## NATIONAL UNIVERSITY OF SINGAPORE MATHEMATICS SOCIETY

## PAST YEAR PAPER SOLUTIONS with credits to Ho Chin Fung

# $\begin{array}{ccc} \textbf{MA2101} & \textbf{Linear Algebra II} \\ & \text{AY } 2007/2008 \text{ Sem } 1 \end{array}$

### SECTION A

#### Question 1

(i)  $S = \{1, x, x^2\}$ . Then

$$T(1) = 1 + x^2$$

$$T(x) = x + 2x^2$$

$$T(x^2) = 1 + x^2$$

So

$$[T]_{\mathcal{S}} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 2 & 1 \end{pmatrix}$$

(ii) The characteristic polynomial of T is given by

$$c_T(x) = \begin{vmatrix} x-1 & 0 & -1 \\ 0 & x-1 & 0 \\ -1 & 2 & x-1 \end{vmatrix}$$
$$= (x-1)^3 - (x-1)$$
$$= x(x-1)(x-2)$$

So the eigenvalues of T are 0, 1 and 2.

(iii) Since T has 3 distinct eigenvalues, it is diagonalisable. To find eigenvectors of  $[T]_S$  corresponding to  $\lambda = 0$ , we solve the linear system  $([T]_S - 0I)\mathbf{x} = \mathbf{0}$ :

$$\left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \end{array}\right) \xrightarrow{R_3 - R_1 - 2R_2} \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array}\right)$$

Set  $x_3 = t$ . Then  $x_1 = -t$  and  $x_2 = 0$ . In particular, with t = -1, we obtain  $\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ .

Similar computations show that the vectors  $\begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$  are eigenvectors of  $[T]_S$ 

corresponding to  $\lambda = 1$  and  $\lambda = 2$  respectively. So  $1 - x^2$ , 2 - x, and  $1 + x^2$  are eigenvectors of T corresponding to  $\lambda = 0$ , 1, and 2 respectively. Let  $\mathcal{B} = \{1 - x^2, 2 - x, 1 + x^2\}$ .

Then  $\mathcal{B}$  is a basis for  $P_2(\mathbb{R})$  and  $[T]_{\mathcal{B}} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix}$ .

(i) We compute

$$\det[T_2 \circ T_1]_{\mathcal{B}_1} = \begin{vmatrix} 2 & 1 & 2 \\ 1 & 1 & 0 \\ 0 & 1 & -2 \end{vmatrix} \\
= 0.$$

So  $[T_2 \circ T_1]_{\mathcal{B}_1}$  is not invertible. Since  $[T_2 \circ T_1]_{\mathcal{B}_1}$  is the matrix of  $T_2 \circ T_1$  with respect to  $\mathcal{B}_1$ ,  $T_2 \circ T_1$  is also not invertible.

(ii) Let

$$[T_1]_{\mathcal{B}_2,\mathcal{B}_1} = \begin{pmatrix} a & -1 & c \\ b & 1 & d \end{pmatrix},$$
$$[T_2]_{\mathcal{B}_2,\mathcal{B}_1} = \begin{pmatrix} 1 & e \\ 0 & f \\ -1 & g \end{pmatrix}.$$

Then

$$[T_2]_{\mathcal{B}_1,\mathcal{B}_2}[T_1]_{\mathcal{B}_2,\mathcal{B}_1} = \begin{pmatrix} 1 & e \\ 0 & f \\ -1 & g \end{pmatrix} \begin{pmatrix} a & -1 & c \\ b & 1 & d \end{pmatrix}$$

$$\begin{pmatrix} 2 & 1 & 2 \\ 1 & 1 & 0 \\ 0 & 1 & -2 \end{pmatrix} = [T_2 \circ T_1]_{\mathcal{B}_1} = \begin{pmatrix} a+eb & -1+e & c+ed \\ bf & f & fd \\ -a+bg & 1+g & -c+dg \end{pmatrix}.$$

Comparing each entry, we obtain

$$a = 0$$
,  $b = 1$ ,  $c = 2$ ,  $d = 0$ ,  $e = 2$ ,  $f = 1$ ,  $g = 0$ .

So

$$[T_1]_{\mathcal{B}_2,\mathcal{B}_1} = \begin{pmatrix} 0 & -1 & 2 \\ 1 & 1 & 0 \end{pmatrix},$$
  
 $[T_2]_{\mathcal{B}_1,\mathcal{B}_2} = \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ -1 & 0 \end{pmatrix}.$ 

(iii) Since  $T_1$  is a linear transformation, we have

$$T_{1}(a + bx + cx^{2}) = aT_{1}(1) + bT_{1}(x) + cT_{1}(x^{2})$$

$$= a \begin{pmatrix} 0 \\ 1 \end{pmatrix} + b \begin{pmatrix} -1 \\ 1 \end{pmatrix} + c \begin{pmatrix} 2 \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} -b + 2c \\ a + b \end{pmatrix}.$$

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(i) We have

$$\det(xI - A) = c_A(x)$$
  
=  $(x-1)^6(x-2)^5(x-3)^2$ .

In particular, when x = 0,

$$\det(-A) = (-1)^{6}(-2)^{5}(-3)^{2}$$
$$(-1)^{6+5+2}\det(A) = (1)(-32)(9)$$
$$(-1)\det(A) = -288$$
$$\det(A) = 288.$$

(ii) All the possible Jordan canonical forms of A are

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(i) Let

$$X = \begin{pmatrix} b & 2a+b \\ a-2b & -2b \end{pmatrix} \in W.$$

Then,

$$X \quad = \quad a \left( \begin{array}{cc} 0 & 2 \\ 1 & 0 \end{array} \right) + b \left( \begin{array}{cc} 1 & 1 \\ -2 & -2 \end{array} \right).$$

So,  $\begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}$  and  $\begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix}$  spans W. Observe that  $\begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}$  and  $\begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix}$  are not linear multiples of each other. So  $\left\{ \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \right\}$  is linearly independent. Therefore,  $\left\{ \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \right\}$  is a basis for W.

(ii) For any  $X \in W^{\perp}$ ,

$$||F - \mathbf{proj}_{W^{\perp}}(F)|| \le ||F - X||.$$

Therefore,  $\mathbf{proj}_{W^{\perp}}(F)$  can be one such matrix for P.

Now we find  $\mathbf{proj}_{W^{\perp}}(F)$ . Applying Gram-Schmidt process to  $\left\{ \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \right\}$ , we obtain  $\left\{ \frac{1}{\sqrt{5}} \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}, \frac{1}{\sqrt{10}} \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \right\}$ , an orthonormal basis for W. So,

$$\mathbf{proj}_{W}(F) = \left\langle \begin{pmatrix} 4 & 0 \\ 5 & 7 \end{pmatrix}, \frac{1}{\sqrt{5}} \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix} \right\rangle \frac{1}{\sqrt{5}} \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix} + \left\langle \begin{pmatrix} 4 & 0 \\ 5 & 7 \end{pmatrix}, \frac{1}{\sqrt{10}} \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \right\rangle \frac{1}{\sqrt{10}} \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} = \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix} + (-2) \begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} = \begin{pmatrix} -2 & 0 \\ 5 & 4 \end{pmatrix}.$$

Then,

$$\begin{aligned} \mathbf{proj}_{W^{\perp}}(F) &= F - \mathbf{proj}_{W}(F) \\ &= \begin{pmatrix} 4 & 0 \\ 5 & 7 \end{pmatrix} - \begin{pmatrix} -2 & 0 \\ 5 & 4 \end{pmatrix} \\ &= \begin{pmatrix} 6 & 0 \\ 0 & 3 \end{pmatrix}. \end{aligned}$$

Therefore,

$$P = \begin{pmatrix} 6 & 0 \\ 0 & 3 \end{pmatrix}.$$

(a) False.

Let 
$$S = \{v_1, v_2, v_3\}$$
, where  $v_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ ,  $v_3 = \begin{pmatrix} 0 \\ 2 \\ 0 \end{pmatrix}$ . Clearly,  $S$  is a linearly dependent subset of  $\mathbb{R}^3$ . However  $v_1 = c_2v_2 + c_3v_3$  has no solutions.

(b) True.

Since  $T_2$  is an isomorphism,  $T_2^{-1}$  exists.

Let  $\mathbf{y}_1 \in \mathcal{R}(T_1)$ .

Then  $\exists \boldsymbol{x}_1 \in V \text{ s.t. } T_1(\boldsymbol{x}_1) = \boldsymbol{y}_1.$ 

$$y_1 = T_1(T_2T_2^{-1}(x_1))$$
  
=  $T_1T_2(T_2^{-1}(x_1)) \in \mathcal{R}(T_1T_2)$   
 $\therefore \mathcal{R}(T_1) \subseteq \mathcal{R}(T_1T_2).$ 

Let  $\boldsymbol{y}_2 \in \mathcal{R}(T_1T_2)$ .

Then  $\exists x_2 \in V \text{ s.t. } T_1 T_2(x_2) = y_2$ .

$$y_2 = T_1 T_2(x_2)$$

$$= T_1(T_2(x_2)) \in \mathcal{R}(T_1)$$

$$\therefore \mathcal{R}(T_1 T_2) \subseteq \mathcal{R}(T_1).$$

Thus,  $\mathcal{R}(T_1T_2) = \mathcal{R}(T_1)$ . Moreover, since V is finite dimensional,  $\mathcal{R}(T_1), \mathcal{R}(T_1T_2) \subset V$  are also finite dimensional. Therefore,  $rank(T_1T_2) = rank(T_1)$ .

(c) False.

Let 
$$A = \begin{pmatrix} 0 & 1 \\ 0 & -1 \end{pmatrix}$$
,  $B = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$ .

Let  $A = \begin{pmatrix} 0 & 1 \\ 0 & -1 \end{pmatrix}$ ,  $B = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$ . Both A and B have 2 eigenvalues each and are therefore diagonalizable. However,  $A + B = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$  $\begin{pmatrix} 0 & 2 \\ 0 & 0 \end{pmatrix} = 2J_2(0)$ , which is not diagonalizable.

### Question 6

(a) Claim: span $\{A^m : m = 0, 1, 2, 3, \dots\} = \text{span}\{A^n : n = 0, 1, 2, \dots, k-1\}.$ Let the minimum polynomial of A be  $m_A$ . By the division algorithm,  $\forall m \geq 0 \in \mathbb{Z}$ ,

$$x^m = m_A(x)q(x) + r(x)$$

for some polynomial p(x) and r(x) where  $\deg[r(x)] < \deg[m_A(x)] = k$ . Sub x = A,

$$A^{m} = m_{A}(A)q(A) + r(A)$$
$$= (0_{n})q(A) + r(A)$$
$$= r(A).$$

 $A^m$  can be expressed as a polynomial in A of powers up to (k-1). So,  $\forall m \ge 0 \in \mathbb{Z}, A^m \in \text{span}\{A^n : n = 0, 1, 2, \dots, k - 1\}.$ 

Thus, we have,  $\operatorname{span}\{A^m: m=0,1,2,3,\dots\} \subseteq \operatorname{span}\{A^n: n=0,1,2,\dots,k-1\}.$ Next, clearly,  $\operatorname{span}\{A^m: m=0,1,2,3,\dots\} \supseteq \operatorname{span}\{A^n: n=0,1,2,\dots,k-1\}.$ Therefore,

$$span\{A^m : m = 0, 1, 2, 3, \dots\} = span\{A^n : n = 0, 1, 2, \dots, k - 1\}.$$

Claim:  $\{A^n : n = 0, 1, 2, \dots, k-1\}$  is linearly independent.

Suppose not.

Then  $\exists c_1, c_2, \dots, c_{k-1} \in \mathbb{F}$  such that

$$c_{k-1}A^{k-1} + c_{k-2}A^{k-2} + \dots + c_1A + a_0I_n = 0_n.$$

The LHS is a polynomial in A that is equal to  $0_n$  and it is of degree (k-1) < k. This contradict to the condition that the minimal polynomial of A is of degree k.

Therefore,  $\{A^n : n = 0, 1, 2, \dots, k-1\}$  is linearly independent.

Therefore,  $\dim(\text{span}\{A^m: m=0,1,2,3,\dots\}) = \dim(\text{span}\{A^n: n=0,1,2,\dots,k-1\}) = k$ .

(b) Let  $c = m_A(0_{\mathbb{F}})$  be the constant term in the minimum polynomial of A. Then, there exists  $f(x) \in \mathbb{F}[x]$ , with degree  $(\deg[m_A(x)] - 1)$ , such that

$$m_A(x) = (f(x))x + c$$
  
i.e.  $xf(x) = m_A(x) - c$ .

Claim:  $c \neq 0_{\mathbb{F}}$ . Suppose not, then  $xf(x) = m_A(x)$ . Sub x = A, we have

$$Af(A) = m_A(A)$$
  
 $f(A) = A^{-1}0_n \quad (\because A \text{ is invertible.})$   
 $= 0_n.$ 

This results in a contradiction as  $\deg[f(x)] < \deg[m_A(x)]$ . So,  $c \neq 0_{\mathbb{F}}$  and thus  $c^{-1}$  exists. Now, let  $g(x) = -(c^{-1})f(x)$ . Then

$$xg(x) = -(c^{-1})xf(x)$$

$$= -(c^{-1})(m_A(x) - c)$$

$$= 1_{\mathbb{F}} - (c^{-1})m_A(x)$$

Sub x = A, we have

$$Ag(A) = I_n - (c^{-1})m_A(A)$$
  
=  $I_n - (c^{-1})0_n$   
=  $I_n$ 

This give us  $A^{-1} = g(A)$ .

(i) We have

$$\det \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1r} \\ a_{21} & a_{22} & \cdots & a_{2r} \\ \vdots & \vdots & & \vdots \\ a_{r1} & a_{r2} & \cdots & a_{rr} \end{pmatrix} \neq 0$$

$$\Rightarrow \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1r} \\ a_{21} & a_{22} & \cdots & a_{2r} \\ \vdots & \vdots & & \vdots \\ a_{r1} & a_{r2} & \cdots & a_{rr} \end{pmatrix} \text{ is invertible,}$$

$$\Rightarrow \text{ the set } \left\{ \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{r1} \end{pmatrix}, \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{r2} \end{pmatrix}, \dots, \begin{pmatrix} a_{1r} \\ a_{2r} \\ \vdots \\ a_{rr} \end{pmatrix} \right\} \text{ is linearly independent,}$$

$$\Rightarrow \text{ the system } \begin{cases} c_{1}a_{11} + c_{2}a_{12} + \cdots + c_{r}a_{1r} = 0 \\ c_{1}a_{21} + c_{2}a_{22} + \cdots + c_{r}a_{2r} = 0 \\ \vdots \\ c_{1}a_{r1} + c_{2}a_{r2} + \cdots + c_{r}a_{rr} = 0 \end{cases}$$

$$\Rightarrow \text{ the system } \begin{cases} c_{1}a_{11} + c_{2}a_{12} + \cdots + c_{r}a_{1r} = 0 \\ c_{1}a_{21} + c_{2}a_{22} + \cdots + c_{r}a_{2r} = 0 \\ \vdots \\ c_{1}a_{r1} + c_{2}a_{r2} + \cdots + c_{r}a_{rr} = 0 \end{cases}$$

$$\Rightarrow \text{ the system } \begin{cases} c_{1}a_{11} + c_{2}a_{12} + \cdots + c_{r}a_{rr} = 0 \\ c_{1}a_{21} + c_{2}a_{22} + \cdots + c_{r}a_{rr} = 0 \\ \vdots \\ c_{1}a_{n1} + c_{2}a_{n2} + \cdots + c_{r}a_{nr} = 0 \end{cases}$$

$$\Rightarrow \text{ the set } \begin{cases} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{pmatrix}, \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{n2} \end{pmatrix}, \dots, \begin{pmatrix} a_{1r} \\ a_{2r} \\ \vdots \\ a_{nr} \end{pmatrix} \end{cases} \text{ is linearly independent.}$$

(ii) Let dr(A) denotes the determinant rank of A. Then there exists S, an  $r \times r$  submatrix of A, where r = dr(A) and  $det(S) \neq 0$ .

Let A' be an  $n \times n$  matrix formed by rearranging the rows of A, in such a way that S can be obtained by deleting the last (n-r) rows and some (n-r) columns from A'.

Since  $\det S \neq 0$ , then by the result from part (i), there exists a set of r linearly independent vectors in the column set of A'. Therefore, the dimension of the column space of A' is at least r. Since A and A' are row-equivalent, we have  $\operatorname{Rank} A = \operatorname{Rank} A'$ . Thus,

$$\operatorname{Rank} A = \operatorname{Rank} A' \ge r = \operatorname{dr}(A).$$

Let k = (RankA). Certainly, we can find a set of k linearly independent columns in the column set of A. Keeping these k columns, we delete the remaining (n - k) columns from A to form B. Since B has k linearly independent columns, the rank of B is again k. Therefore, we can then find a set of k linearly independent rows in the row set of B. Next, delete the remaining (n - k) rows from B to form C. The rank of C is again k. C is an  $k \times k$  submatrix of A. C is therefore full rank and has a nonzero determinant. By the definition of determinant rank, C cannot be larger than  $dr(A) \times dr(A)$ . So, we have

$$\operatorname{Rank} A = k < \operatorname{dr}(A)$$
.

Therefore, dr(A) = RankA. The determinant rank of A is equal to the rank of A.

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(i)  $\forall \boldsymbol{u} \in S$ , we have  $\boldsymbol{u} \in U$ . Therefore, S is a subset of U. Since U is a vector space,  $\mathbf{0} \in U$ . Since W is a subspace,  $T(\mathbf{0}) = \mathbf{0} \in W$ . Therefore,  $\mathbf{0} \in S$ .  $\forall \boldsymbol{u}_1, \boldsymbol{u}_2 \in S, \alpha_1, \alpha_2 \in \mathbb{F}$ , we have

$$T(\mathbf{u}_1), T(\mathbf{u}_2) \in W$$

$$\Rightarrow \quad \alpha_1 T(\mathbf{u}_1) + \alpha_2 T(\mathbf{u}_2) \in W$$

$$\Rightarrow \quad T(\alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2) \in W$$

$$\Rightarrow \quad \alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2 \in S.$$

Therefore, S is a subspace of U.

(ii) Since  $T:U\to V$  is a linear transformation, by the Dimension Theorem, we have

$$dim(U) = rank(T) + nullity(T)$$
$$= dim(R(T)) + nullity(T).$$

Similarly,  $T|_S: S \to W$  is a linear transformation, and as  $\ker(T) \subseteq S$ , we have

$$dim(S) = rank(T|S) + nullity(T|S)$$
$$= dim(R(T) \cap W) + nullity(T).$$

We also have

$$\dim(R(T)\cap W) = \dim(R(T)) + \dim(W) - \dim(R(T) + W)$$
 i.e. 
$$\dim(R(T)) - \dim(R(T)\cap W) = \dim(R(T) + W) - \dim(W).$$

Thus,

$$\dim(U) - \dim(S) = \dim(R(T)) - \dim(R(T) \cap W)$$
$$= \dim(R(T) + W) - \dim(W)$$
$$\leq \dim(V) - \dim(W).$$

Therefore,

$$\dim(S) \ge \dim(U) - \dim(V) + \dim(W).$$

### Question 9

(i) Since T is a self-adjoint operator, the eigenvalues of T are real. Let  $\lambda \in \mathbb{R}$  be an eigenvalue of T, i.e.  $T(\boldsymbol{v}) = \lambda \boldsymbol{v}$  for some  $\boldsymbol{v} \in V$ . Then

$$\begin{aligned}
\langle T(\boldsymbol{v}), \boldsymbol{v} \rangle &\geq 0 \\
\langle \lambda \boldsymbol{v}, \boldsymbol{v} \rangle &\geq 0 \\
\lambda \langle \boldsymbol{v}, \boldsymbol{v} \rangle &\geq 0 \\
\lambda \|\boldsymbol{v}\|^2 &\geq 0 \\
\lambda &\geq 0. \quad (:: \|\boldsymbol{v}\|^2 \geq 0)
\end{aligned}$$

Therefore, all the eigenvalues of T are nonnegative.

(ii) Since T is a self-adjoint operator, T is diagonalisable. Then we may represent T by  $PAP^{-1}$  where P is a invertable matrix and A is a diagonal matrix having eigenvalues of T as diagonal entries.

$$A = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{pmatrix} , \text{ where } \lambda_i \text{'s are eigenvalues of } T.$$

By the result of part (i), all  $\lambda_i$ 's are nonnegative. Let

$$B = \begin{pmatrix} \sqrt{\lambda_1} & 0 & \dots & 0 \\ 0 & \sqrt{\lambda_2} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \sqrt{\lambda_n} \end{pmatrix}.$$

Consider  $(PBP^{-1})^2$ .

$$(PBP^{-1})^{2} = (PBP^{-1})(PBP^{-1})$$

$$= PB(P^{-1}P)BP^{-1}$$

$$= PBIBP^{-1}$$

$$= PB^{2}P^{-1}$$

$$= P\begin{pmatrix} \sqrt{\lambda_{1}} & 0 & \dots & 0 \\ 0 & \sqrt{\lambda_{2}} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \sqrt{\lambda_{n}} \end{pmatrix}^{2} P^{-1}$$

$$= P\begin{pmatrix} \lambda_{1} & 0 & \dots & 0 \\ 0 & \lambda_{2} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_{n} \end{pmatrix} P^{-1}$$

$$= PAP^{-1}$$

Let S be a linear operator on V represented by  $PBP^{-1}$ . Then  $S^2 = T$ .

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