Useful facts about square matrices

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1 Notation/Definition cheat sheet

Definition 1.1 Let $\mathbf{M}(n, \mathbb{F})$ denote the set of $n \times n$ matrices over field \mathbb{F} .

Specifically, let $\mathbf{M}(n, \mathbb{R})$ denote the set of real $n \times n$ matrices and $\mathbf{M}(n, \mathbb{C})$ the set of complex $n \times n$ matrices.

Definition 1.2 Let $I_n \in \mathbf{M}(n, \mathbb{F})$ be the matrix s.t. for all matrices A in $\mathbf{M}(n, \mathbb{F})$,

$$I_n A = A I_n = A \tag{1}$$

Call this matrix the identity matrix. When n and \mathbb{F} are unambiguous or arbitrary, denote the identity matrix as simply I.

1.1 Matrix Operations

Definition 1.3 For matrix $A \in \mathbf{M}(n, \mathbb{F})$ with elements a_{ij} , let A^T be the transpose of A defined by $[A^T]_{ij} = a_{ji}$. Equivalently,

$$A^{T} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}^{T} = \begin{pmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{pmatrix}$$
(2)

Definition 1.4 For $A \in \mathbf{M}(n, \mathbb{C})$ with elements a_{ij} , define \overline{A} as the <u>conjugate</u> of A defined by the map

$$a_{ij} \mapsto a_{ij}^* \tag{3}$$

ie. take the complex conjugate of each element.

Definition 1.5 For $A \in \mathbf{M}(n, \mathbb{C})$, define A^{\dagger} as the <u>Hermitian conjugate</u> (or conjugate transpose) of A defined by

$$A^{\dagger} = (\overline{A})^T = \overline{A^T} \tag{4}$$

Equivalently defined by the map

$$a_{ij} \mapsto a_{ji}^* \tag{5}$$

Definition 1.6 For matrices $A, B \in \mathbf{M}(n, \mathbb{F})$, let [A, B] be the <u>commutator</u> of A and B defined by

$$[A,B] = AB - BA \tag{6}$$

Similarly, let $\{A, B\}$ be the <u>anti-commutator</u> of A and B defined by

$$\{A, B\} = AB + BA \tag{7}$$

We say A and B commute iff

$$[A,B] = 0 \iff AB = BA \tag{8}$$

and say A and B anti-commute iff

$$\{A, B\} = 0 \iff AB = -BA \tag{9}$$

Definition 1.7 For matrix $A \in \mathbf{M}(n, \mathbb{F})$, let Tr[A] be the <u>trace</u> of A defined by

$$Tr[A] = \sum_{i}^{n} a_{ii} \tag{10}$$

ie. the sum of diagonal elements of A.

1.2 Matrix Properties

Definition 1.8 For $A \in \mathbf{M}(n, \mathbb{R})$, say A is symmetric iff

$$A = A^T \tag{11}$$

and say A is anti-symmetric or skew-symmetric iff

$$A = -A^T \tag{12}$$

Definition 1.9 For $A \in \mathbf{M}(n, \mathbb{C})$, say A is <u>Hermitian</u> iff

$$4 = A^{\dagger} \tag{13}$$

and say A is anti-Hermitian or skew-Hermitian iff

$$A = -A^{\dagger} \tag{14}$$

Definition 1.10 Say $A \in \mathbf{M}(n, \mathbb{F})$ with elements a_{ij} is <u>diagonal</u> iff $a_{ij} = 0 \forall i \neq j$. I.e. A has off-diagonal elements of only 0.

Definition 1.11 Say $A \in \mathbf{M}(n, \mathbb{F})$ is <u>invertible</u> iff \exists a unique matrix A^{-1} s.t.

$$AA^{-1} = A^{-1}A = I \tag{15}$$

If such an A^{-1} exists, call it the <u>inverse</u> of A.

Definition 1.12 Say $A \in \mathbf{M}(n, \mathbb{R})$ is orthogonal iff A is invertible and

$$A^{-1} = A^T \tag{16}$$

ie. the inverse of A is also its transpose.

Definition 1.13 For $A \in \mathbf{M}(n, \mathbb{C})$, say A is unitary iff A is invertible and

$$A^{\dagger} = A^{-1} \tag{17}$$

ie. the inverse of A is also its Hermitian conjugate.

Definition 1.14 For $A \in \mathbf{M}(n, \mathbb{R})$, say A is <u>diagonalizable</u> iff \exists orthogonal matrix S and diagonal matrix D s.t.

$$A = SDS^T \tag{18}$$

For $A \in \mathbf{M}(n, \mathbb{C})$, call A diagonalizable if \exists unitary S and diagonal D s.t.

$$A = SDS^{\dagger} \tag{19}$$

1.3 Vectors

We will need a notion of vectors and vector notation to describe some matrix properties.

Definition 1.15 Let an $n \times 1$ matrix be called a <u>column vector</u> and a $1 \times n$ matrix be called a <u>row vector</u>.

Definition 1.16 For real n-dimensional column vectors u and v, define $u \cdot v$ as the dot product defined as

$$u \cdot v = u^T v = \sum_{i}^{n} u_i v_i \tag{20}$$

where u_i, v_i are the *i*th elements of *u* and *v* respectively.

Similarly, for complex n-dimensional u and v, define the dot product as

$$u \cdot v = u^{\dagger}v = \sum_{i}^{n} u_{i}^{*}v_{i} \tag{21}$$

Remark 1.17 Note that the above dot product is a specific case of the inner product (denoted $\langle u, v \rangle$) which, along with a vector space, defines an inner product space.

1.4 Basis

Definition 1.18 Say two vectors v and u are linearly independent iff

$$\langle v, u \rangle = 0 \tag{22}$$

Say set of vectors $\{v_1, v_2, ..., v_n\}$ is linearly independent if for all $i \neq j$, v_i and v_j are linearly independent.

Definition 1.19 For set of vectors $V = \{v_1, v_2, ..., v_k\}$ over field F, let the span of V (denoted Sp[V]) be the set of linear combinations of vectors in V, ie.

$$Sp[V] = \{c_1v_1 + c_2v_2 + \dots + c_kv_k \mid c_i \in F\}$$
(23)

Definition 1.20 For a set of vectors V, say $B = \{b_1, ..., b_k\}$ forms a <u>basis</u> of V iff

1. B is linearly independent

2. Sp[B] = V

Additionally, define the <u>dimension</u> of V as the cardinality of a basis of V.

1.5 Eigenvectors/Eigenvalues

TODO

1.6 Misc.

Sometimes, it is convenient to represent a matrix by its rows or columns:

$$A = \begin{pmatrix} --- & r_1 & --- \\ --- & r_2 & --- \\ & \vdots & \\ --- & r_n & --- \end{pmatrix} = \begin{pmatrix} | & | & | & | \\ | & | & | \\ c_1 & c_2 & \dots & c_n \\ | & | & | \\ | & | & | \end{pmatrix}$$
(24)

where $r_i = (a_{i1}, a_{i2}, ..., a_{in})$ and $c_j = (a_{1j}, a_{2j}, ..., a_{nj})$. Sometimes we use the notation $a_{i*} = (a_{i1}, a_{i2}, ..., a_{in})$ for rows and $a_{*j} = (a_{1j}, a_{2j}, ..., a_{nj})$ for columns.

2 Operator properties

2.1 Transpose

Definition given in Defn. 1.3. For the following, assume $A,B\in \mathbf{M}(n,\mathbb{F})$ and $c\in\mathbb{F}$

Property 2.1 Matrix transpose is distributive over addition and scalar multiplication

Pf.

$$(A+cB)^{T} = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix} + \begin{pmatrix} cb_{11} & \dots & cb_{1n} \\ \vdots & \ddots & \vdots \\ cb_{n1} & \dots & cb_{nn} \end{pmatrix})^{T}$$
(25)

$$= \begin{pmatrix} a_{11} + cb_{11} & \dots & a_{1n} + cb_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} + cb_{n1} & \dots & a_{nn} + cb_{nn} \end{pmatrix}^{T}$$
(26)

$$= \begin{pmatrix} a_{11} + cb_{11} & \dots & a_{n1} + cb_{n1} \\ \vdots & \ddots & \vdots \\ \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots \end{pmatrix}$$
(27)

$$\begin{array}{cccc} \left(a_{1n} + cb_{1n} & \dots & a_{nn} + cb_{nn}\right) \\ \left(a_{11} & \dots & a_{n1}\right) & \left(b_{11} & \dots & b_{n1}\right) \end{array}$$

$$= \begin{pmatrix} \vdots & \ddots & \vdots \\ a_{1n} & \dots & a_{nn} \end{pmatrix} + c \begin{pmatrix} \vdots & \ddots & \vdots \\ b_{1n} & \dots & b_{nn} \end{pmatrix}$$
(28)

$$=A^T + cB^T \tag{29}$$

Property 2.2 $(AB)^T = B^T A^T$

$$(AB)^{T} = \begin{pmatrix} \dots & a_{1} & \dots \\ \dots & a_{2} & \dots \\ \vdots & \vdots \\ \dots & a_{n} & \dots \end{pmatrix} \begin{pmatrix} | & | & | & | \\ | & | & | \\ b_{1} & b_{2} & \dots & b_{n} \\ | & | & | \\ | & | & | \end{pmatrix})^{T}$$
(30)
$$= \begin{pmatrix} a_{1} \cdot b_{1} & a_{1} \cdot b_{2} & \dots & a_{1} \cdot b_{n} \\ a_{2} \cdot b_{1} & a_{2} \cdot b_{2} & \dots & a_{2} \cdot b_{n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n} & b_{n} & a_{n} & b_{n} & \dots & a_{n} & b_{n} \end{pmatrix}^{T}$$
(31)

$$= \begin{pmatrix} a_{1} \cdot b_{1} & a_{2} \cdot b_{2} & \dots & a_{n} \cdot b_{n} \end{pmatrix}$$

$$= \begin{pmatrix} a_{1} \cdot b_{1} & a_{2} \cdot b_{1} & \dots & a_{n} \cdot b_{1} \\ a_{1} \cdot b_{2} & a_{2} \cdot b_{2} & \dots & a_{n} \cdot b_{2} \\ \vdots & \vdots & \ddots & \vdots \end{pmatrix}$$
(32)

$$\begin{pmatrix}
a_1 \cdot b_n & a_2 \cdot b_n & \dots & a_n \cdot b_n \\
\dots & b_1 & \dots \\
\dots & b_2 & \dots \\
\vdots & \vdots \\
\dots & b_n & \dots
\end{pmatrix}
\begin{pmatrix}
\begin{vmatrix} & & & & & \\ & & & & \\ & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & & \\ & & & & & \\ &$$

$$=B^T A^T \tag{34}$$

where b_i , a_i are denoting row and column vectors as appropriate (should probably make this more rigorous in fact).

2.2 Conjugate

Definition given in Defn. 1.4.

Property 2.3 Matrix conjugate is distributive over addition and scalar multiplication

Pf.

Pf.

$$\overline{A + cB} = \overline{\begin{pmatrix} a_{11} + cb_{11} & a_{12} + cb_{12} & \dots & a_{1n} + cb_{1n} \\ a_{21} + cb_{21} & a_{22} + cb_{22} & \dots & a_{2n} + cb_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} + cb_{n1} & a_{n2} + cb_{n2} & \dots & a_{nn} + cb_{nn} \end{pmatrix}}$$
(35)
$$= \begin{pmatrix} \frac{\overline{a_{11} + cb_{11}}}{a_{21} + cb_{21}} & \frac{\overline{a_{12} + cb_{12}}}{a_{22} + cb_{22}} & \dots & \overline{a_{1n} + cb_{1n}} \\ \frac{1}{a_{21} + cb_{21}} & \frac{1}{a_{22} + cb_{22}} & \dots & \overline{a_{2n} + cb_{2n}} \\ \frac{1}{a_{n1} + cb_{n1}} & \frac{\overline{a_{12} + cb_{12}}}{a_{n2} + cb_{n2}} & \dots & \overline{a_{1n} + cb_{nn}} \end{pmatrix}$$
(36)
$$= \begin{pmatrix} \overline{a_{11} + c\overline{b_{11}}} & \overline{a_{12} + c\overline{b_{12}}} & \dots & \overline{a_{1n} + c\overline{b_{1n}}} \\ \overline{a_{21} + c\overline{b_{21}}} & \overline{a_{22} + c\overline{b_{22}}} & \dots & \overline{a_{2n} + c\overline{b_{2n}}} \\ \vdots & \vdots & \ddots & \vdots \\ \overline{a_{n1} + c\overline{b_{n1}}} & \overline{a_{n2} + c\overline{b_{n2}}} & \dots & \overline{a_{nn} + c\overline{b_{nn}}} \end{pmatrix}$$
(37)
$$= \overline{A} + c^* \overline{B}$$
(38)

where in Eq. 37 we use the fact that the complex conjugate distributes over addition and multiplication.

2.3 Hermitian Conjugate

Definition given in Defn. 1.5.

Corollary 2.4 For $A, B \in \mathbf{M}(n, \mathbb{C}), c \in \mathbb{C}, (A + cB)^{\dagger} = A^{\dagger} + c^*B^{\dagger}$ Pf.

$$(A+cB)^{\dagger} = \overline{(A+cb)^T} \tag{39}$$

$$=\overline{A^T + cB^T} \tag{40}$$

$$=\overline{A^T} + \overline{c}\overline{B^T} \tag{41}$$

$$=A^{\dagger} + c^* B^{\dagger} \tag{42}$$

2.4 Commutator

Definition in Defn. 1.6

Property 2.5 For all matrices A, B,

$$[A, B] = AB - BA = -(BA - AB) = -[B, A]$$
(43)

Property 2.6 Commutator is a billinear form, ie.

$$[\lambda(A_1 + A_2), B] = \lambda[A_1, B] + \lambda[A_2, B]$$
(44)

$$[A, \lambda(B_1 + B_2)] = \lambda[A, B_1] + \lambda[A, B_2]$$
(45)

Pf.

$$[\lambda(A_1 + A_2), B] = \lambda(A_1 + A_2)B - B\lambda(A_1 + A_2)$$
(46)

$$= \lambda (A_1 B - B A_1) + \lambda (A_2 B - B A_2)$$
(47)

$$=\lambda[A_1,B] + \lambda[A_2,B] \tag{48}$$

By Prop. 2.5, this also applies to $A, \lambda(B_1 + B_2)$.

Property 2.7 For matrices A, B,

$$[A^{2}, B] = A[A, B] + [A, B]A$$
(49)

In general,

$$[A^{n}, B] = \sum_{j=0}^{n-1} A^{j} [A, B] A^{n-j-1}$$
(50)

Pf. TODO

Similarly,

Property 2.8 For matrices A, B, C s.t. [A, B] = [A, C] = 0

$$[A, BC] = 0 \tag{51}$$

Ie. if a matrix A commutes with two matrices B, C, it also commutes with the product BC.

Pf.

$$[A, BC] = ABC - BCA \tag{52}$$

$$= ABC - BAC + BAC - BCA \tag{53}$$

$$= (AB - BA)C + B(AC - CA)$$
(54)

$$= [A, B]C + B[A, C]$$
 (55)

$$=0\tag{56}$$

2.5 Trace

Definition given in Defn. 1.7.

Property 2.9 *Trace is distributive over addition and scalar multiplication* Pf.

For $A, B \in \mathbf{M}(n, \mathbb{F})$,

$$\operatorname{Tr}[A+cB] = \operatorname{Tr}\begin{bmatrix} a_{11}+cb_{11} & a_{12}+cb_{12} & \dots & a_{1n}+cb_{1n} \\ a_{21}+cb_{21} & a_{22}+cb_{22} & \dots & a_{2n}+cb_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1}+cb_{n1} & a_{n2}+cb_{n2} & \dots & a_{nn}+cb_{nn} \end{bmatrix}$$
(57)

$$= (a_{11} + cb_{11}) + (a_{22} + cb_{22}) + \dots + (a_{nn} + cb_{nn}$$
(58)

$$= (a_{11} + a_{22} + \dots + a_{nn}) + c(b_{11} + b_{22} + \dots + b_{nn})$$
(59)

$$= \operatorname{Tr}[A] + c\operatorname{Tr}[B] \tag{60}$$

Property 2.10 *Trace is invariant to transpose* Pf.

$$\operatorname{Tr}[A^T] = \sum_{i=1}^{n} [A^T]_{ii} = \sum_{i=1}^{n} [A]_{ii} = \operatorname{Tr}[A]$$
 (61)

Property 2.11 Trace is invariant to cyclic permutation. I.e. for any matrices $A, B, C \in \mathbf{M}(n, \mathbb{F})$,

$$Tr[ABC] = Tr[CAB] = Tr[BCA]$$
(62)

This is known as the $\underline{cyclic \ property}$ of the trace and is general for any number of products.

Pf.

Consider the trace of k products. The general case of k = n can be proven with just k = 2.

Base case (k = 2):

Let A, B be matrices in $\mathbf{M}(n, \mathbb{F})$ s.t. A has rows a_{i*} and columns a_{*j} . Similarly B has rows b_{i*} and columns b_{*j} . I.e.,

$$A = \begin{pmatrix} \dots & a_{1*} & \dots \\ \dots & a_{2*} & \dots \\ \vdots & \\ \dots & a_{n*} & \dots \end{pmatrix} = \begin{pmatrix} | & | & | & | \\ | & | & | & | \\ a_{*1} & a_{*2} & \dots & a_{*n} \\ | & | & | & | \\ | & | & | & | \end{pmatrix}$$
(63)

$$B = \begin{pmatrix} \dots & b_{1*} & \dots \\ \dots & b_{2*} & \dots \\ \vdots & \\ \dots & b_{n*} & \dots \end{pmatrix} = \begin{pmatrix} | & | & | & | \\ | & | & | & | \\ b_{*1} & b_{*2} & \dots & b_{*n} \\ | & | & | & | \\ | & | & | & | \end{pmatrix}$$
(64)

(65)

Then,

$$\operatorname{Tr}[AB] = \operatorname{Tr}\begin{bmatrix} \dots & a_{1*} & \dots \\ \dots & a_{2*} & \dots \\ \vdots & \vdots \\ \dots & a_{n*} & \dots \end{bmatrix} \begin{pmatrix} \begin{vmatrix} & & & & & \\ & & & & \\ b_{*1} & b_{*2} & \dots & b_{*n} \\ \begin{vmatrix} & & & & \\ & & & & \\ & & & & & \\ \end{vmatrix} \end{bmatrix}$$
(66)

$$= \operatorname{Tr} \begin{bmatrix} a_{1*} \cdot b_{*1} & a_{1*} \cdot b_{*2} & \dots & a_{1*} \cdot b_{*n} \\ a_{2*} \cdot b_{*1} & a_{2*} \cdot b_{*2} & \dots & a_{2*} \cdot b_{*n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n*} \cdot b_{*1} & a_{n*} \cdot b_{*2} & \dots & a_{n*} \cdot b_{*n} \end{bmatrix}$$
(67)

$$=\sum_{i}^{n} a_{i*} \cdot b_{*i} = \sum_{i}^{n} \sum_{j}^{n} a_{ij} b_{ji}$$
(68)

$$=\sum_{j}^{n}\sum_{i}^{n}b_{ji}a_{ij}=\sum_{j}^{n}b_{j*}\cdot a_{*j}$$
(69)

$$= \operatorname{Tr} \begin{bmatrix} b_{1*} \cdot a_{*1} & b_{1*} \cdot a_{*2} & \dots & b_{1*} \cdot a_{*n} \\ b_{2*} \cdot a_{*1} & b_{2*} \cdot a_{*2} & \dots & b_{2*} \cdot a_{*n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n*} \cdot a_{*1} & b_{n*} \cdot a_{*2} & \dots & b_{n*} \cdot a_{*n} \end{bmatrix}$$
(70)

$$= \operatorname{Tr} \begin{bmatrix} \dots & b_{1*} & \dots \\ \dots & b_{2*} & \dots \\ & \vdots & \\ \dots & b_{n*} & \dots \end{bmatrix} \begin{pmatrix} | & | & | & | \\ | & | & | & | \\ a_{*1} & a_{*2} & \dots & a_{*n} \\ | & | & | & | \\ | & | & | & | \end{pmatrix}]$$
(71)

$$= \operatorname{Tr}[BA] \tag{72}$$

Illustrative case: (k = 3)

$$Tr[ABC] = Tr[(AB)C] = Tr[C(AB)]$$

$$= Tr[CAB] = Tr[(CA)B] = Tr[B(CA)]$$

$$= Tr[BCA]$$
(73)

General case: (k = n)

$$Tr[A_{1}A_{2}...A_{n-1}A_{n}] = Tr[A_{1}(A_{2}...A_{n-1}A_{n})] = Tr[(A_{2}...A_{n-1}A_{n})A_{1}]$$
(74)

$$= Tr[A_{2}...A_{n-1}A_{n}A_{1}]$$

$$= ...$$

$$= Tr[A_{i}A_{i+1}...A_{1}A_{2}...A_{n-1}A_{n}...A_{i-2}A_{i-1}]$$

$$= Tr[(A_{i}A_{i+1}...A_{1}A_{2}...A_{n-1}A_{n}...A_{i-2})A_{i-1}]$$

$$= Tr[A_{i-1}(A_{i}A_{i+1}...A_{1}A_{2}...A_{n-1}A_{n}...A_{i-2})]$$

$$= Tr[A_{i-1}A_{i}A_{i+1}...A_{1}A_{2}...A_{n-1}A_{n}...A_{i-2}]$$

$$= ...$$

Corollary 2.12 *Trace is invariant to change of basis.* Pf.

 $\operatorname{Tr}[U^{\dagger}AU] = \operatorname{Tr}[UU^{\dagger}A] = \operatorname{Tr}[A]$ for unitary U by Prop 2.11.

3 Symmetric matrix properties

Definition 3.1 Let $\mathbf{S}(n) \subset \mathbf{M}(n,\mathbb{R})$ denote the set of symmetric real $n \times n$ matrices (as defined in Defn. 1.8).

Property 3.2 S(n) is closed under addition and scalar multiplication.

Pf.

For $A, B \in \mathbf{S}(n)$,

$$(A + cB)^{T} = A^{T} + cB^{T} = A + cB$$
(75)

Property 3.3 For symmetric A, B, their product $AB \in \mathbf{S}(n) \iff AB = BA$ Pf.

 $AB \in \mathbf{S}(n) \implies AB = BA$

$$AB = (AB)^T = B^T A^T = BA \tag{76}$$

 $AB \in \mathbf{S}(n) \iff AB = BA$

$$(AB)^T = B^T A^T = BA = AB \tag{77}$$

Corollary 3.4 For $m \in \mathbb{Z}$, if $A \in \mathbf{S}(n)$, then $A^m \in \mathbf{S}(n)$ Pf.

Follows directly from Prop. 3.3 and the fact that a matrix commutes with itself.

4 Anti-symmetric/skew-symmetric matrix properties

Definition 4.1 Let SS(n) denote the set of skew-symmetric real $n \times n$ matrices (as defined in Defn. 1.8.

For the following properties, assume $A, B \in \mathbf{SS}(n)$

Property 4.2 SS(n) closed under addition and scalar multiplication. Pf.

$$(A + cB)^{T} = A^{T} + cB^{T} = -A - cB = -(A + cB)$$
(78)

5 Orthogonal matrix properties

TODO

6 Hermitian matrices

Definition 6.1 Let $\mathbf{H}(n) \subset \mathbf{M}(n, \mathbb{C})$ denote the set of Hermitian $n \times n$ matrices (as defined in 1.9).

Property 6.2 H(n) closed under addition

Pf. For $A, B \in \mathbf{H}(n)$,

$$(A+B)^{\dagger} = A^{\dagger} + B^{\dagger} = (A+B)$$
 (79)

by 2.4

Key result for quantum mechanics:

Property 6.3 If A is Hermitian, then A has real eigenvalues

Pf.

Let λ_i be the i^{th} eigenvalue of A with eigenvector v_i . For convenience, assume the eigenvectors are normalized to 1 $(v_i \cdot v_i = v_i^{\dagger} v_i = 1)$. Then,

$$Av_i = \lambda_i v_i \tag{80}$$

$$v_i^{\dagger} A v_i = \lambda_i v_i^{\dagger} v_i \tag{81}$$

$$v_i^{\dagger} A v_i = \lambda_i \tag{82}$$

Taking the hermitian conjugate of Eq. 80,

$$v_i^{\dagger} A^{\dagger} = \lambda_i^* v_i^{\dagger} \tag{83}$$

$$v_i^{\dagger} A^{\dagger} v_i = \lambda_i^* v_i^{\dagger} v_i \tag{84}$$

$$v_i^{\mathsf{T}} A v_i = \lambda_i^* \tag{85}$$

Subtracting Eqs. 82 and 85,

$$0 = \lambda_i - \lambda_i^* \tag{86}$$

$$= \operatorname{Re}[\lambda_i] + \operatorname{Im}[\lambda_i]i - (\operatorname{Re}[\lambda_i] - \operatorname{Im}[\lambda_i]i)$$
(87)

$$=2\mathrm{Im}[\lambda_i]\tag{88}$$

Then each of λ_i are purely real, since $\text{Im}\lambda_i = 0$.

Property 6.4 If A, B are two Hermitian matrices, then the commutator [A, B] of A and B is skew-Hermitian

Pf.

$$[A,B] = AB - BA \tag{89}$$

$$= (BA)^{\dagger} - (AB)^{\dagger} \tag{90}$$

$$= (BA - AB)^{\dagger} \tag{91}$$

$$= [B, A]^{\dagger} = -[A, B]^{\dagger}$$
(92)

where we've used the trivial property of the commutator that for any matrices A, B, the following is true: [A, B] = AB - BA = -(BA - AB) = -[B, A].