# Linear Algebra

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## Axioms for Number Systems

Let S be a number system  $(S, +, \times)$ <u>Axioms for Addition</u>: A1:  $\forall \alpha, \beta \in S \quad \alpha + \beta = \beta + \alpha$ A2:  $\forall \alpha, \beta, \gamma \in S \quad (\alpha + \beta) + \gamma = \alpha + (\beta + \gamma)$ A3:  $\exists 0 \in S \ s.t. \ \forall \alpha \in S \quad 0 + \alpha = \alpha + 0 = \alpha$ A4:  $\forall \alpha \in S \ \exists (-\alpha) \in S \ s.t. \ \alpha + (-\alpha) = (-\alpha) + \alpha = 0$ <u>A4:</u>  $\forall \alpha, \beta \in S \ \exists (-\alpha) \in S \ s.t. \ \alpha + (-\alpha) = (-\alpha) + \alpha = 0$ <u>Axioms for Multiplication</u>: M1:  $\forall \alpha, \beta \in S \quad \alpha\beta = \beta\alpha$ M2:  $\forall \alpha, \beta, \gamma \in S \quad (\alpha\beta)\gamma = \alpha(\beta\gamma)$ M3:  $\exists 1 \in S \ s.t. \ \forall \alpha \in S \quad 1\alpha = \alpha 1 = \alpha$ 

M4:  $\forall \alpha \in S^* \exists \alpha^{-1} \in S \ s.t. \ \alpha \alpha^{-1} = \alpha^{-1} \alpha = 1$ 

#### Distributivity:

D1:  $\forall \alpha, \beta, \gamma \in S \quad (\alpha + \beta)\gamma = \alpha\gamma + \beta\gamma$ 

**Def**<sup><u>n</u></sup>. A set S with addition & multiplication satisfying A1-A4, M1-M4 & D1 is a <u>field</u> if  $1 \neq 0$ .

# **Vector Spaces**

**Def**<sup><u>n</u></sup>. A vector space over a field K is a set V with addition & scalar multiplication, so  $\forall v, w \in V \exists v + w \in V \& \forall \alpha \in K \forall v \in V \exists \alpha v \in V$ 

i) Addition Satisfies A1-A4 ii)  $\alpha(\boldsymbol{v} + \boldsymbol{w}) = \alpha \boldsymbol{v} + \alpha \boldsymbol{w}$ iii)  $(\alpha + \beta)\boldsymbol{v} = \alpha \boldsymbol{v} + \beta \boldsymbol{v}$ iv)  $(\alpha\beta)\boldsymbol{v} = \alpha(\beta\boldsymbol{v})$ v)  $1\boldsymbol{v} = \boldsymbol{v}1 \ \forall \boldsymbol{v} \in V$ iv)  $1\boldsymbol{v} = \boldsymbol{v}1 \ \forall \boldsymbol{v} \in V$ iv) If  $\alpha \boldsymbol{v} = \boldsymbol{0}$  then  $\alpha = 0$  or  $\boldsymbol{v} = \boldsymbol{0}$ 

#### Linear Independence, Spanning & Bases of Vector Spaces

**Def**<sup><u>n</u></sup>. Let V be a vector space over  $K, v_1, \ldots, v_n \in V$  the vectors are linearly independent if  $\exists \alpha_1, \ldots, \alpha_n \in K$  not all 0 s.t.  $\alpha_1 v_1 + \cdots + \alpha_n v_n = 0$ , otherwise linearly dependent.

**Lemma.** If  $v_1, \ldots, v_n$  are linearly dependent, either  $v_1 = 0$  or  $v_r$  is a linear combination of  $v_1, \ldots, v_{r-1}$ .  $P_r^{f}$ . Trivial

**Def**<sup><u>n</u></sup>.  $v_1, \ldots, v_n$  span V if  $\forall v \in V \exists \alpha_1, \ldots, \alpha_n$  s.t.  $\alpha_1 v_1 + \cdots + \alpha_n v_n = v$ 

**Def**<sup><u>n</u></sup>. If  $v_1, \ldots, v_n$  span V & are linearly independent they form a <u>basis</u> of V.

**Def**<sup><u>n</u></sup>. The unique scalars that determine any given  $v \in V$  are called the <u>coordinates</u> of v.

**Th**<sup>m</sup>. <u>The Basis Theorem</u>: Suppose  $v_1, \ldots, v_m \notin w_1, \ldots, w_n$  are both bases of the vector space V, then m = n.

**Def**<sup><u>n</u></sup>. The number of vectors n in a basis of the finite dimensional vector space V is called the <u>dimension</u>: dim(V) = n.

Sifting: Given  $v_1, \ldots, v_r \in V$  successively look at  $v_1, \ldots, v_r$  keep  $v_i$  unless  $v_i = 0$  or  $v_i$  is a linear combination of  $v_1, \ldots, v_{i-1}$ .

**Lemma.** If  $v_1, \ldots, v_n$  we span  $V \in w$  is a linear combination of  $v_1, \ldots, v_n$  then  $v_1, \ldots, v_n$  span V.

 $P_{-}^{f}$ .  $\boldsymbol{w} = \alpha_1 \boldsymbol{v}_1 + \dots + \alpha_n \boldsymbol{v}_n$  now substitute  $\boldsymbol{w}$ 

**Th**<sup> $\underline{\mathbf{m}}$ </sup>. Suppose  $v_1, \ldots, v_n$  span V then  $\exists$  subsequence of vectors, a basis of V.

 $P_{-}^{f}$ . Sift  $\boldsymbol{v}_{1},\ldots,\boldsymbol{v}_{n}$ 

**Th**<sup>m</sup>. Suppose  $v_1, \ldots, v_n$  are linearly independent in V. We can extend this to a basis of V.

 $P^{\underline{f}}$ . Add  $\boldsymbol{w}_1, \ldots, \boldsymbol{w}_m$  & sift out  $\boldsymbol{w}$ 's.

**Prop**<sup>**n**</sup>. <u>The Exchange Lemma</u>: Suppose  $v_1, \ldots, v_n$  span  $V \notin w_1, \ldots, w_m$  are linearly independent in V then  $m \leq n$ .

 $P_{-}^{f}$ . Place  $w_1$  infront of  $v_1, \ldots, v_n$  & sift removing at less one vector. now repeat for  $w_i$  removing at less one vector each time. Hence  $m \leq n$ . QED

**Cor**<sup><u>ly</u></sup>. If n vectors form a basis of V then n - 1 vectors cannot span V  $\mathcal{E}$  n + 1 vectors cannot be independent.

 $P^{\underline{f}}$ . Of <u>The Basis Theorem</u>: Since  $v_i$ 's span  $V \& w_j$ 's are linearly independent  $n \leq m$  by Exchange Lemma. Since  $w_j$ 's span  $V \& v_i$ 's are linearly independent  $m \leq n$  by Exchange Lemma. Hence n = m. QED

#### Subspaces

**Def**<sup><u>n</u></sup>. A subspace of V is a non-empty subset  $W \subset V$  s.t. W is closed under addition & scalar multiplication. i.e.  $u, v \in W \ \alpha, \beta \in K \implies \alpha u + \beta v \in W$ 

**Prop**<sup><u>n</u></sup>. If  $W_1 \notin W_2$  are subspaces of V then so is  $W_1 \cap W_2$ 

 $P^{\underline{f}}$ . Trivial

*Note.*  $W_1 \cup W_2$  not necessarily a subspace.

**Def**<sup><u>n</u></sup>. Let  $W_1 \& W_2$  be subspaces of V then  $W_1 + W_2$  is defined to be  $\boldsymbol{v} \in V$  s.t.  $\boldsymbol{v} = \boldsymbol{w}_1 + \boldsymbol{w}_2$  for some  $\boldsymbol{w}_1 \in W_1 \& \boldsymbol{w}_2 \in W_2$  or  $W_1 + W_2 := \{\boldsymbol{w}_1 + \boldsymbol{w}_2 : \boldsymbol{w}_1 \in W_1 \& \boldsymbol{w}_2 \in W_2\}$ 

**Prop**<sup><u>n</u></sup>.  $W_1 + W_2$  is the smallest subspace to contain  $W_1 \notin W_2$ .

 $P_{-}^{f}$ . Any subspace of V containing  $W_1 \& W_2$  must contain  $W_1 + W_2$  QED

QED

QED

QED

At this point we drop bold face notation for vectors

**Th<sup>m</sup>**. If V is a finite dimensional vector space  $\mathcal{B}$  W<sub>1</sub>, W<sub>2</sub> subspaces of V then:

 $\dim(W_1 + W_2) = \dim(W_1) + \dim(W_2) - \dim(W_1 \cap W_2)$ 

$$\begin{split} P^{\underline{f}}. \mbox{ Let } \dim(W_1+W_2) &= r \ \& \ e_1, \dots, e_r \ be \ a \ basis \ of \ W_1 \cap W_2. \ Extend \ this \ to \ e_1, \dots, e_r, f_1, \dots, f_s \ to \ be \ a \ basis \ of \ W_1 \ s.t. \ \dim(W_1) &= r+s. \ Also \ extend \ to \ e_1, \dots, e_r, g_1, \dots, g_t \ to \ be \ a \ basis \ of \ W_2 \ s.t. \ \dim(W_2) &= r+t \\ \forall w_1 \in W_1, \ w_1 &= \alpha_1 e_1 + \dots + \alpha_r e_r + \beta_1 f_1 + \dots + \beta_s f_s \\ \forall w_2 \in W_2, \ w_2 &= \gamma_1 e_1 + \dots + \gamma_r e_r + \delta_1 g_1 + \dots + \delta_t g_t \\ hence \ w_1 + w_2 &= (\alpha_1 + \gamma_1) e_1 + \dots + (\alpha_r + \gamma_r) e_r + \beta_1 f_1 + \dots + \beta_s f_s + \delta_1 g_1 + \dots + \delta_t g_t \in W_1 + W_2 \ so \ e_i, f_j, g_k \ span \ W_1 + W_2 \\ Suppose: \ \alpha_1 e_1 + \dots + \alpha_r e_r + \beta_1 f_1 + \dots + \beta_s f_s + \gamma_1 g_1 + \dots + \gamma_t g_t = 0 \\ then \ \alpha_1 e_1 + \dots + \alpha_r e_r + \beta_1 f_1 + \dots + \beta_s f_s = -\gamma_1 g_1 - \dots - \gamma_t g_t, \ s.t. \ LHS \in W_1, \ RHS \in W_2 \\ \Longrightarrow \ both \ \in W_1 \cap W_2 \ with \ basis \ e_i. \\ Now \ -\gamma_1 g_1 - \dots - \gamma_t g_t = \delta_1 e_1 + \dots + \delta_r e_r \ i.e: \ \delta_1 e_1 + \dots + \delta_r e_r + \gamma_1 g_1 + \dots + \gamma_t g_t = 0 \\ e_i, g_k \ basis \ of \ W_2 \ \Longrightarrow \ all \ \delta^c s \ \& \ \gamma^c s = 0 \ leaving \ \alpha_1 e_1 + \dots + \alpha_r e_r + \beta_1 f_1 + \dots + \beta_s f_s = 0 \\ e_i, f_j \ basis \ of \ W_1 \ \Longrightarrow \ all \ \alpha^c s \ \& \ \beta^c s = 0 \ so \ e_i, f_j, g_k \ are \ linearly \ independent, \ hence \ e_i, f_j, g_k \ are \ a \ basis \ of \ W_1 + W_2, \ hence \end{split}$$

$$\dim(W_1 + W_2) = \dim(W_1) + \dim(W_2) - \dim(W_1 \cap W_2)$$

QED

QED

**Prop**<sup>**n**</sup>.  $v_1, \ldots, v_n \in V$  all linear combinations form a subspace of V.

 $P^{f}_{-}$ . Trivial

**Def**<sup><u>n</u></sup>.  $W_1, W_2$  subspaces of V are complementary if  $W_1 \cap W_2 = \{0\}$  &  $W_1 \cup W_2 = V$ .

**Prop**<sup>**n**</sup>.  $W_1, W_2$  subspaces of  $V, W_1 \notin W_2$  are complementary  $\iff v \in V$  can be written uniquely as  $v = w_1 + w_2$  where  $w_1 \in W_1 \notin w_2 \in W_2$ .

 $\begin{array}{l} P_{2}^{f}. \ `` \Longrightarrow `` Suppose W_{1}, W_{2} \text{ complementary then } W_{1}+W_{2}=V \text{ so can find } w_{1} \in W_{1} \& w_{2} \in W_{2} \ s.t. \ v=w_{1}+w_{2} \\ \text{suppose } w_{1}' \in W_{1} \& w_{2}' \in W_{2} \ s.t. \ v=w_{1}'+w_{2}' \ \text{now } w_{1}+w_{2}=w_{1}'+w_{2}', \ w_{1}-w_{1}'=w_{2}'-w_{2} \ LHS \in W_{1} \\ RHS \in W_{2} \ \text{hence both } \in W_{1} \cap W_{2} \ \text{but } W_{1} \cap W_{2} = \{0\} \ \text{hence } w_{1}=w_{1}' \& w_{2}=w_{2}'. \\ `` \Leftarrow `` \ Suppose \ \text{every } v \in V \ \text{can be uniquely written } v=w_{1}+w_{2}, \ \text{with } w_{1} \in W_{1} \ w_{2} \in W_{2} \\ (\text{Obv.}) \ W_{1}+W_{2}=V. \ \text{If } 0 \neq v \in W_{1} \cap W_{2} \ \text{then } v=v+0, \ v \in W_{1} \ v=0+v, \ v \in W_{2}. \ \text{Hence } \\ W_{1} \cap W_{2} = \{0\} \implies W_{1}, W_{2} \ \text{complementary}. \end{aligned}$ 

## Linear Transformations

**Def**<sup><u>n</u></sup>. Let U & V be vector spaces over K, a linear transformation or linear map T from U to V is a function  $T: U \to V$  s.t.  $T(\alpha u_1 + \beta u_2) = \alpha T(u_1) + \beta T(u_2) \forall u_1, u_2 \in U \& \forall \alpha, \beta \in K$ 

**Lemma.** *i*) 
$$T(0_U) = 0_V$$

*ii)* 
$$T(-u) = -T(u)$$

 $P^{\underline{f}}$ . i)  $T(0_U) = T(0_U + 0_U) = T(0_U) + T(0_U) \implies T(0_U) = 0_V$ 

ii) 
$$T((-1)u) = (-1)T(u)$$

QED

**Prop**<sup>**n**</sup>. Let U, V be vector spaces over  $K, u_1, \ldots, u_n \in U$  basis  $v_1, \ldots, v_n \in V$  then  $\exists !T \ s.t. \ T : U \to V$  linear map with  $T(u_i) = v_i$ .

 $P_{-}^{f}$ . Let  $u \in U$  then  $u = \alpha_{1}u_{1} + \dots + \alpha_{n}u_{n}$  so  $T(u) = T(\alpha_{1}u_{1} + \dots + \alpha_{n}u_{n}) = \alpha_{1}v_{1} + \dots + \alpha_{n}v_{n} = v \in V$ hence T is uniquely determined. QED **Def**<sup><u>n</u></sup>. Let  $T: U \to V$  be a linear map. The image of T Im(T) is the set of vectors  $v \in V$  s.t. v = T(u) for some  $u \in U$ . The kernel of T ker(T) is the set of vectors  $u \in U$  s.t.  $T(u) = 0_V$ .

 $Im(T) := \{T(u) : u \in U\}, \quad \ker(T) := \{u \in U : T(u) = 0_V\}$ 

**Prop**<sup>**n**</sup>. Let  $T: U \to V$  be a linear map, then Im(T) is a subspace of  $V \notin ker(T)$  is a subspace of U.

 $P^{\underline{f}}. \ \alpha v_1 + \beta v_2 = \alpha T(u_2) + \beta T(u_2) = T(\alpha u_1) + T(\beta u_2) = T(\alpha u_1 + \beta u_2) \in \operatorname{Im}(T) \text{ for } u_1, u_2 \in U \& \alpha, \beta \in K$  $T(\alpha u_1 + \beta u_2) = T(\alpha u_1) + T(\beta u_2) = \alpha T(u_1) + \beta T(u_2) = \alpha 0_V + \beta 0_V = 0_V \text{ where } \alpha u_1 + \beta u_2 \in \ker(T)$ QED

**Def**<sup>**n**</sup>. Let  $T: U \to V$  be a linear map dim(Im(T)) is called the <u>rank</u>, dim(ker(T)) is called the nullity.

**Th**<sup>m</sup>. <u>Rank-Nullity Theorem</u>: Let U, V be vector spaces over K with U finite dimensional. Let  $T : U \to V$  be a linear map. Then

$$\operatorname{rank}(T) + \operatorname{null}(T) = \dim(U)$$

 $\begin{array}{l} P^{\underline{f}}. \text{ Since } \ker(T) \text{ is a subspace of } U \text{ (Both finite dimensional). Let } \operatorname{null}(T) = s \And e_1, \ldots, e_s \text{ be a basis of } \ker(T). \text{ Now extend to a basis of } U: e_1, \ldots, e_s, f_1, \ldots, f_r. \text{ Now } \dim(U) = s + r. \ T(e_1), \ldots, T(e_s), T(f_1), \ldots, T(f_r) \text{ span } \operatorname{Im}(T) \And \operatorname{since} T(e_1), \ldots, T(e_s) \text{ all } = 0_V \text{ then } T(f_1), \ldots, T(f_r) \text{ span } \operatorname{Im}(T). \\ \text{Suppose } \alpha_1 T(f_1) + \cdots + \alpha_r T(f_r) = 0_V \text{ then } T(\alpha_1 f_1 + \cdots + \alpha_r f_r) = 0_V \text{ so } \alpha_1 f_1 + \cdots + \alpha_r f_r \in \ker(T) \text{ but } e_1, \ldots, e_s \text{ is a basis of } \ker(T) \text{ hence } \exists \beta_j \in K \text{ s.t. } \alpha_1 f_1 + \cdots + \alpha_r f_r = \beta_1 e_1 + \cdots + \beta_s e_s \Longrightarrow \alpha_1 f_1 + \cdots + \alpha_r f_r - \beta_1 e_1 - \cdots - \beta_s e_s = 0_U \text{ but } e_1, \ldots, e_s, f_1, \ldots, f_r \text{ is a basis of } U \text{ hence } \alpha_i, \beta_j = 0 \ \forall i \Longrightarrow f_1, \ldots, f_r \text{ linearly independent, hence } f_1, \ldots, f_r \text{ is a basis of } \operatorname{Im}(T) \text{ hence } \operatorname{rank}(T) + \operatorname{null}(T) = r + s = \dim(U) \end{array}$ 

QED

**Cor**<sup><u>ly</u></sup>. Let  $T: U \to V$  be a linear map where dim $(U) = \dim(V) = n$ . Then the following properties of T are equivalent:

- 1. T is surjective.
- 2.  $\operatorname{rank}(T) = n$
- 3. null(T) = 0
- 4. T is injective.
- 5. T is bijective.

 $\begin{array}{l} P^{\underline{f}}. \text{ (i)} \implies \operatorname{Im}(T) = V \implies \operatorname{rank}(T) = \dim(V) = n \implies (\text{ii}) \\ (\text{ii}) \implies \operatorname{Im}(T) \text{ subspace of } V \text{ dimension } n \implies \operatorname{Im}(T) = V \implies (\text{i}) \\ (\text{ii}) \implies \dim(U) = n = \operatorname{rank}(T) + \operatorname{null}(T) \implies (T) = 0 \implies (\text{iii}) \\ (\text{iii}) \implies \ker(T) = \{0_V\} T(u_1) = T(u_2) \implies T(u_1 - u_2) = 0_V \implies u_1 - u_2 \in \ker(T) = \{0_V\} \implies u_1 = u_2 \implies (\text{iv}) \\ (\text{iv}) \implies (\text{iii}) \\ \text{finally (i)}\&(\text{iv}) \iff (\text{v}) \end{array}$ 

 $Def^{\underline{n}}$ . If the above is satisfied then T is called non-singular, otherwise singular.

**Def**<sup><u>n</u></sup>. Addition & Scalar multiplication of linear maps: Let  $T_1 : U \to V$  &  $T_2 : U \to V$  then define  $\alpha T_1 + \beta T_2 : U \to V$  to be  $(\alpha T_1 + \beta T_2)(u) = \alpha T_1(u) + \beta T_2(u) \ \forall \alpha, \beta \in K \ \forall u \in U$ 

**Def**<sup><u>n</u></sup>. Composition of linear maps: Let  $T_1: U \to V \& T_2: V \to W$  then define  $T_2 \circ T_1: U \to W$  to be  $(T_2 \circ T_1)(u) = T_2(T_1(u)) \forall u \in U$ 

#### Linear Transformations & Matrices

Let  $T: U \to V$  be a linear map, where  $\dim(U) = n \& \dim(V) = \emptyset e_1, \dots, e_n$  is a basis of  $U \& f_1, \dots, f_m$  a basis V. Now

$$T(e_{1}) = \alpha_{11}f_{1} + \alpha_{21}f_{2} + \dots + \alpha_{m1}f_{m}$$

$$T(e_{2}) = \alpha_{12}f_{1} + \alpha_{22}f_{2} + \dots + \alpha_{m2}f_{m}$$

$$\vdots :$$

$$T(e_{n}) = \alpha_{1n}f_{1} + \alpha_{2n}f_{2} + \dots + \alpha_{mn}f_{m}$$

$$with \alpha_{ij} \in K$$

$$(1)$$

or  $T(e_j) = \sum_{i=1}^m \alpha_{ij} f_i$  for  $1 \le j \le n \& A = (\alpha_{ij})$  is the matrix of the linear map T. This can be written  $[T]_f^e = A$ 

**Th<sup>m</sup>**. Let U, V be vector spaces over K of dimensions n, m respectively. For a giver choice of bases of  $U \notin V$  there is a one to one correspondence between the set  $\operatorname{Hom}_{K}(U, V)$  of linear maps  $U \to V \notin$  the set  $K^{m \times n}$  of  $m \times n$  matrices over K.

 $P^{\underline{f}}$ . Use the above formulation

QED

QED

**Prop**<sup>**n**</sup>. Let  $T: U \to V$  be a linear map. Let  $A = (\alpha_{ij})$  represent T wrt given bases of  $U \ & V$ . Then  $T(u) = v \iff Au = v$  for  $u \in U \ & v \in V$ .

$$P_{\underline{-}}^{f}. T(u) = T\left(\sum_{j=1}^{n} \lambda_{j} e_{j}\right) = \sum_{j=1}^{n} \lambda_{j} T(e_{j}) = \sum_{j=1}^{n} \lambda_{j} \left(\sum_{i=1}^{m} \alpha_{ij} f_{i}\right) = \sum_{i=1}^{m} \underbrace{\left(\sum_{j=1}^{n} \alpha_{ij} \lambda_{j}\right)}_{=*} f_{i} = \sum_{i=1}^{m} \underbrace{\mu_{i}}_{*=} f_{i}$$
OED

**Prop**<sup>**n**</sup>. Let  $T_1, T_2 : U \to V$  be linear maps & A & B the respective matrices (& wrt the same bases). Then  $\alpha T_1 + \beta T_2$  has matrix  $\alpha A + \beta B$ .

 $P^{\underline{f}}$ . Trivial

**Th<sup>m</sup>.** Let  $T_1: V \to W$  be a linear map with  $\ell \times m$  matrix  $A = (\alpha_{ij}) & \text{if let } T_2: U \to V$  be a linear map with  $m \times n$  matrix  $B = (\beta_{ij})$ . Then the composite map  $T_1 \circ T_2$  has matrix AB.

 $P_{-}^{f}$ . Similar to  $P_{-}^{f}$  of  $T(u) = v \iff Au = v$  but  $T_{1}(T_{2}(u)) = ABu$  QED

# Elementary Operations & Rank of a Matrix

Elementary Row Operations:

- (R1) For  $i \neq j$  add a multiple of  $r_j$  to  $r_i$  ( $r_i, r_j$  are rows).
- (R2) Interchange two rows.
- (R3) Multiply a row by a non-zero scalar.

Def<sup><u>n</u></sup>. A matrix satisfying:

- i) All zero rows below all non-zero rows.
- ii) Let  $r_1, \ldots, r_s$  be the non-zero rows, then all  $r_i$  has a 1 as its first entry.
- iii) The first non-zero entry of each row is strictly to the right og the first non-zero entry of the row above.
- iv) If row i is non-zero all entries below the first non-zero element are zero.

is said to be in upper echelon form.

**Def**<sup><u>n</u></sup>. A matrix in upper echelon form satisfying:

v) If row i is non-zero then all entries above and below the first non-zero element are zero.

is said to be in row reduced form.

**Th**<sup>m</sup>. Every matrix can be brought to row reduced form by elementary row operations.

 $P^{\underline{f}}$ . Algorithm:

- 1) If  $\alpha_{ij}$  & all entries below are zero move one place to the right (i, j+1) & goto (1) unless j=n
- 2) If  $\alpha_{ij} = 0$  but not all entries below are, apply (R2) to exchange rows.
- 3) If  $\alpha_{ij} \neq 1$  apply (R3) using  $\alpha_{ij}^{-1}$ .
- 4) Now apply (R1) s.t. all entries above & below that every entry are zero.
- 5) Move down one & right one (i + 1, j + 1) unless i = m or j = n. QED

Elementary Column Operations:

- (C1) For  $i \neq j$  adda multiple of  $c_i$  to  $c_i$  ( $c_i, c_j$  are columns).
- (C2) Interchange two columns.
- (C3) Multiply a column by a non-zero scalar.

**Th<sup>m</sup>**. By applying elementary row & column operations a matrix can be brought into the form  $\left( \begin{array}{c|c} I_s & 0_{s,n-s} \\ \hline 0 & 0_{s,n-s} \end{array} \right).$ 

 $P^{\underline{f}}$ . Row reduce & use column operations.

Def<sup><u>n</u></sup>. A matrix in the above format is said to be in <u>Smith normal form</u>.

**Lemma.** Let  $T: U \to V$  be a linear map &  $e_1, \ldots, e_n$  a basis of U then the rank of T is the largest independent subset of  $T(e_1)$ , dots,  $T(e_n)$ .

- **Def**<sup><u>n</u></sup>. 1) The row space of A is the subspace of  $K^n$  spanned by the rows of A. The row rank is the dimension of the row space.
  - 2) The column space of A is the subspace of  $K^n$  spanned by the columns of A. The column rank is the dimension of the column space.

**Th<sup>m</sup>**. Suppose the linear map T has matrix A then rank(T) = column rank(A)

 $P^{f}_{-}$ . Use above lemma.

**Th**<sup> $\underline{m}$ </sup>. Applying elementary row operations of elementary column operations does not change the row  $\mathcal{E}$  column rank.

 $P^{\underline{f}}$ . Obviousness.

 $\mathbf{Cor}^{\underline{ly}}$ . The number of non-zero rows in the Smith normal form of a matrix A is equal to both the row  $\mathcal{E}$  column rank.

 $P_{-}^{f}$ . Elementary row & column operations don't change row or column ranks so:

 $\operatorname{row}\operatorname{rank}(A) = \operatorname{row}\operatorname{rank}(\operatorname{Smith} \operatorname{normal} \operatorname{form} \operatorname{of} A) = s = \operatorname{column}\operatorname{rank}(\operatorname{Smith} \operatorname{normal} \operatorname{form} \operatorname{of} A) = \operatorname{column}\operatorname{rank}(A)$ 

**Cor**<sup>ly</sup>. The rank of a matrix A is equal to the number of rows that are non-zero in upper echelon form  $P^{\underline{f}}$ . Non-zero rows in upper echelon form are linearly independent. QED

**Th<sup>m</sup>**. Let A be the augmented  $n \times (m+1)$  matrix of a linear system. Let B be the matrix obtained from A by removing the last column. The system of equations have a solution  $\iff \operatorname{rank}(A) = \operatorname{rank}(B)$ 

QED

QED

QED

#### The inverse of a Linear Transformation & of a Matrix

**Def**<sup>**n**</sup>. Let  $T: U \to V$  be a linear transformation with corresponding matrix  $A(m \times n)$ . If  $\exists T^{-1}: V \to U$  with  $TT^{-1} = I_V \& T^{-1}T = I_U$  then T is said to be <u>invertible</u> &  $T^{-1}$  is called the <u>inverse</u>. If so,  $A^{-1}$  is the  $(n \times m)$  matrix &  $AA^{-1} = I_m \& A^{-1}A = I_n$ , then A is said to be <u>invertible</u> &  $A^{-1}$  is called the <u>inverse</u>.

**Lemma.** Let A be a matrix of a linear map T. T is invertible  $\iff A$  is invertible.  $T^{-1} & A^{-1}$  are unique.

 $P_{-}^{f}$ . Bijection between matrices & linear mpas.

QED

QED

**Th**<sup>m</sup>. A linear map T is invertible  $\iff$  T is non-singular. In particular if T is invertible then m = n so only square matrices are invertible.

 $P_{-}^{f}$ . Map required to be injective so is inverse, hence bijection etc.

**Prop**<sup>**n**</sup>. The row reduced form of an invertible  $n \times n$  matrix A is  $I_n$ 

Elementary matrices:

- $E(n)_{\lambda,i,j}^1$   $(i \neq j)$   $n \times n$  matrix like the identity butt with a non-zero entry  $\lambda$  in the  $(i, j)^{th}$  position.
- $E(n)_{i,j}^2$  Like the  $n \times n$  identity with  $i^{th} \& j^{th}$  rows interchanged.
- $E(n)^3_{\lambda,i}$  ( $\lambda \neq 0$ ) like the  $n \times n$  identity  $\lambda$  in the  $(i,i)^{th}$  position.

Th<sup>m</sup>. An invertible matrix is a product of elementary matrices.

**Th**<sup> $\underline{\mathbf{m}}$ </sup>**.** Let A be an  $n \times n$  matrix:

- i) The homogeneous system ax = 0 has a non-trivial solution  $\iff A$  is singular.
- ii) The system Ax = b has a unique solution  $\iff A$  is non-singular.
- $P^{\underline{f}}$ . i) The solution is nullspace(A) if T corresponds to A nullspace(T) = ker(T) =  $\{0\}$   $\iff$  null(T) = 0  $\iff$  T non-singular gives no solutions hence require A singular.
  - ii) If A singular then its nullity> 0 so nullspace $(A) \neq \{0\} \implies$  no solutions OR solutions are x+nullspace(A) hence not unique.

QED

## The Determinant of a Matrix

**Def**<sup><u>n</u></sup>. A permutation  $\phi$  is said to be <u>even</u> if it is a composition of an even number of transpositions & sign( $\phi$ ) = +1 & <u>odd</u> if a composition of an odd number of transpositions & sign( $\phi$ ) = -1.

**Def**<sup><u>n</u></sup>. The <u>determinant</u> of an  $n \times n$  matrix  $A = (\alpha_{ij})$  is the scalar quantity

$$\underline{\det(A)} := \sum_{\phi \in S_n} \operatorname{sign}(\phi) \alpha_{1\phi(1)} \cdots \alpha_{n\phi(n)}$$

Th<sup>m</sup>. Elementary row operations affect the determinant as follows:

- $i) \det(I_n) = 1.$
- ii) Applying (R2) changes the determinants sign.
- iii) If A has two equal rows det(A) = 0.
- iv) Applying (R1) does not change the determinant.
- v) Applying (R3) multiplies the determinant by  $\lambda$ .

 $P^{\underline{f}}$ . i)  $\alpha_{ij} = 0 \ \forall i \neq j$  so only identity permutation is non-zero  $\det(I) = \alpha_{11} \cdots \alpha_{nn} = 1$ .

- ii)  $\det(B) = \sum_{\phi \in S_n} \operatorname{sign}(\phi) \beta_{1\phi(1)} \cdots \beta_{n\phi(n)}$  let  $\varphi = \phi \circ (i, j)$  now  $\operatorname{sign}(\varphi) = -\operatorname{sign}(\phi)$  hence  $= \sum_{\varphi \in S_n} -\operatorname{sign}(\phi) \alpha_{1\varphi(1)} \cdots \alpha_{n\varphi(n)} = -\det(A).$
- iii) Use part (ii) & swap rows that are the same now  $det(A) = -det(A) \implies det(A) = 0$ .
- iv)

$$det(B) = \sum_{\phi \in S_n} sign(\phi) \alpha_{1\phi(1)} \cdots (\alpha_{i\phi(i)} + \lambda \alpha_{j\phi(j)}) \cdots \alpha_{n\phi(n)}$$
$$= \sum_{\phi \in S_n} sign(\phi) \alpha_{1\phi(1)} \cdots \alpha_{n\phi(n)} + \lambda \sum_{\phi \in S_n} sign(\phi) \alpha_{1\phi(1)} \cdots \alpha_{j\phi(j)} \alpha_{j\phi(j)} \cdots \alpha_{n\phi(n)}$$

Second term = 0 since  $\alpha_{j\phi(j)}$  repeated is the same as two equal rows.

v) 
$$\det(B) = \sum_{\phi \in S_n} \operatorname{sign}(\phi) \alpha_{1\phi(1)} \cdots \lambda \alpha_{i\phi(i)} \cdots \alpha_{n\phi(n)} = \lambda \sum_{\phi \in S_n} \operatorname{sign}(\phi) \alpha_{1\phi(1)} \cdots \alpha_{n\phi(n)} = \lambda \det(A)$$
  
QED

Def<sup>n</sup>. A matrix is upper triangular if all entries below the leading diagonal are zero.

Cor<sup>ly</sup>. The determinant of an upper triangular matrix is the product of its diagonal entries.

**Def**<sup><u>n</u></sup>. Let  $A = (\alpha_{ij})$  be an  $m \times n$  matrix. Define the transpose  $\underline{A}^T$  of A to be the  $n \times m$  matrix  $(\beta_{ij}) = (\alpha_{ji})$ .

**Th**<sup>m</sup>. Let  $A = (\alpha_{ij})$  be an  $n \times n$  matrix, then  $\det(A^T) = \det(A)$ .  $P^{\underline{f}}$ .

$$det(A^{T}) = \sum_{\phi \in S_{n}} sign(\phi)\beta_{1\phi(1)} \cdots \beta_{n\phi(n)}$$
$$= \sum_{\phi \in S_{n}} sign(\phi)\alpha_{\phi(1)1} \cdots \alpha_{\phi(n)n}$$
$$= \sum_{\phi \in S_{n}} sign(\phi)\alpha_{1\phi(1)} \cdots \alpha_{n\phi(n)}$$
$$= det(A)$$

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Cor<sup>ly</sup>. Elementary column operations affect the determinant as row operations do.

**Th<sup>m</sup>.** An  $n \times n$  matrix A has det $(A) = 0 \iff A$  is singular.

 $P^{\underline{f}}$ . Row reduce A to obtain A',  $\det(A) = 0 \iff \det(A') = 0$ , A' is upper triangular hence  $\det(A) = 0 \iff$  one diagonal entry is  $0 \iff A'$  has a zero row  $\iff$  row rank $(A) < n \iff A$  is singular.

**Lemma.** If E is an elementary matrix & B any matrix both  $n \times n$  then det(EB) = det(E) det(B).

 $P_{-}^{f}$ . det $(E^{1}) = 1$ , det $(E^{2}) = -1$ , det $(E^{3}) = \lambda$  & consider row operations. QED

**Th**<sup>**m**</sup>. For any two  $n \times n$  matrices  $A \notin B$ : det(AB) = det(A) det(B).

$$P^{\underline{f}}. \det(AB) = \det(E_1) \det(E_2 \cdots E_n B) = \det(E_1) \cdots \det(E_n) \det(B) = \det(E_1 \cdots E_n) \det(B)$$
$$= \det(A) \det(B)$$
QED

**Def**<sup><u>n</u></sup>. Let  $A = (a_{ij})$  be an  $n \times n$  matrix. Let  $A_{ij}$  be the matrix  $(n-1) \times (n-1)$  obtained by eliminating the  $i^{th}$  row &  $j^{th}$  column of A.  $M_{ij} = \det(A_{ij})$  is called the  $(i, j)^{th}$  minor of A.

**Def**<sup>**n**</sup>. 
$$c_{ij} = (-1)^{i+j} M_{ij} = (-1)^{i+j} \det(A_{ij})$$
 is called the  $\underline{(i,j)^{th}}$  cofactor of  $A$ .

**Th<sup>m</sup>**. Let A be an  $n \times n$  matrix.

i) Expansion by  $i^{th}$  row: det $(A) = \alpha_{i1}c_{i1} + \cdots + \alpha_{in}c_{in} = \sum_{j=1}^{n} \alpha_{ij}c_{ij}$ .

ii) Expansion by  $j^{th}$  column:  $det(A) = \alpha_{1j}c_{1j} + \dots + \alpha_{nj}c_{nj} = \sum_{i=1}^{n} \alpha_{ij}c_{ij}$ .

 $P^{\underline{f}}$ .

$$\sum_{\substack{\phi \in S_n \\ \phi(n) = n}} \operatorname{sign}(\phi) \alpha_{1\phi(1)} \cdots \alpha_{n\phi(n)} = \sum_{\substack{\phi \in S_{n-1}}} \operatorname{sign}(\phi) \alpha_{1\phi(1)} \cdots \alpha_{n-1\phi(n-1)}$$

now det
$$(B) = \beta_{nn} N_{nn} = (-1)^{i+j} \alpha_{ij} M_{ij} = \alpha_{ij} c_{ij}$$
 hence det $(A) = \sum \alpha_{ij} c_{ij}$  for fixed *i* or *j*. QED

**Def**<sup><u>n</u></sup>. Let A be an  $n \times n$  matrix. Define the adjugate of A,  $\operatorname{adj}(A)$ , to be the  $n \times n$  matrix with  $(i, j)^{th}$  element the cofactor  $c_{ji}$ . i.e. the transpose of the matrix of cofactors.

**Th**<sup>$$\mathbf{m}$$</sup>.  $Aadj(A) = det(A)I_n = adj(A)A$ 

$$P^{f}_{-} A = (\alpha_{ij}) \operatorname{adj}(A) = (c_{ij}) \text{ so } A\operatorname{adj}(A) = \left(\sum_{j=1}^{n} \alpha_{ij} c_{kj}\right). \text{ If } k \neq i \sum_{j=1}^{n} \alpha_{ij} c_{kj} = 0 \text{ hence}$$
$$\operatorname{Aadj}(A) = \operatorname{det}(A)I_{n}$$
QED

**Cor**<sup>ly</sup>. If det(A)  $\neq 0$  then  $A^{-1} = \frac{1}{\det(A)} adj(A)$ .

# Change of Basis & Equivalent Matrices

**Prop**<sup>**n**</sup>. The change of basis matrix is invertible. i.e. if P is the change of basis matrix from  $e_i$ 's to  $e'_i$ 's  $\mathcal{E} Q$  is the change of basis ,matrix from  $e'_i$ 's to  $e_i$ 's then  $P = Q^{-1}$ .

 $P^{\underline{f}}$ . Consider  $I_U: U \to U \to U$  where the first & last spaces use the same bases  $\implies PQ = I_n \implies P = Q^{-1}$  QED

**Prop**<sup>**n**</sup>. Pv = v' where  $v = \alpha_1 e_1 + \dots + \alpha_n e_n$  &  $v' = \beta_1 e'_1 + \dots + \beta_n e'_n$ .

 $P^{\underline{f}}$ . Obviousness

**Th<sup>m</sup>**. Let  $T: U \to V$  be a linear map e, e' bases of U, f, f' bases of V

$$P = [Id_U]_{e'}^e \quad Q = [Id_V]_{f'}^f \quad A = [T]_f^e \quad B = [T]_{f'}^{e'}$$

Then  $B = QAP^{-1}$ .

$$P_{-}^{f}$$
.  $e \xrightarrow{P} e' \xrightarrow{B} f'$  hence  $BP \quad e \xrightarrow{A} f \xrightarrow{Q} f'$  hence  $QA \implies BP = QA$  QED

**Def**<sup><u>n</u></sup>.  $A\&B \ m \times n$  matrices are <u>equivalent</u> iff there are invertible matrices  $P \ n \times n$  matrix &  $Q \ m \times m$  matrix s.t. B = QAP.

**Th<sup>m</sup>**. Let A & B be  $m \times n$  matrices over K the following conditions are equivalent:

- i) A & B are equivalent.
- ii) A&B represent the same linear map wrt. different bases.
- iii)  $A \mathfrak{E} B$  have the same rank.
- iv) B can be obtained from A using elementary row  $\mathfrak{C}$  column operations.
- $P_{-}^{t}$ . (i)  $\iff$  (ii) by the previous Th<u>m</u>.

(ii)  $\implies$  (iii) Since A&B represent the same linear map  $\operatorname{rank}(A) = \operatorname{rank}(T) = \operatorname{rank}(B)$ .

(iii)  $\implies$  (iv) Both A&B have rank s so can be brought to Smith normal form by elementary row & column operations. i.e:  $A \leftrightarrow S.N.F. \leftrightarrow B$ 

(iv)  $\implies$  (i)  $B = R_r R_{r-1} \cdots R_1 A C_1 C_2 \cdots C_s$  where  $R_i \& C_i$  are row & column operation elementary matrices, hence B = QAP QED

#### Similar Matrices, Eigenvectors & Eigenvalues

**Def**<sup><u>n</u></sup>. Two  $n \times n$  matrices over K are said to be <u>similar</u> if  $\exists$  an  $n \times n$  matrix P which is invertible with  $B = P^{-1}AP$ .

Def<sup>n</sup>. A matrix similar to a diagonal matrix is said to be diagonalisable.

**Def**<sup><u>n</u></sup>. Let  $T: V \to V$  be a linear map where V is a vector space over K. Suppose  $v \in V$  non-zero & some  $\lambda \in K$  we have  $T(v) = \lambda v$ . Then v is called an eigenvector of T &  $\lambda$  an eigenvalue of T corresponding to v.

**Def**<sup>**n**</sup>. Let A be an  $n \times n$  matrix over K. Suppose some non-zero vector v & scalar  $\lambda \in K$  satisfy  $Av = \lambda v$ . Then v is called an eigenvector of A &  $\lambda$  an eigenvalue of A corresponding to v.

**Th<sup>m</sup>**. Let A be an  $n \times n$  matrix. Then  $\lambda$  is an eigenvalue  $\iff \det(A - \lambda I_n) = 0$ .

 $P_{-}^{f} \stackrel{\text{``}}{\Longrightarrow} \stackrel{\text{``}}{\Longrightarrow} Av = \lambda v \text{ so } Av = \lambda I_{n}v, (A - \lambda I_{n})v = 0 \text{ for a solution } \det(A - \lambda I_{n}) = 0.$  $\stackrel{\text{``}}{\longleftarrow} \det(A - \lambda I_{n}) = 0 \text{ then } A - \lambda I_{n} \text{ is singular, so has solutions } \Longrightarrow \exists x \ s.t. \ Av = \lambda v.$ QED

**Def**<sup><u>n</u></sup>. For an  $n \times n$  matrix A the equation det $(A - \lambda x) = 0$  is called the characteristic equation of A & det $(A - \lambda x)$  is called the characteristic polynomial.

**Th<sup>m</sup>**. Similar matrices have the same characteristic equation & hence the same eigenvalues.

 $P^{\underline{f}}. \text{ Let } A\&B \text{ be similar hence } B = P^{-1}AP \text{ for some } P$  $\det(B - xI_n) = \det(P^{-1}AP - xI_n) = \det(P^{-1}(A - xI_n)P) = \det(P^{-1})\det(A - xI_n)\det(P)$  $= \det(P^{-1})\det(P)\det(A - xI_n) = \det(A - xI_n)$ QED

**Prop**<sup>**n**</sup>. Suppose A is upper triangular, the eigenvalues are just  $\alpha_{ii}$ .

$$P_{-}^{I} \det(A - xI_n) = (\alpha_{11} - x) \cdots (\alpha_{nn} - x) = 0$$
QED

**Th**<sup>m</sup>. Let  $T: V \to V$  be a linear map, the matrix of T is diagonal wrt. to some basis of  $V \iff V$  has a basis consisting of eigenvectors of T.

**Equivalently**: Let A be an  $n \times n$  matrix over K. Then A is similar to a diagonal matrix  $\iff K^n$  has a basis of eigenvectors of A.

 $P_{-}^{f}$ . " $\implies$ " Suppose  $A = (\alpha_{ij})$  is diagonal wrt. a basis of V hence  $T(e_i) = \alpha_{ii}e_i$  so each  $e_i$  is an eigenvector.

" $\Leftarrow$ " Suppose  $e_i$ 's are a basis of V consisting of eigenvectors, so  $T(e_i) = \lambda e_i$  for some  $\lambda_i \in K$  so the matrix  $A = (\alpha_{ij})$  with  $\alpha_{ii} = \lambda_i$ .

**Th<sup>m</sup>**. Let  $\lambda_1, \ldots, \lambda_r$  be distinct eigenvalues with  $v_1, \ldots, v_r$  eigenvectors then  $v_1, \ldots, v_r$  are linearly independent.

 $P^{\underline{f}}$ . Induction on r.

 $\operatorname{Cor}^{\operatorname{ly}}$ . If A has distinct eigenvalues it is diagonalisable.

**Def**<sup><u>n</u></sup>. The scalar product of two vectors  $v = (\alpha_1, \ldots, \alpha_n)$  &  $w = (\beta_1, \ldots, \beta_n)$  in  $\mathbb{R}^n$  is defined to be  $v \cdot w = \sum_{i=1}^n \alpha_i \beta_i$ .

**Def**<sup><u>n</u></sup>. A basis  $b_1, \ldots, b_n$  is called <u>orthonormal</u> if  $b_i \cdot b_i = 1 \quad \forall i \& b_i \cdot b_j = 0 \quad \forall i \neq j$ .

**Def**<sup><u>n</u></sup>. An  $n \times n$  matrix A is said to be symmetric if  $A^T = A$ .

**Def**<sup><u>n</u></sup>. An  $n \times n$  matrix A is said to be orthogonal if  $A^T = A^{-1}$  or indeed  $AA^T = A^TA = I_n$ .

**Prop**<sup>n</sup>. An  $n \times n$  matrix A over  $\mathbb{R}$  is orthogonal  $\iff$  the rows  $r_1, \ldots, r_n$  form an orthonormal basis of  $\mathbb{R}^n$  $\iff$  the columns  $c_1, \ldots, c_n$  form an orthonormal basis of  $\mathbb{R}^n$ .

 $P^{\underline{f}}$ . Orthogonal matrix A is invertible hence row & column ranks are n & hence the rows & columns form bases.  $A^T A = I_n \implies r_i \cdot r_i = 1 \ \forall i \ \& \ r_i \cdot r_j = 0 \ \forall i \neq j$ . Similarly for the converse. QED

**Prop**<sup>**n**</sup>. Let A be a real symmetric matrix then A has a n eigenvalue in  $\mathbb{R}$  & all complex eigenvalues of A are in  $\mathbb{R}$ .

$$\begin{array}{l} P^{\underline{f}}. \ \det(A - \lambda I_n) \in \mathbb{R}[\lambda]_{deg=n} \quad Av = \lambda v \ \& \ A\bar{v} = \bar{\lambda}\bar{v} \ (\text{since } \bar{A} = A). \ \text{Also,} \ A^T = A \ \text{so} \ v^T A^T = v^T A = \lambda v^T \\ \text{so} \quad \begin{pmatrix} v^T A)\bar{v} = \lambda v^T \bar{v} \\ v^T (A\bar{v}) = v^T \bar{\lambda}\bar{v} \end{array} \right\} \implies \lambda v^T \bar{v} = \bar{\lambda} v^T \bar{v} \implies (\lambda - \bar{\lambda}) v^T \bar{v} = 0 \ \text{since} \ v^T \bar{v} \neq 0 \implies \lambda = \bar{\lambda} \ \text{hence} \ \lambda \in \mathbb{R} \\ \text{QED} \end{array}$$

**Prop**<sup><u>n</u></sup>. Let A be a real symmetric matrix  $\mathscr{C}$  let  $\lambda_1, \lambda_2$  be two distinct eigenvalues  $v_1, v_2$  then  $v_1 \cdot v_2 = 0$ .

 $P^{\underline{f}}. Av_1 = \lambda_1 v_1 \& Av_2 = \lambda_2 v_2 \text{ since } v_1^T A^T = \lambda_1 v_1^T, v_1^T Av_2 = \lambda_1 v_1^T v_2$  & similarly  $v_2^T Av_1 = \lambda v_2^T v_1 = \lambda_2 v_2^T v_1 \implies v_1^T Av_2 = \lambda_2 v_1^T v_2$   $\therefore (\lambda_2 - \lambda_1) v_1^T v_2 = 0 \text{ since } \lambda_2 - \lambda_1 \neq 0 \quad v_1 \cdot v_2 = 0$ QED

**Th<sup>m</sup>**. Let A be a real symmetric matrix.  $\exists$  a real orthogonal matrix P with  $P^{-1}AP = P^TAP$  diagonal.

 $P_{\cdot}^{f}$ . (Case of distinct eigenvalues only)  $\lambda_{i} \in \mathbb{R}$  eigenvalues  $\implies v_{i} \in \mathbb{R}^{n}$  eigenvectors  $v_{i} \cdot v_{j} = v_{i}^{T}v_{j} = 0$ for  $i \neq j$  since  $v_{i} \neq 0$   $v_{i} \cdot v_{i} = \alpha_{i} > 0$  so replacing each  $v_{i}$  with  $v_{i}/\sqrt{\alpha_{i}}$  we know  $v_{i} \cdot v_{i} = 1 \forall i$ . All the  $v_{i}$  are independent & form a basis (orthogonal). But  $P^{-1}AP$  is the diagonal matrix with entries  $\lambda_{1}, \ldots, \lambda_{n}$ . QED