Chapter 13. Symmetry Groups

Groups

Definitions:

A **group** is a set G with an operation \circ that is closed and associative, has an identity e, and every element g has an inverse g^{-1} such that $g \circ g^{-1} = e = g^{-1} \circ g \circ g$.

A group G is **Abelian** if it is commutative: $g \circ h = h \circ g$ for all g, h in G.

A **coset** of G is a set $Gh = \{gh: g \in G\}$, where $h \in G$.

A **subgroup** is a subset of G that is a group under o.

A **normal subgroup** is a subgroup H that satisfies $g \circ H = H \circ g$ for all g in G, or equivalently $H = g^{-1} \circ H \circ g$.

A group is **simple** if it contains no non-trivial normal subgroup.

The simple groups are the fundamental "building blocks" of more complex groups.

Theorem. There are precisely 4 **classical** and 5 **exceptional** simple Lie groups.

- Classical Families: A_m , B_m , C_m , D_m having dimensions m(m+2), m(2m+1), m(2m+1), and m(2m-1), respectively where $m \in \mathbb{Z}^+$.
- Exceptional Groups: E₆, E₇, E₈, F₄, G₂ of dimension 78, 133, 248, 52, and 14 respectively

Theorem. The simple finite groups have been classified into classical and exceptional groups. The largest exceptional group has $\approx 10^{60}$ elements and is known as **the monster**.

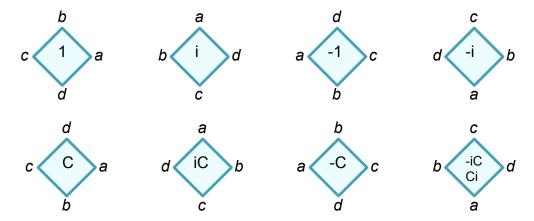
Definition. The **Product Group** of groups G and H is $\mathbf{GxH} = \{(g,h): g \in G, h \in H\}$ with group operation $(g_1,h_1) \circ (g_2,h_2) = (g_1 \circ g_2,h_1 \circ h_2)$.

Definition. Let N be a subgroup of G. The **Factor Space G/N** is the collection of cosets $G \circ n$ along with the operation $(G \circ n_1) \circ (G \circ n_2) = G \circ (n_1 \circ n_2)$.

Theorem. If N is normal then G/N is a group, called the **Factor Group**.

Theorem. (Problem [13.10]) $H \cong G \times H/G$.

Symmetries of a Square



Definitions:

Non-reflecting Group: $\langle i \rangle = \{1, i, -1, -i\}$

Reflecting Group: < i, $C > = \{1, i, -1, -i, C, iC, -C, -iC = Ci\}$

C is complex conjugation: $a + bi \mapsto a - bi$. 1 is the null rotation, which is the group identity element. i is the 90° counter-clockwise rotation of the square

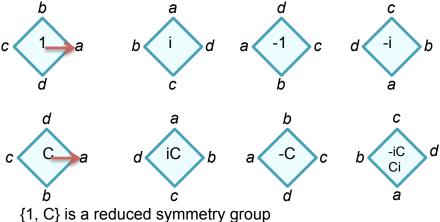
Convention: ab means b acts first.

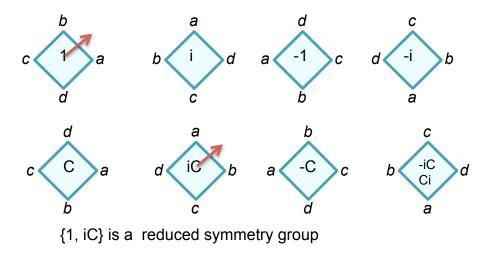
A subgroup of a symmetry group is called a **reduced symmetry group**.

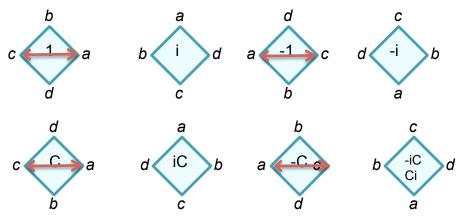
Examples:

Normal subgroups of < i, C > :{1, -1, C, -C}, {1, -1}, {1, -1} Non-normal subgroups of < i, C > :{1, -C}, {1, iC}, {1, C} For example, $\{1, C\}$ $i = \{i, Ci\} \neq \{i, -Ci\} = i \{1, C\}$

Example [13.6]: Reduced symmetry groups can be generated using one or more arrows.







{1, -1, C, -C} is a reduced symmetry group

Symmetries of a Sphere

Definitions:

A group G whose underlying set is continuous is called a **Lie Group**. **SO(3)** is the group of non-reflective symmetries of a 3-sphere **O(3)** is the **Orthogonal Group**. It consists of both the reflective and non-reflective symmetries of a sphere.

 $O(3) = SO(3) \cup T$, the disjoint union of O(3) with the coset of reflective symmetries

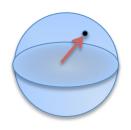
T = R SO(3) = $\{Rg : g \in SO(3)\}$ where **R** is the reflection operator on the sphere.

Recall problem [12.7]: SO(3) is group isomorphic to the solid sphere \mathcal{R} of radius π with antipodal points identified.

Theorem. (Problem [13.7]) SO(3) and $\{1, R\}$ are the only normal subgroups of O(3), where **1** is the null rotation. (Penrose overlooked that the latter group is normal.)

Examples. Reduced Symmetry Groups

The set of rotations that fix a point on the sphere forms a non-normal subgroup. It is the set of rotations having the arrow as its axis.





Marking the sphere with vertices of a regular polyhedra reduces to the finite group of rotations of the sphere that take each vertex to one of the others. Such reduced symmetry groups are non-normal.

Linear Transformations and Matrices

Definition. Let V and W be vector spaces.

- f: V→ W is a homomorphism if it preserves the vector space structure:
 f(au + bv) = af(u) + bf(v) for all vectors u and v and scalars a and b.
 - I am (/ M) is the set of homemorphisms from \/ to \//
- Hom(V,W) is the set of homomorphisms from V to W.
- A(V) = Hom(V,V).
- A linear transformation is a member $T \in A(V)$.
 - That is, a linear transformation is a function T: $V \rightarrow V$ such that T(au + bv) = aTu + bTv.

Theorem. [13.12, 13.13] Let $V = \mathbb{R}^3$, using (x^1, x^2, x^2) instead of (x, y, z). Then a linear transformation T takes the form T: $x^r \mapsto T^r s x^s = ax^1 + bx^2 + cx^3$.

Note. Linear transformations are represented by matrices:

$$T: \begin{pmatrix} x^{1} \\ x^{2} \\ x^{3} \end{pmatrix} \mapsto \begin{pmatrix} T_{1}^{1} & T_{2}^{1} & T_{3}^{1} \\ T_{1}^{2} & T_{2}^{2} & T_{3}^{2} \\ T_{1}^{3} & T_{2}^{3} & T_{3}^{3} \end{pmatrix} \begin{pmatrix} x^{1} \\ x^{2} \\ x^{3} \end{pmatrix} \quad \text{or} \quad x \mapsto Tx$$

In diagrammatic form this is \mapsto



Theorem. If R = ST then $R_c^a = S_b^a T_c^b$. That is, the composition, R, of 2 linear transformations is the result of matrix multiplication of S and T. In diagrammatic notation:

Example. TI = T = IT is written in diagrammatic form as

and, in
$$\mathbb{R}^3$$
, $I = \delta_b^a = \begin{cases} 1 & \text{if } a = b \\ 0 & \text{if } a \neq b \end{cases} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ where a , b range over $\{1, 2, 3\}$.

Definitions. A linear transformation T is **singular** if Dim(TV) < Dim W; that is, T is not *onto*.

Theorem. [13.17] T is singular iff $\exists v \neq 0$ such that $\forall v \neq 0$.

Corollary. [Bud] T is 1-1 iff T is non-singular iff T is onto.

Proof: T is 1-1
$$\Leftrightarrow \forall v \neq w \ \mathsf{T}(v-w) = \mathsf{T}(v) - \mathsf{T}(w) \neq 0 \ \Leftrightarrow \ \forall u \neq 0 \ \mathsf{T}(u) \neq 0$$

$$\Leftrightarrow \ \mathsf{T} \text{ is non-singular } \Leftrightarrow \ \mathsf{T} \text{ is onto.}$$

$$(*) \text{ Set } v = 3u \text{ and } w = 2u.$$

Theorem. [13.18] If T is nonsingular, then it has an inverse T⁻¹.

Theorem. [13.19]
$$T^{-1} = \binom{1}{n} = \binom{n}{n}$$

Definition. The **transpose** of the matrix $T = (T_j^i)$ is the matrix $T^T = (T_j^i)$.

Definition. A matrix T is **orthogonal** if $T^{-1} = T^{T}$.

Determinants and Traces

Definition. Det
$$T = \frac{1}{n!}$$

$$= \frac{1}{n!} \in {}^{ab\cdots d} T^e_a T^f_b \cdots T^h_d \varepsilon_{ef\cdots h}.$$

Proof: Let $P_{a...g}$ be the set of permutations of (a, ..., g). Then

$$= n! \varepsilon_{a \cdots g} \in {}^{r \cdots x} T^{[a}_{r} \cdots T^{g]}_{x}$$

$$=\frac{n!}{n!}\varepsilon_{\mathbf{a}\cdots\mathbf{g}}\in^{r\cdots\mathbf{x}}\sum_{\boldsymbol{\pi}\in\mathcal{P}_{\mathbf{a}\mathbf{b}\cdots\mathbf{g}}}\mathrm{Sign}(\boldsymbol{\pi})\mathsf{T}^{\boldsymbol{\pi}(\mathbf{a})}{}_{r}\cdots\mathsf{T}^{\boldsymbol{\pi}(\mathbf{g})}{}_{\mathbf{x}}=\underbrace{\quad n!}_{\mathbf{z}}\varepsilon_{\mathbf{a}\cdots\mathbf{g}}\;\mathsf{T}^{\mathbf{a}}{}_{r}\cdots\mathsf{T}^{\mathbf{g}}{}_{\mathbf{x}}\in^{r\cdots\mathbf{x}}$$

(*) π is the composition of transmutations (i.e., of pairwise permutations).

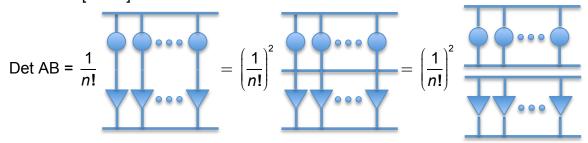
Let π^* : $\begin{array}{c} c \mapsto e \\ e \mapsto c \end{array}$ be a transmutation. Then

$$\begin{split} &\varepsilon_{a\cdots c\cdots e\cdots g}\in {}^{r\cdots t\cdots v\cdots x}\operatorname{Sign}\left(\pi^{\,*}\right)\mathsf{T}^{\pi(a)}_{r}\cdots\mathsf{T}^{\pi(c)}_{t}\cdots\mathsf{T}^{\pi(e)}_{v}\cdots\mathsf{T}^{\pi(g)}_{x}\\ &=\varepsilon_{a\cdots c\cdots e\cdots g}\in {}^{r\cdots t\cdots v\cdots x}\operatorname{Sign}\left(\pi^{\,*}\right)\mathsf{T}^{a}_{r}\cdots\mathsf{T}^{e}_{t}\cdots\mathsf{T}^{c}_{v}\cdots\mathsf{T}^{g}_{x}\\ &=\varepsilon_{a\cdots e\cdots c\cdots g}\in {}^{r\cdots t\cdots v\cdots x}\operatorname{Sign}\left(\pi^{\,*}\right)\mathsf{T}^{a}_{r}\cdots\mathsf{T}^{c}_{t}\cdots\mathsf{T}^{e}_{v}\cdots\mathsf{T}^{g}_{x} \ \ (\operatorname{Rename}\ c\mapsto e\ \&\ e\mapsto c)\\ &=\operatorname{Sign}\left(\pi^{\,*}\right)\varepsilon_{a\cdots c\cdots e\cdots g}\in {}^{r\cdots t\cdots v\cdots x}\operatorname{Sign}\left(\pi^{\,*}\right)\mathsf{T}^{a}_{r}\cdots\mathsf{T}^{c}_{t}\cdots\mathsf{T}^{c}_{v}\cdots\mathsf{T}^{g}_{x}\\ &=\varepsilon_{a\cdots g}\,\mathsf{T}^{a}_{r}\cdots\mathsf{T}^{g}_{x}\in {}^{r\cdots x}\,. \end{split}$$

So, for any permutation π , we have

$$\varepsilon_{\mathbf{a}\cdots\mathbf{g}}\!\in^{r\cdots\mathbf{x}}\mathrm{Sign}\!\left(\pi\right)\!\mathsf{T}^{\pi(\mathbf{a})}_{r}\cdots\mathsf{T}^{\pi(\mathbf{g})}_{\mathbf{x}}=\varepsilon_{\mathbf{a}\cdots\mathbf{g}}\,\mathsf{T}^{\mathbf{a}}_{r}\cdots\mathsf{T}^{\mathbf{g}}_{\mathbf{x}}\!\in^{r\cdots\mathbf{x}}$$

Theorem. [13.22]



= DetA DetB

Theorem. (p.260 – no proof given) Matrix A is singular iff Det A = 0.

Proof: From [13.19], A is non-singular iff Det $A \neq 0$.

Definition. Vectors v and w are **orthogonal** if $v \cdot w = 0$. That is, the angle between them is 90° .

Theorem. A matrix is orthogonal (i.e., $T^T = T^{-1}$) iff its column vectors are mutually orthogonal.

Example. Orthogonal 2 x 2 Matrices: A and B

Let
$$A = \begin{pmatrix} v & w \end{pmatrix}$$

$$= \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}.$$

$$A^{T} = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}.$$

$$A^{T} = A^{-1}:$$

$$AA^{T} = \begin{pmatrix} \sin^{2}\theta + \cos^{2}\theta & 0 \\ 0 & \sin^{2}\theta + \cos^{2}\theta \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$$
Similarly $A^{T}A = I$

So A is an orthogonal matrix

Det A = Det A^T =
$$\cos^2 \theta + \sin^2 \theta = 1$$

The column vectors of A are orthogonal: $v \perp w$

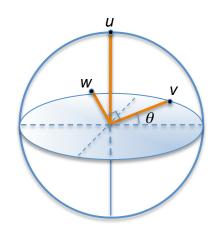
Let
$$B = \begin{pmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{pmatrix} = B^T$$
. Then $B B^T = I$, Det $B = Det B^T = -1$, and its

column vectors v and w_1 are orthogonal.

Examples. Orthogonal 3 x 3 Matrices: A, B, and C

Let
$$v = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix}$$
, $w = \begin{pmatrix} -\sin \theta \\ \cos \theta \\ 0 \end{pmatrix}$, and $u = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$.

Let
$$A = \begin{pmatrix} v & w & u \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
.



$$\mathbf{A}^{\mathsf{T}} = \left(\begin{array}{ccc} \mathsf{Cos} \; \theta & \mathsf{Sin} \; \theta & \mathsf{0} \\ - \; \mathsf{Sin} \; \theta & \mathsf{Cos} \; \theta & \mathsf{0} \\ \mathsf{0} & \mathsf{0} & \mathsf{1} \end{array} \right). \; \mathsf{A} \; \mathsf{is} \; \mathsf{orthogonal}, \; \mathsf{its} \; \mathsf{columns} \; \mathsf{are} \; \mathsf{orthogonal} \; \mathsf{vectors},$$

and its determinant is +1.

Let B =
$$\begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & -1 \end{pmatrix}$$
. B is orthogonal and its determinant is -1. \checkmark

Let C be a θ -rotation of A about an axis $\{t(a,b,c): 0 < t < \infty, a^2 + b^2 + c^2 = 1\}$:

$$\mathbf{C} = \left(\begin{array}{ccc} \frac{1}{2} \Big[1 + a^2 - b^2 - c^2 + \Big(1 - a^2 + b^2 + c^2 \Big) \mathrm{Cos} \ \theta \Big] & 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[- c \, \mathrm{Cos} \ \frac{\theta}{2} + ab \, \mathrm{Sin} \ \frac{\theta}{2} \Big] & 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[b \, \mathrm{Cos} \ \frac{\theta}{2} + ac \, \mathrm{Sin} \ \frac{\theta}{2} \Big] \\ 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[c \, \mathrm{Cos} \ \frac{\theta}{2} + ab \, \mathrm{Sin} \ \frac{\theta}{2} \Big] & \frac{1}{2} \Big[1 - a^2 + b^2 - c^2 + \Big(1 + a^2 - b^2 + c^2 \Big) \mathrm{Cos} \ \theta \Big] & 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[- a \, \mathrm{Cos} \ \frac{\theta}{2} + bc \, \mathrm{Sin} \ \frac{\theta}{2} \Big] \\ 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[- b \, \mathrm{Cos} \ \frac{\theta}{2} + ac \, \mathrm{Sin} \ \frac{\theta}{2} \Big] & 2 \mathrm{Sin} \ \frac{\theta}{2} \Big[a \, \mathrm{Cos} \ \frac{\theta}{2} + bc \, \mathrm{Sin} \ \frac{\theta}{2} \Big] & \frac{1}{2} \Big[1 - a^2 - b^2 + c^2 + \Big(1 + a^2 + b^2 - c^2 \Big) \mathrm{Cos} \ \theta \Big] \end{array} \right]$$

It can be directly verified that C is an orthogonal matrix with mutually orthogonal column vectors and determinant +1. ✓

Definition. A **symmetry** of a vector space (V,+) is a transformation $T: V \mapsto V$ that is 1-1 and onto that preserves the vector space structure:

$$T(a v + b w) = a Tv + b Tw$$

Definition. The **General Linear Group GL**(*n*) is the group of symmetries of an *n*-dimensional vector space.

Theorem. GL(n) is the group of non-singular $(n \times n)$ matrices.

Proof. Let $T \in GL(n)$. Since T(a v + b w) = a Tv + b Tw, T is a linear transformation. Were T singular, then by [13.17] Dim $TV < n \Rightarrow T$ is not onto, a contradiction. Therefore T is a non-singular linear transformation. Thus in any basis, T is represented by a non-singular matrix.

Definition. The **Special Linear Group SL**(n) is the subset of GL(n) having determinant = 1.

Theorem. SL(n) is a normal subgroup of GL(n).

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Proof. First, SL(n) is a group:
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Closed: If S_1, S_2 \in SL(n), then Det(S_1S_2) = Det(S_1) Det(S_2) = 1

\Rightarrow S_1 S_2 \in SL(n).

Identity: Det(I) = 1 \Rightarrow I \in SL(n)

Inverse: 1 = Det(I) = Det(S_1S_1^{-1}) = Det(S_1) Det(S_1^{-1}) = Det(S_1^{-1})

\Rightarrow S_1^{-1} \in SL(n)
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Also, SL(n) is normal:

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Let S \in SL(n) and G \in GL(n). Then Det(G^{-1} S G) = Det(G^{-1}) Det(S) Det(G) = Det(G^{-1}) Det(G)
= Det(G G^{-1}) = Det(I) = 1
\Rightarrow G^{-1} S G \in SL(n) \Rightarrow G^{-1} SL(n) G = SL(n)
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The groundwork has now been laid to introduce the table, below, that shows the relationships between GL(3), O(3), SL(3), general linear transformations, orthogonality, determinants, and symmetries. The table shows that $SL(3) \subseteq O(3) \subseteq GL(3) \subseteq A(\mathbb{R}^3)$, and GL(3) is both the set of symmetries of \mathbb{R}^3 and the set of non-singular matrices. It also shows that the orthogonal group O(3) is a proper subset of the set of orthogonal matrices (shaded blue).

In general, non-singular matrices squeeze and stretch the unit sphere (or the reflected sphere) into an ellipsoid. However, singular matrices are more severe. They squash the unit sphere down to a 2-dimensional circle or ellipse or even to a line or a point.

Only orthogonal matrices preserve the sphere without squeezing or stretching any portion of it. This is achieved by limiting its operation to rotations and reflections. However, if determinant $\neq \pm 1$ then orthogonal matrices also expand or contract the sphere.

Non-orthogonal matrices also squeeze, stretch, or preserve the sphere but not as rotations. Rather, the matrix columns would contain non-orthogonal vectors. In such a case the angle between the 1st and 2nd column vectors might be less than

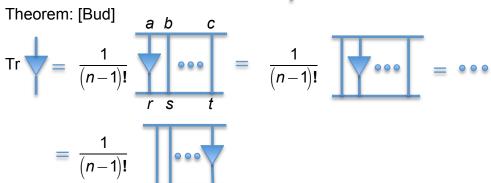
 $A(\mathbb{R}^3) = 3 \times 3$ Matrices

Determinant	Orthogonal	Sphere maps to a	Matrix Type		
0	Yes	Circle or line or point	Singular		
	No	Ellipse or line or point	Sirigulai		
Between -1 and 0	Yes	Contracted reflected sphere			
	No	Contracted reflected ellipsoid			
Between 0 and +1	Yes	GL(3)			
	No				
-1	Yes	Yes Reflected sphere			
	No	No Reflected ellipsoid O(3)			
+1	Yes	SL(3) = sphere	singular		
	No	Ellipsoid	Symmetries		
< -1	Yes	Expanded reflected sphere	of \mathbb{R}^3		
	No	Expanded reflected ellipsoid			
> 1	Yes	Expanded sphere			
	No	Expanded ellipsoid			

 90° , squeezing the sphere along associated plane. The angle between the 2^{nd} and 3^{rd} vectors would then be greater than 90° , stretching the sphere along that plane.

Matrices with positive determinant act on the sphere. Matrices with negative determinant behave exactly the same but act on the reflected sphere.

Definition. The **Trace** of A is
$$T_r(A) = Tr = T_k^k = T_1^1 + \cdots + T_n^n$$
.



Proof: Let $\mathcal{Q}_{ab...c}$ and $\mathcal{Q}_{rs...t}$ be the sets of permutations of (a,b,...,c) and (r,s,...,t),

respectively. Let B =
$$\frac{a \ b \ c}{r \ s \ t}$$

$$=\in^{r\mathbf{s}\cdots t}\varepsilon_{\mathbf{a}\mathbf{b}\cdots\mathbf{c}}\mathsf{T}^{\mathbf{a}}{}_{\mathbf{r}}\delta^{\mathbf{b}}_{\mathbf{s}}\cdots\delta^{\mathbf{c}}_{t}=\sum_{\boldsymbol{\pi}\in\mathcal{Q}_{\mathbf{a}\mathbf{b}\cdots\mathbf{c}}}\sum_{\boldsymbol{\pi}'\in\mathcal{Q}_{\mathbf{s}\cdots t}}\in^{\boldsymbol{\pi}'(r)\boldsymbol{\pi}'(\mathbf{s})\cdots\boldsymbol{\pi}'(t)}\varepsilon_{\boldsymbol{\pi}(\mathbf{a})\boldsymbol{\pi}(\mathbf{b})\cdots\boldsymbol{\pi}(\mathbf{c})}\mathsf{T}^{\boldsymbol{\pi}(\mathbf{a})}{}_{\boldsymbol{\pi}'(\mathbf{c})}\delta^{\boldsymbol{\pi}(\mathbf{b})}_{\boldsymbol{\pi}'(\mathbf{s})}\cdots\delta^{\boldsymbol{\pi}(\mathbf{c})}_{\boldsymbol{\pi}'(t)}.$$

Fix π . The only non-zero term in the sum is

$$\in^{\pi(\mathbf{a})\pi(\mathbf{b})\cdots\pi(\mathbf{c})}\varepsilon_{\pi(\mathbf{a})\pi(\mathbf{b})\cdots\pi(\mathbf{c})}\mathsf{T}^{\pi(\mathbf{a})}_{\pi(\mathbf{a})}\delta_{\pi(\mathbf{b})}^{\pi(\mathbf{b})}\cdots\delta_{\pi(\mathbf{c})}^{\pi(\mathbf{c})}=\mathsf{T}^{\pi(\mathbf{a})}_{\pi(\mathbf{a})}.$$

I showed in Problem [13.22] that $\in^{xy\cdots z} \varepsilon_{xy\cdots z} = 1$ for any fixed (x,y,\ldots,z) .

Thus, B = $\sum_{\pi \in \mathscr{Q}_{ab\cdots c}} \mathsf{T}^{\pi(a)}_{\pi(a)}$. This sum has n! terms composed of (n-1)! terms equal to T^a_a , (n-1)! terms equal to T^b_b , ..., and (n-1)! terms equal to T^c_c . So,

B =
$$(n-1)!$$
 ($T_a^a + T_b^b + ... + T_c^c$) = $(n-1)!$ Tr (A) = $(n-1)!$ Tr Similarly for the other figures.

Theorem. [13.24] $\text{Det}\big(I+\in A\big)=1+\in T_r\big(A\big)$ if we ignore 2^{nd} order and higher \in terms.

Theorem. [13.25] Det $e^{A} = e^{T_{r}(A)}$.

Definition. An **Eigenvector** is a non-zero vector v for which $\exists \lambda \in \mathbb{C}$ such that $Tv = \lambda v \Leftrightarrow (T - \lambda I)v = 0$. λ is called an **Eigenvalue**.

Note: $Det(T - \lambda I) = 0$ and so $(T - \lambda I)$ is singular

Theorem. [13.26] $\operatorname{Det}(T-\lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda) = 0$ is a polynomial equation of degree n.

Definition. λ has multiplicity r means that λ appears r times in the equation above. Eigenvalue multiplicities are called **degeneracies** in Quantum Mechanics.

Definition. The set of Eigenvectors corresponding to λ is a linear space called an **Eigenspace**.

Theorem. If *d* is the dimension of the Eigenspace of λ and *r* is the multiplicity of λ then $1 \le d \le r$.

Theorem. [13.27] Let $\{\lambda_i\}$ be the set of Eigenvalues of an $n \times n$ matrix T, and let r_i be the multiplicity of λ_i . Then $\sum r_i = n$.

Corollary. A linear transformation T has at least 1 Eigenvector.

Theorem. [13.30] Suppose $\{e_k\}$ and $\{f_k\}$ are bases for a vector space V, and $f_k = Te_k$. Then

$$f_j = \left(\begin{array}{c} \mathsf{T}^1_{\ j} \\ \vdots \\ \mathsf{T}^n_{\ j} \end{array}\right).$$

That is, the components of f_j in basis $\{e_k\}$ are $\left(\mathsf{T}^1_{\ j},\cdots,\mathsf{T}^n_{\ j}\right)$.

Theorem. [13.31] If the Eigenspace dimension of every multiple Eigenvector equals its multiplicity, then there is a basis for V composed of Eigenvectors, and the matrix of T in this basis is

$$T = \left(\begin{array}{ccc} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{array} \right).$$

The next theorem states that even when the hypothesis of the above theorem is not satisfied, the matrix of T can at least be written in upper triangular form.

Theorem. (Note 13.12): **Jordan Canonical Form:** Let $\left\{\lambda_i\right\}$ be the set of

Eigenvalues of an $n \times n$ matrix T, and let r_i be the multiplicity of λ_i . Then there is a basis for V such that the matrix of T in this basis is

	λ_1	1											0	
		$\lambda_{_{\! 1}}$	1										:	
			٠.	•••	1									
					$\lambda_{\rm 1}$	0								
						λ_{2}	1							
T =							٠.	·						
								٠٠.	1	0				
									λ_{n-1}		1			
										λ_n		٠.		
	:	.•									λ_n	٠.	1	
	Ò	•										•	1	
	U	•••											λ_n	