## Complex Inner Product Spaces

A vector space $V$ over the field $\mathbb{C}$ of complex numbers is called a (complex) inner product space if it is equipped with a pairing

$$
\langle,\rangle: V \times V \rightarrow \mathbb{C}
$$

which has the following properties:

1. (Sesqui-symmetry) For every $v, w$ in $V$ we have $\langle w, v\rangle=\overline{\langle v, w\rangle}$.
2. (Linearity) For every $u, v, w$ in $V$ and $a$ in $\mathbb{C}$ we have langleau $+v, w\rangle=$ $a\langle u, w\rangle+\langle v, w\rangle$.
3. (Positive definite-ness) For every $v$ in $V$, we have $\langle v, v\rangle \geq 0$. Moreover, if $\langle v, v\rangle=0$, the $v$ must be 0 .

Sometimes the third condition is replaced by the following weaker condition:
3'. (Non-degeneracy) For every $v$ in $V$, there is a $w$ in $V$ so that $\langle v, w\rangle \neq 0$.
We will generally work with the stronger condition of positive definite-ness as is conventional.

A basis $f_{1}, f_{2}, \ldots$ of $V$ is called orthogonal if $\left\langle f_{i}, f_{j}\right\rangle=0$ if $i<j$. An orthogonal basis of $V$ is called an orthonormal (or unitary) basis if, in addition $\left\langle f_{i}, f_{i}\right\rangle=1$.

Given a linear transformation $A: V \rightarrow V$ and another linear transformation $B: V \rightarrow V$, we say that $B$ is the adjoint of $A$, if for all $v, w$ in $V$, we have $\langle A v, w\rangle=\langle v, B w\rangle$. By sesqui-symmetry we see that $A$ is then the adjoint of $B$ as well. Moreover, by positive definite-ness, we see that if $A$ has an adjoint $B$, then this adjoint is unique. Note that, when $V$ is infinite dimensional, we may not be able to find an adjoint for $A!$ In case it does exist it is denoted as $A^{*}$.

A linear transformation $A: V \rightarrow V$ is called a self-adjoint operator on $V$ (with respect to $\langle$,$\rangle ) if we have, for all v$ and $w$ in $V$,

$$
\langle A v, w\rangle=\langle v, A w\rangle
$$

Comparing with the definition above, this is equivalent to the assertion that $A$ is its own adjoint.

An invertible linear transformation $U: V \rightarrow V$ is called a unitary operator on $V$ (with respect to $\langle$,$\rangle ) if we have, for all v$ and $w$ in $V$,

$$
\langle U v, U w\rangle=\langle v, w\rangle
$$

Since $U$ has an inverse $U^{-1}$, we can also write this as

$$
\langle U v, w\rangle=\left\langle v, U^{-1} w\right\rangle
$$

Comparison with the above identity shows that $U^{*}=U^{-1}$.
Given a linear transformation $T: V \rightarrow V$ which has an adjoint $T^{*}$. We say that $T$ is normal if $T^{*}$ commutes with $T$. Note that an operator commutes with itself and with its inverse (if the inverse exists). This shows that self-adjoint operators and unitary operators are normal.

## Gram-Schmidt Orthogonalisation

Given a complex inner product space and a basis $v_{1}, v_{2}, \ldots v_{n}$, we would like to produce and orthogonal/orthonormal basis. The procedure is essentially identical to the earlier one for the real case.

We define $f_{1}=v_{1}$ and put

$$
w_{k}=v_{k}-\frac{\left\langle v_{1}, v_{k}\right\rangle}{\left\langle v_{1}, v_{1}\right\rangle} v_{1} \text { for } k \geq 2
$$

This makes $\left\langle f_{1}, w_{k}\right\rangle=0$ for all $k \geq 2$. We can then repeat the process with $w_{2}, \ldots, w_{n}$. This results in a basis (Exercise: Why is it a basis?) $f_{1}, f_{2}, \ldots, f_{n}$, so that $\left\langle f_{i}, f_{j}\right\rangle=0$ for $i<j$. In other words, this is an orthogonal basis.

Defining $e_{i}=f_{i} /\left\langle f_{i}, f_{i}\right\rangle^{1 / 2}$, we see that this is an orthonormal or unitary basis.
In particular, this shows that any finite dimensional complex inner product space has a orthonormal (unitary) basis.

Suppose $f: V \rightarrow \mathbb{C}$ is a linear map and let $f\left(e_{i}\right)=a_{i}$. We then define the vector $w_{f}=\sum_{i=1}^{n} \overline{a_{i}} e_{i}$; note that this sum does not make sense if $V$ is infinite dimensional!

Now note that $\left\langle e_{i}, w_{f}\right\rangle=f\left(e_{i}\right)$. This can be used to show that $\left\langle v, w_{f}\right\rangle=f(v)$ for all vectors $v$ in $V$ since both sides are linear maps on $V$ and they are equal on the basis vectors. Moreover, if $\langle v, w\rangle=\left\langle v, w^{\prime}\right\rangle$ for all $v$, then it is easy to show that $w=w^{\prime}$. Thus, we have shown that any linear map $f: V \rightarrow \mathbb{C}$ on a finite dimensional complex inner product space is of the form $v \mapsto\langle v, w\rangle$ for a suitably chosen vector $w$ in $V$.

Given an orthonormal basis $e_{1}, \ldots, e_{n}$ of a complex inner product space, let us try to understand the inner product in terms of co-ordinates. Given vectors $v=\sum_{i=1}^{n} v_{i} e_{i}$ and $w=\sum_{i=1}^{n} w_{j} e_{j}$, we see that

$$
\langle v, w\rangle=\sum_{i=1}^{n} v_{i} \overline{w_{i}}=\left(\begin{array}{lll}
\overline{w_{1}} & \ldots & \overline{w_{n}}
\end{array}\right)\left(\begin{array}{c}
v_{1} \\
\vdots \\
v_{n}
\end{array}\right)
$$

Note that this expression will not be correct if we use a basis which is not orthonormal. Now if $U$ is a unitary transformation, and we define $u_{i}=U e_{i}$, the $u_{1}, \ldots, u_{n}$ is another orthonormal basis. Writing $u_{i}=\sum_{j=1}^{n} u_{j, i} e_{j}$ as a column
vector with entries $u_{i, j}$, and using the calculation of the inner-product above, we see that

$$
\left(\begin{array}{ccc}
\overline{u_{1,1}} & \ldots & \overline{u_{n, 1}} \\
\vdots & \ddots & \vdots \\
\overline{u_{n, 1}} & \ldots & \overline{u_{n, n}}
\end{array}\right) \cdot\left(\begin{array}{ccc}
u_{1,1} & \ldots & u_{1, n} \\
\vdots & \ddots & \vdots \\
u_{n, 1} & \ldots & u_{n, n}
\end{array}\right)=\left(\begin{array}{ccc}
1 & \ldots & 0 \\
\vdots & \ddots & \vdots \\
0 & \ldots & 1
\end{array}\right)
$$

In other words, the matrix $M$ associated with a unitary transformation in an orthonormal basis satisfies $\bar{M}^{t} M=1$, i.e. it is a unitary matrix. This is why an orthonormal basis is sometimes also called a unitary basis.
Similarly, let us examine the matrix associated with a self-adjoint transformation $A$. Writing $A e_{i}=\sum_{i=1}^{n} a_{i, j} e_{j}$, we have (since $e_{1}, \ldots, e_{n}$ is orthonormal)

$$
a_{i, j}=\left\langle A e_{i}, e_{j}\right\rangle=\left\langle e_{i}, A e_{j}\right\rangle=\overline{\left\langle A e_{j}, e_{i}\right\rangle}=\overline{a_{j, i}}
$$

Thus the matrix associated with $A$ is Hermitian, i.e. $A=\bar{A}^{t}$.
Now suppose that $A: V \rightarrow V$ is a linear transformation and for a vector $w$ in $V$, let us define the linear functional $f=f_{A, w}: v \mapsto\langle A v, w\rangle$. When $V$ is a finite dimensional complex inner product space, we have seen that there is a vector $w_{f}$ so that $f(v)=\left\langle v, w_{f}\right\rangle$. Thus, we get a set-theoretic map $B: V \rightarrow V$ defined by $w \mapsto w_{f}$. In other words, we have the identity $\langle A v, w\rangle=\langle v, B w\rangle$ for all $v$ and $w$ in $V$. It is now an easy exercise to check that $B$ is a linear map from the complex vector space $V$ to itself. This shows that any linear transformation on a finite-dimensional complex inner-product space has an adjoint. As above, we can easily show that if $M$ is the matrix associated with $A$ in on orthonormal basis, then its adjoint is given by $\bar{M}^{t}$. For that reason we sometimes use the notation $M^{*}=\bar{M}^{t}$.

Note: The above statements converting from linear transformations to matrices are only valid when one uses an orthonormal basis.

## Orthogonal complements and Projections

Given a subspace $W$ of a complex inner-product space $V$, we can define its orthogonal complement

$$
W^{\perp}=\{v \in V \mid\langle v, w\rangle=0 \forall w \in W\}
$$

as the collection of all vectors $v$ whose inner-product with any vector $w$ in $W$ is 0 .

We note that the pairing restricted to $W$ makes it into a complex inner product space. Thus, if $W$ is finite dimensional, then we can find an orthonormal basis $e_{1}, \ldots, e_{m}$ of $W$.

Given any vector $v$ in $V$, we can define $P v$ by the formula

$$
P v=\sum_{i=1}^{m}\left\langle v, e_{i}\right\rangle e_{i}
$$

(Note the similarity with the formula used above for Gram-Schmidt orthogonalisation.) It is clear that $P: V \rightarrow V$ is a linear transformation.

We check that $P e_{i}=e_{i}$ for all $i$ and $P v$ lies in $W$ for every $v$ in $V$. It follows that $P^{2}=P$. So $P$ is an idempotent transformation $V \rightarrow V$. Now,

$$
\left.\langle P v, w\rangle=\sum_{i=1}^{m}\left\langle v, e_{i}\right\rangle e_{i}, w\right\rangle=\sum_{i=1}^{m} \overline{\left.\left\langle w, e_{i}\right\rangle e_{i}, v\right\rangle}=\overline{\langle P w, v\rangle}=\langle v, P w\rangle
$$

In other words, $P$ is also self-adjoint, so it is a self-adjoint idempotent.
As we have seen in our study of idempotents $P$, we can write any vector $v$ as $(1-P) v+P v$, in other words $V$ is decomposed in the image of $P$ and the image of $(1-P)$. Further, if $P$ is self-adjoint, then

$$
\langle(1-P) v, P w\rangle=\langle P(1-P) v, w\rangle=\langle 0, w\rangle=0
$$

Thus, if $W$ is the image of $P$ then the image of $1-P$ is contained in $W^{\perp}$. Thus, $V$ is the orthogonal sum of $W$ and $W^{\perp}$.

In summary, a self-adjoint idempotent $P$ (also called an orthogonal projection) leads to an orthogonal decomposition of the vector space into the image of $P$ and the image of $1-P$. If $W$ is a finite dimensional subspace of $V$, then there is an orthogonal idempotent $P$ whose image is $W$.

## Eigenvector for an operator

Given a linear transformation $A: V \rightarrow V$ on a finite dimensional complex vector space, we can choose a basis of $V$ and represent it as a matrix which we also denote by $A$. The polynomial equation $\operatorname{det}(A-T \cdot 1)=0$ has a solution $T=\lambda$ over the field of complex numbers (by an application of the fundamental theorem of algebra). It follows that $\operatorname{det}(A-\lambda \cdot 1)=0$. Thus, by the usual procedure for solving a finite system of linear equations, we can find a non-zero vector $v$ so that $(A-\lambda \cdot 1) v=0$; equivalently $A v=\lambda v$. Thus, we have proved that a linear transformation $A: V \rightarrow V$ on a finite dimensional complex vector space has an eigenvector.

Exercise: Consider the vector space $V$ consisting of all polynomials in $z^{-1}$ which can be considered as functions on the $\mathbb{C}-\{0\}$. The operator $D: V \rightarrow V$ sends each such polynomial function to its derivative. In other words $z^{-k} \mapsto-k z^{-k-1}$. Check that there is no eigenvector.
Thus, the finite dimensionality of $V$ is important. For infinite dimensional spaces we will have to work harder to find eigenvectors!

## Self-Adjoint Operators

Given a self-adjoint operator $A: V \rightarrow V$ on a complex inner-product space and an eigenvector $v \neq 0$, so that $A v=\lambda v$ for eigenvalue $\lambda$. We then have

$$
\lambda\langle v, v\rangle=\langle\lambda v, v\rangle=\langle A v, v\rangle=\langle v, A v\rangle=\langle v, \lambda v\rangle=\bar{\lambda}\langle v, v\rangle
$$

Since $v \neq 0$, we have $\langle v, v\rangle \neq 0$ so we get $\lambda=\bar{\lambda}$. In other words, we have shown that any eigenvalue of a self-adjoint operator is real.
If $v$ is an eigenvector of $A$, let us consider the space

$$
v^{\perp}=\{w \in V \mid\langle v, w\rangle=0\}
$$

We check that if $w$ is in $v^{\perp}$, then

$$
\langle v, A w\rangle=\langle A v, w\rangle=\lambda\langle v, w\rangle=0
$$

Hence, $A$ sends $v^{\perp}$ to itself. It is obvious that $v^{\perp}$ is a complex inner-product space and that the restriction of $A$ to $v^{\perp}$ is self-adjoint.

In particular, we see by induction on dimension that given a self-adjoint operator $A$ on a finite-dimensional complex inner product space $V$, there is an orthogonal (by scaling we can make it orthonormal) basis of $V$ consisting of eigenvectors of A. Moreover, the eigenvalues are real numbers.

## Unitary Operators

Given a unitary operator $U: V \rightarrow V$ on a complex inner-product space and an eigen vector $v \neq 0$, so that $U v=\lambda v$ for eigenvalue $\lambda$. We then have

$$
\langle v, v\rangle=\langle U v, U v\rangle=\langle\lambda v, \lambda v\rangle=|\lambda|^{2}\langle v, v\rangle
$$

Hence, we see that $|\lambda|^{2}=1$. In other words, $\lambda$ lies on the unit circle $S^{1}$ in the complex plane $\mathbb{C}$. Thus any eigenvalue of a unitary matrix has absolute value 1 .
If $v$ is an eigenvector of $A$, let us consider the space

$$
v^{\perp}=\{w \in V \mid\langle v, w\rangle=0\}
$$

We check that if $w$ is in $v^{\perp}$, then

$$
\lambda\langle v, U w\rangle=\langle U v, U w\rangle=\langle v, w\rangle=0
$$

Since $\lambda \neq 0$ (it has absolute value 1), it follows that $U w$ also lies in $v^{\perp}$. In other words, $U$ takes $v^{\perp}$ to itself. It is obvious, as before, that $v^{\perp}$ is a complex innerproduct space and that the restriction of $U$ to $v^{\perp}$ is a unitary transformation.
In particular, we see by induction on dimension that give a self-adjoint operator $U$ on a finite-dimensional complex inner product space $V$, there is an orthonormal
basis of $V$ consisting of eigenvectors of $A$. Moreover, the eigenvalues are of absolute value 1.

The above results can be extended to normal operators. We will show that a linear transformation of a finite dimensional inner product space to itself is normal if and only if there is an orthonormal basis consisting of eigenvectors of the transformation.

