}{l} v, v' \in V, w, w' \in W \\ a \in \text{the field} \end{array}$$

This "forces $\bar{\mu}$ to be bilinear."

Let's look at this more explicitly...

$$V \times W \longrightarrow F \longrightarrow V \otimes W$$

(v, w) $a_1 v_1 w_1 + a_2 v_2 w_2 + \dots + a_n v_n w_n$

$a_n \in \text{dim. vector space}$

$\hookrightarrow a_1 v_1 w_1 + a_2 v_2 w_2 + \dots + a_n v_n w_n + A \in F/A$

$\equiv V \otimes W$

ex.

$$vw + vw' + A \quad \text{a coset of } A \text{ in } F$$

$$= v(w+w') - v(w+w') + vw + vw' + A$$

$$= v(w+w') + A$$

It's easy! "A". just absorbs every thing that you would cancel by hand if you assumed that the product was bilinear.

Often, the "A" is dropped and the product is written with \otimes as in

$$v(w+w') + A \equiv v \otimes (w+w')$$

The bottom part of the diagram

$$V \times W \rightarrow F \rightarrow F/A = V \otimes W$$

$$\bar{\mu}(a_1v_1w_1 + a_2v_2w_2) = a_1\mu(v_1, w_1) + a_2\mu(v_2, w_2)$$

↑ guaranteed to be linear by the freeness of F .

$$\bar{\bar{\mu}}(v \otimes (w+w')) = \bar{\bar{\mu}}(v \otimes w + v \otimes w') =$$

$$\uparrow \bar{\mu}(v \otimes w) + \bar{\mu}(v \otimes w') = \mu(v, w) + \mu(v, w')$$

guaranteed to be linear and to annihilate A .

~~if $\{v_i\}$ and $\{w_j\}$ are bases, then~~, ~~then~~ $\bar{\bar{\mu}}$ is well-defined.

If $K \subset V$ and $L \subset W$ are bases, clearly any

$$v \otimes w = (\sum_i a_i k_i) \otimes (\sum_j b_j l_j) = \sum_i \sum_j a_i b_j (k_i \otimes l_j)$$

\Rightarrow and any element of $V \otimes W$ can be written as a finite linear combination of $k \otimes l$ with $k \in K, l \in L$. $\Rightarrow V \otimes W$ is the free vector space on $K \times L$ as a set.

OK, but what are tensors good for?

ex: You can represent mappings $g \in \text{Mor}(V, W)$ as tensors in $V^* \otimes W$

$$(a_1 f_1 \otimes w_1 + a_2 f_2 \otimes w_2 + \dots + a_n f_n \otimes w_n) : v \mapsto a_1 f_1(v) w_1 + a_2 f_2(v) w_2 + \dots + a_n f_n(v) w_n$$

In the finite dimensional case, this is an isomorphism
 $V^* \otimes W \cong \text{Mor}(V, W)$ [$\Rightarrow \text{Mor}(V, W)$ is a $\dim(V) \cdot \dim(W)$
dimensional vector space, incidentally].

For example: $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$

can be written as

$$\cos \theta x \otimes f_x - \sin \theta y \otimes f_x + \sin \theta x \otimes f_y + \cos \theta y \otimes f_y \in (\mathbb{R}^2)^* \times \mathbb{R}^2$$

example: $V = \mathbb{R}^4$, the "Lorentz metric tensor" is

$$dx \otimes dx + dy \otimes dy + dz \otimes dz - dt \otimes dt = g \in V^* \otimes V^*$$

sometimes written " $dx^2 + dy^2 + dz^2 - dt^2$ " in G.R. books.

example: The quaternions H are usually introduced as a 4-d real vector space with basis $1, i, j, k$ with a bilinear associative product defined by

$$i^2 = j^2 = k^2 = -1 \quad ij = -ji = k, \quad jk = -kj = i, \quad ki = -ik = j$$

but what does this mean exactly?

Let H be the free vector space on $\{1, i, j, k\}$.

Bilinear products on H are isomorphic to $M_4(H \otimes H, H)$ via

$$\begin{array}{ccc} H \times H & \longrightarrow & H \otimes H \\ & \searrow \text{Product} & \downarrow \text{product} \\ & & H \end{array}$$

and these are isomorphic to H valued functions on

$$\begin{array}{ccc} \{1, i, j, k\} \times \{1, i, j, k\} & \longrightarrow & H \otimes H \\ & \searrow \times & \downarrow \\ & & H \end{array}$$

so $\{\text{bilinear products on } H\} \cong \{4 \times 4 H \text{ valued matrices}\}$

\Rightarrow Bilinear products on H form a 64-dimensional vector space. To define the product we have to supply a

4×4 matrix:

	1	i	j	k
1	1	i	j	k
i	i	-1	k	-j
j	j	-k	-1	i
k	k	j	-i	-1

Let's try the same line of reasoning for functions having more properties than just being bilinear. For example, consider

$$\text{area}(v, v') = \begin{array}{c} v \\ \swarrow \quad \searrow \\ v' \end{array} \quad \begin{array}{l} \text{area in} \\ \text{parallelogram} \in \mathbb{R} \\ v, v' \in \mathbb{R}^3 = V \end{array}$$

area is evidently bilinear. Also, however $\text{area}(0, v') = \text{area}(v, 0) = 0$
 $\Rightarrow \text{area}(v, v') = -\text{area}(v', v)$ it is anti-symmetric.

$$\begin{array}{ccc} V \times V & \xrightarrow{\quad} & F & \xrightarrow{\quad} & F/B \\ & \searrow \text{area} & \downarrow & \nearrow \overline{\text{area}} & \\ & & Z & & \end{array} \quad B = \left\{ \begin{array}{l} \text{Subspace of } F \\ \text{generated by previous } A \\ \text{and all elements} \\ vv' + v'v \end{array} \right\}$$

$F/B = "V \wedge V"$ "exterior" or "wedge" product

$$\begin{array}{ccc} V \times V & \xrightarrow{\quad} & F & \xrightarrow{\quad} & F/B = V \wedge V \\ \underbrace{(v, v')} & & \underbrace{q_1 v_1 v'_1 + q_2 v_2 v'_2 + \dots + q_n v_n v'_n} & & \underbrace{a_1 v_1 \wedge v'_1 + a_2 v_2 \wedge v'_2 + a_3 v_3 \wedge v'_3 + \dots + a_n v_n \wedge v'_n} \end{array}$$

$$v \wedge v = 0$$

$$v \wedge v' = -v' \wedge v \text{ etc.}$$

You can easily see that if $\{x, y, z\}$ is a basis for $V = \mathbb{R}^3$,
 $\{x \wedge y, y \wedge z, z \wedge x\}$ is a basis for $V \wedge V$.

These spaces have a nice geometric flavor because of their relationship to area or volume. Something very nice happens with $V \wedge V \wedge V \wedge \dots \wedge V \dots$

Although we've done a lot of wedging,

$$\underbrace{V \wedge V \wedge V \wedge \dots \wedge V}_{n \text{ times}} \quad \dim V = n \\ \text{basis} = \{\epsilon_1, \epsilon_2, \dots, \epsilon_n\}$$

is actually one-dimensional! To see this, notice that

$$v_1 \wedge v_2 \wedge \dots \wedge v_n = \left(\sum_{i=1}^m a_i \epsilon_i \right) \wedge \left(\sum_{i=1}^m b_i \epsilon_i \right) \wedge \dots \wedge \left(\sum_{i=1}^m c_i \epsilon_i \right)$$

expanded contains only terms with each basis vector occurring exactly once (since $\epsilon_i \wedge \epsilon_i = 0$). Using $\epsilon_i \wedge \epsilon_j = -\epsilon_j \wedge \epsilon_i$, all such terms can be re-arranged in a standard order, so

$$v_1 \wedge v_2 \wedge \dots \wedge v_n = a_1 \epsilon_1 \wedge \epsilon_2 \wedge \dots \wedge \epsilon_m$$

$\wedge^n V$ is one dimensional. Now, let $\varphi \in \text{Hom}(V, V)$ and

consider

$$v_1 \wedge v_2 \wedge \dots \wedge v_n \mapsto \varphi(v_1) \wedge \varphi(v_2) \wedge \dots \wedge \varphi(v_n) = K_\varphi v_1 \wedge v_2 \wedge \dots \wedge v_n$$

Since $\wedge^n V$ is one dimensional, the action of the map is just multiplication by some constant K_φ . You can easily see that this constant only depends on φ and is independent of which $v_1 \wedge v_2 \wedge \dots \wedge v_n$ you start with [proof: Any

$$w_1 \wedge w_2 \wedge \dots \wedge w_n = a v_1 \wedge v_2 \wedge \dots \wedge v_n \text{ for some real } a \in \mathbb{R}.$$

$$\begin{aligned} \varphi(w_1 \wedge w_2 \wedge \dots \wedge w_n) &= \varphi(av_1) \wedge \varphi(v_2) \wedge \dots \wedge \varphi(v_n) = a K_\varphi v_1 \wedge v_2 \wedge \dots \wedge v_n \\ &= K_\varphi (w_1 \wedge w_2 \wedge \dots \wedge w_n). \end{aligned}$$

K_φ is called "the determinant" of $\varphi = \det \varphi$

Intuitively, it is clear that $\det \varphi$ is the "volume squashing" factor of φ . We expect, for example, for volume preserving maps like rotations to have $\det \varphi = 1$.

It is also now obvious that $\det i_V = 1$ and

$$\det(\varphi \circ \psi) = \det(\varphi) \cdot \det(\psi) \quad \varphi, \psi \in \text{M}_n(V, V)$$

making it a group homomorphism to $\mathbb{R}^{\neq 0}$ as we have noted (or $\mathbb{C}^{\neq 0}$ in the complex case).

Theorem: $\varphi \in \text{M}_n(V, V)$ is an isomorphism iff $\det \varphi \neq 0$.

Proof: If φ is an isomorphism, $\varphi \circ \varphi^{-1} = i_V \Rightarrow \det(\varphi) \cdot \det(\varphi^{-1}) = 1 \Rightarrow \det \varphi \neq 0$. Conversely, suppose that φ has some non-zero $v \in V$ in its kernel. Choose a basis $v, e_2, e_3, \dots, e_n \Rightarrow v \in \text{Ker } \varphi$. Then $\varphi(v \wedge e_2 \wedge e_3 \wedge \dots \wedge e_n) = \varphi(v) \wedge \varphi(e_2) \wedge \dots \wedge \varphi(e_n) = 0 = \det \varphi \cdot v \wedge e_2 \wedge \dots \wedge e_n$ $\Rightarrow \det \varphi = 0$.

Fact. An "eigenvector" of φ is a nonzero $v \in V$ s.t. $\varphi(v) = \lambda v$ for $\lambda \in \mathbb{R}$. $\Rightarrow v \in \text{Ker}(\varphi - \lambda \cdot 1) \Rightarrow \det(\varphi - \lambda \cdot 1) = 0$.

Isn't that easy?

These volume forms will reappear as "differential forms" when we do integration on manifolds.

Speaking of Eigenvalues, why are we always looking for them?

$$\underbrace{\text{End}(V) = M_n(V, V)}_{\dim V = n}$$

If you are trying to understand $L \in \text{End } V$, it helps to know an eigenvector $Lv = \lambda v$, $v \neq 0$ because it gives you an invariant subspace $[v] = \{av : a \in F\}$. You can then split

$$V \cong [v] \oplus U \quad \text{where } U \text{ is complementary to } [v].$$

At least for some vectors, then, you know exactly what L does; it just multiplies by λ . The more eigenvectors we know, the more we can split V into one-dimensional subspaces where the action of L is trivial. In the ideal case we might be able to find n different eigenvectors so that V entirely splits up

$$V \cong [v_1] \oplus [v_2] \oplus [v_3] \oplus \dots \oplus [v_n]$$

so that we have an entire basis of eigenvectors and the action of L is very easy to describe. This is often, but not always achievable [rotation by 90° in \mathbb{R}^2 has, for example no eigenvectors at all].

This is a subject where the field makes a difference. The complex results are simpler, so we take $F = \mathbb{C}$ from now on.

Theorem: Every operator on an n -dimensional complex vector space has an eigen vector.

Proof: Let $L \in \text{End}(V)$ and choose any nonzero v .

If $\dim V = n$, choose $n+1$ vectors like so

$$v, Lv, L^2v, \dots, L^n v$$

Since $\dim V = n$, these vectors must be dependent i.e. there must be $a_0, a_1, \dots, a_n \in \mathbb{C}$ not all zero such that

$$a_0 v + a_1 Lv + a_2 L^2v + \dots + a_n L^n v = 0$$

However, we can factor this like a polynomial \Rightarrow

$$a(L - \lambda_1)(L - \lambda_2)\dots(L - \lambda_k)v = 0 \quad a \neq 0$$

\Rightarrow one of the factors in the chain must zero: it's argument

$$\Rightarrow (L - \lambda_k)v = 0 \quad \text{for some nonzero } v.$$

Theorem: Eigenvectors with different eigenvalues are independent.

Proof: Suppose $Lv_i = \lambda_i v_i$ with $v_i \neq 0$ and all the λ_i different. Let $i = 1, \dots, m$. Suppose that v_1, v_2, \dots, v_m are dependent. Let v_k be the first vector spanned by its predecessors. Then

$$v_k = a_1 v_1 + a_2 v_2 + \dots + a_{k-1} v_{k-1}$$

Applying $(L - \lambda_k)$ to both sides,

$$0 = a_1(\lambda_1 - \lambda_k)v_1 + a_2(\lambda_2 - \lambda_k)v_2 + \dots + a_{k-1}(\lambda_{k-1} - \lambda_k)v_k$$

$$\Rightarrow a_1 = a_2 = \dots = a_{k-1} = 0 \quad (\text{since the } \lambda_i \text{ are all different}) \Rightarrow v_k = 0 \Rightarrow \Leftarrow.$$

\Rightarrow the whole list v_1, v_2, \dots, v_k must be independent.

Analogous results for real vector spaces are slightly harder to prove (see Axler's book).

Theorem: Every operator on a finite dimensional nonzero real vector space has an invariant subspace of dimension 1 or 2.

Theorem: Every operator on an odd-dimensional real vector space has an eigenvalue.

This is why, for example "coordinate axes can be 2-dimensional" in Minkowski space (see Geroch's Minkowski space chapter).

Inner Product Spaces

A vector space V with a function $\langle \cdot, \cdot \rangle : V \times V \rightarrow F$ is an inner product space if

$$\langle v + av', w \rangle = \langle v, w \rangle + a \langle v', w \rangle \quad \text{linearity in the 1st arg.}$$

$$\langle v, v \rangle \geq 0 \quad \langle v, v \rangle = 0 \Leftrightarrow v = 0 \quad \text{positivity}$$

$$\langle v, w \rangle = \langle w, v \rangle^*$$

The associated norm is $\|v\| \equiv \langle v, v \rangle^{1/2}$. $\langle v, w \rangle = 0 \Leftrightarrow v \perp w$

Warning: Definitions of the term "inner product" vary a bit. By this definition, for example the Minkowski metric is not an inner product.

Fact (Pythagoras) : $v \perp w \Rightarrow \|v+w\|^2 = \|v\|^2 + \|w\|^2$.

Fact (Cauchy-Schwartz) : $\langle v, w \rangle \leq \|v\| \|w\|$.

Fact (Non degeneracy) : $\langle v, x \rangle = 0 \text{ for all } x \in V \Rightarrow v = 0$.

Fact (Triangle) : $\|v+w\| \leq \|v\| + \|w\|$

Examples:

$$V = \mathbb{R}^3, \langle v, w \rangle = v_x w_x + v_y w_y + v_z w_z$$

$$V = \mathbb{C}^3, \langle v, w \rangle = v_x w_x^* + v_y w_y^* + v_z w_z^*$$

$V = \{ \text{continuous real functions on } [0,1] \}$

$$\langle f, g \rangle = \int_0^1 f(x) g(x) dx$$

Fact: Gram-Schmidt algorithm

Given a basis $e_1, e_2, \dots, e_n \in V$, we can construct an orthonormal basis by the well known Gram-Schmidt algorithm.

$$\gamma_1 = e_1$$

$$\gamma_n = e_n - \sum_{i=1}^{n-1} \langle e_n, \gamma_i \rangle \gamma_i / \langle \gamma_i, \gamma_i \rangle$$

Fact: If U is a subspace of V , $U \oplus U^\perp$ are complementary with

$$U^\perp = \{ w \in V : \langle u, w \rangle = 0 \text{ for all } u \in U \}$$

Fact (Parallelogram) : $\|v+w\|^2 + \|v-w\|^2 = 2(\|v\|^2 + \|w\|^2)$.

Theorem: Suppose U is a subspace of V and $v \in V$.

Then $\|v - \pi_U v\| \leq \|v - u\|$ for all $u \in U$ with equality iff $u = \pi_U v$.

Proof. Suppose $u \in U$,

$$\begin{aligned}\|v - \pi_U v\|^2 &\leq \|v - \pi_U v\|^2 + \|\pi_U v - u\|^2 \\&= \|(v - \pi_U v) + (\pi_U v - u)\|^2 \quad (\text{Pythagoras}) \\&= \|v - u\|^2\end{aligned}$$

Problem: Fit a continuous function on $[0, 1]$ to a 5th degree polynomial. \nwarrow ∞ dimensional

Solution: Let $V = \{\text{real continuous functions on } [0, 1]\}$ with $\langle f, g \rangle = \int_0^1 f(x)g(x) dx$

Let $U \subset V$ be $\text{Poly} \leq 5$.

Let $E_0, E_1, E_2, E_3, E_4, E_5$ be the orthonormal basis of

U gotten from $1, x, x^2, x^3, x^4, x^5$ by Gram-Schmidt.

$\|f - u\|$ is minimized by $u = \pi_U f = \sum_{i=0}^5 \langle f, E_i \rangle E_i$.

This works very well - much better than a Taylor expansion, for example (typically).

Operators on Inner Product Spaces (finite dimensional)

An inner product on a finite dimensional vector space V fixes a natural isomorphism $V \cong V^*$. To see what this is, denote $x \mapsto \langle v, x \rangle$ by " $\langle v, \cdot \rangle \in V^*$ ". Let

$$\Psi: v \mapsto \langle v, \cdot \rangle, \quad \Psi: V \rightarrow V^*$$

so that the nondegeneracy of $\langle \cdot, \cdot \rangle$ is equivalent to $\ker \Psi = \{0\}$.

Given $V \xrightarrow{L} V$, "the" adjoint " L^+ " of L is an operator satisfying

$$\langle w, Lv \rangle = \langle L^+ w, v \rangle \quad v, w \in V$$

We can say "the adjoint" (and promote L to a function) because adjoints are unique. To see this, note that

$$v \mapsto \langle u, Lv \rangle$$

is an element of V^* and must therefore be $\Psi(x_u)$ for some unique $x_u \in V$. $\Rightarrow \Psi(x_u)(v) = \langle x_u, v \rangle = \langle u, Lv \rangle$ for all $u, v \in V$. Let $L^+ u = x_u$.

Adjoints have a number of nice and easily proved properties:

$$(L^+)^+ = L \quad ('^+' is an "involution")$$

$$1^+ = 1, \quad (LM)^+ = M^+ L^+$$

L has an inverse iff L^+ has an inverse ...

An operator which has $L = L^+$ is called self adjoint

"self adjoint" $\equiv \begin{cases} \text{"symmetric" in real vector spaces} \\ \text{"hermitian" in complex vector spaces} \end{cases}$

Self adjoint operators also have nice and easily proved properties.

Fact: Eigenvalues of a self-adjoint operator are real.

Proof: Suppose that $Lv = \lambda v$ is an eigenvector.

$$\langle Lv, v \rangle = \lambda \langle v, v \rangle = \langle v, Lv \rangle = \lambda^* \langle v, v \rangle \Rightarrow \lambda = \lambda^*.$$

Theorem: Every self-adjoint operator has an eigenvector.

Proof: Axler

Def: Operator L is normal if $L^+L = LL^+$.

Complex Spectral Theorem: Let L be an operator on a complex finite dimensional vector space. V has an orthonormal basis of eigenvectors of L iff L is normal.

Real Spectral Theorem: Let L be an operator on a real finite dimensional vector space. V has an orthonormal basis of eigenvectors of L iff L is self adjoint.

An operator L is said to be an isometry if it preserves the norm:

$$\|Lv\| = \|v\| \quad \text{for all } v \in V$$

isometry $\Leftrightarrow \begin{cases} \text{"orthogonal" in real vector spaces} \\ \text{"unitary" in complex vector spaces} \end{cases}$

Fact: $\langle Lv, Lw \rangle = \langle v, w \rangle$ for all $v, w \in V$.

Proof: $\langle Lv, Lw \rangle = \frac{1}{4} (\|Lv + Lw\|^2 - \|Lv - Lw\|^2)$
 $= \frac{1}{4} (\|v + w\|^2 - \|v - w\|^2) = \langle v, w \rangle.$

Fact: L has an inverse and $L^{-1} = L^+$.

Proof: $\langle (L^+ L - 1)v, w \rangle = 0$ for all $w \Rightarrow L^+ L = 1$.

Fact: L is normal.

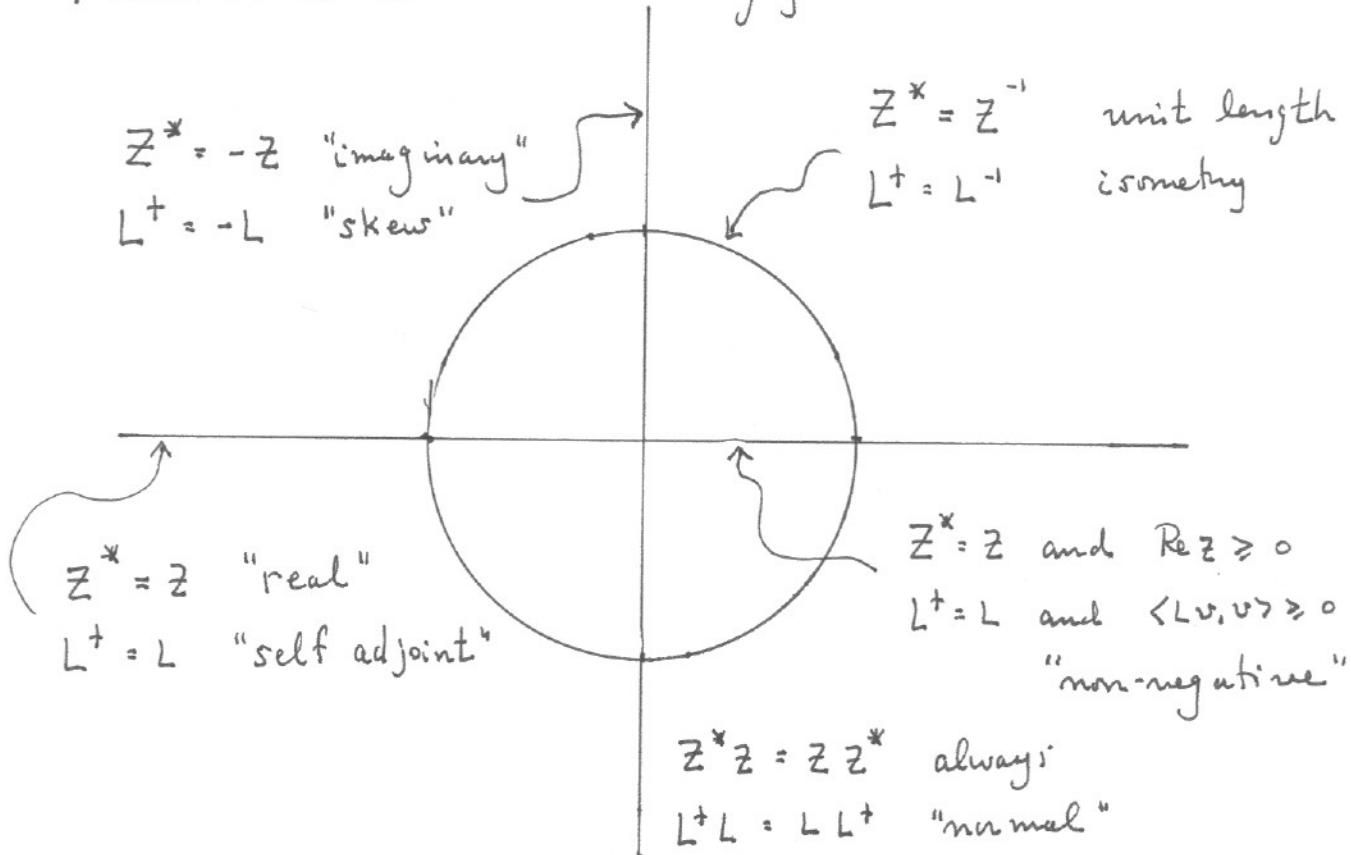
Proof: $L L^+$ is self adjoint \Rightarrow has an inverse. $(L L^+)^{-1} = L^{+^{-1}} L^{-1} = 1$

$$\Rightarrow L L^+ = L^+ L = 1.$$

Fact: L^+ is an isometry.

Proof: $\|L^+ v\| = \langle L^+ v, L^+ v \rangle = \langle v, L L^+ v \rangle = \langle v, w \rangle = \|v\|.$

There is a beautiful analogy:



You can use this to guess what's true about operators.

example: If z is positive, \sqrt{z} has a positive square root.

Theorem: Every positive operator on V has a unique positive square root.

example: $\sqrt{z} = e^{i\theta/2} (z^* z)^{1/2}$ for some θ .

Theorem: Any operator L can be written

$$L = S (L^* L)^{1/2}$$

where S is an isometry.

See Axler for proofs.