

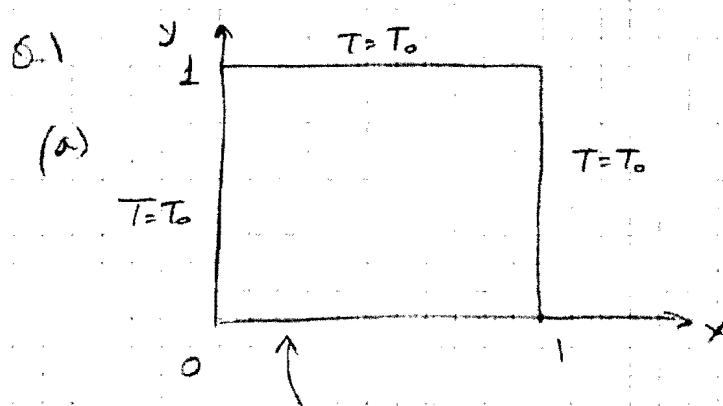
ENGINEERING TRIPPOS PART IB JUNE 2012
CHAIRMAN PROF. P DAVIDSON

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PAPER 7 SOLUTIONS – MATHEMATICAL METHODS

AUTHORS :

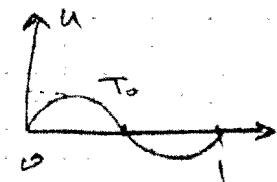
PROF. E MASTORAKOS



If $u = T - T_0$, then : $\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = g(x, y)$

b.c. : $u = 0$ at $x=1, y=0, y=1$

2. $u = T_0 \sin(2\pi x)$ at $y=0$



Using separation of variables: we seek a solution $u(x, y) = X(x)Y(y)$

$$\Rightarrow \frac{1}{X} \frac{d^2 X}{dx^2} = -\frac{1}{Y} \frac{d^2 Y}{dy^2} = \text{constant}$$

Address b.c.

$$u=0 \text{ at } x=1 \Rightarrow u(1, y) = X(1)Y(y) = 0$$

If we take $Y(y) = 0$, it means $u(x, y) = 0$ everywhere $\Rightarrow X(1) = 0$

Similarly, at $x=0$ gives: $X(0) = 0$

$$\text{At } y=1, \quad u(x, 1) = 0 \Rightarrow X(x)Y(1) = 0$$

To avoid the trivial solution, $Y(1) = 0$

$$\text{At } y=0, \quad u(x, 0) = T_0 \sin(2\pi x)$$

$$\Rightarrow X(x) = T_0 \sin(2\pi x) \quad \& \quad Y(0) = 1$$

Going back to the equations:

$$\frac{1}{X} \frac{\partial^2 X}{\partial x^2} = -k^2, \quad \frac{1}{Y} \frac{\partial^2 Y}{\partial y^2} = k^2 \quad (k > 0)$$

(We write the one for the X eqn since we know that it must have a sin dependence)

$$\Rightarrow k = 2\pi$$

$$\Rightarrow Y = A e^{2\pi y} + B e^{-2\pi y}$$

$$X = T_0 \sin(2\pi x)$$

$$\text{At } y=0, Y=1 \Rightarrow A+B=1$$

$$y=1, Y=0 \Rightarrow A e^{2\pi} + B e^{-2\pi} = 0$$

$$\Rightarrow B = 1 - A$$

$$\Rightarrow A e^{2\pi} + (1-A) e^{-2\pi} = 0$$

$$\Rightarrow A = \frac{-e^{-2\pi}}{e^{2\pi} - e^{-2\pi}} \quad ; \quad B = \frac{e^{2\pi}}{e^{2\pi} - e^{-2\pi}}$$

$$\therefore u(x,y) = T_0 \sin(2\pi x) \cdot (A e^{2\pi y} + B e^{-2\pi y})$$

with A & B from
above

(b) It is very difficult to solve the pde for $g \neq 0$. We are seeking $\oint g \cdot d\underline{y}$

So a 2-D version of Gauss theorem may be used:

$$\iint \nabla \cdot \underline{g} \, dA = \oint g \cdot d\underline{y}$$

But since $\underline{g} = -\nabla T$, $\nabla \cdot \underline{g} = -\nabla^2 T$

$$\Rightarrow \text{LHS} = \iint_{[0,1]^2} -g(x,y) \, dx \, dy$$

$$= - \int_0^1 \sin(2\pi x) \, dx \int_0^1 \sin(2\pi y) \, dy$$

$$= - \left(-\frac{1}{2\pi}\right) [\cos(2\pi x)]_0^1 - \left(-\frac{1}{2\pi}\right) [\cos(2\pi y)]_0^1$$

$$= 0$$

(One could have guessed this by considering symmetry about the line $x=0.5$)

$$Q2(a) \quad V = \frac{Q}{4\pi\epsilon r}$$

$$\begin{aligned} \underline{E} &= -\nabla V = -\nabla \left(\frac{Q}{4\pi\epsilon r} \right) \\ &= -\frac{\partial}{\partial r} \left(\frac{Q}{4\pi\epsilon r} \right) \underline{e}_r \\ &= \frac{Q}{4\pi\epsilon r^2} \underline{e}_r \end{aligned}$$

Solenoidal if $\nabla \cdot \underline{E} = 0$

$$\begin{aligned} \text{In spherical coordinates, } \nabla \cdot \underline{E} &= \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \frac{Q}{4\pi\epsilon r^2} \right) \\ &= 0 \quad \text{QED} \end{aligned}$$

$$(b) \quad \underline{E} = \frac{k}{r} \underline{e}_r \Rightarrow \text{force on } q_r = q_r k \frac{1}{r} \underline{e}_r$$

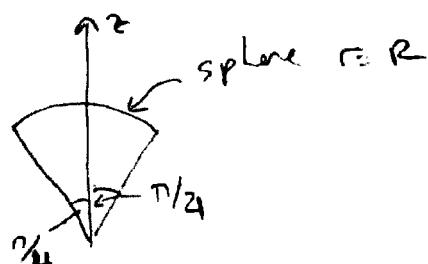
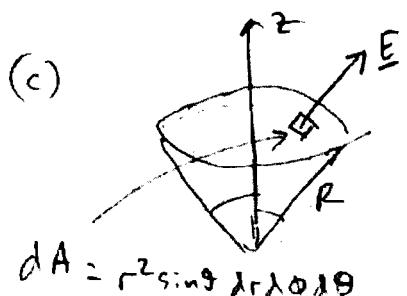
Work needed to move from R_1 to $2R_1$:

$$\int_{R_1}^{R_2} \frac{1}{r} q_r k dr = q_r k \ln \frac{R_2}{R_1}$$

$$\text{If } R_2 = 2R_1, \text{ work} = q_r k \ln 2$$

(Alternatively: work done = $[V(R_2) - V(R_1)] q_r$)

$$\begin{aligned} \text{Since } \frac{\partial V}{\partial r} &= \frac{k}{r} \Rightarrow V = k \ln r + C \\ \Rightarrow V(R_2) - V(R_1) &= k q_r \ln \frac{R_2}{R_1} \end{aligned}$$



Flux of \underline{E} out of surface:

- zero from conical surface since \underline{E} aligned with $\underline{\Gamma}$

- only through part of sphere at radius R

$$\text{Flux} = \int_0^{2\pi} \int_0^{\pi/4} R^2 (R^3 - r^2) \sin \theta d\theta d\phi$$

$$= (R^5 - R^4) \int_0^{2\pi} d\phi \int_0^{\pi/4} \sin \theta d\theta$$

$$= 2\pi R^4 (R-1) [\cos \theta]_0^{\pi/4}$$

$$= 2\pi (1 - \sqrt{2}/2) R^4 (R-1)$$

Divergence Theorem: $\iiint_V \nabla \cdot \underline{E} dv = \int_A \underline{E} \cdot d\underline{A}$

$$\text{LHS} = \nabla \cdot \underline{E} = \frac{1}{r^2} \frac{\partial}{\partial r} (r^5 - r^4) = 5r^2 - 4r$$

$$\int_0^{2\pi} \int_0^{\pi/4} \int_0^R (5r^2 - 4r) r^2 \sin \theta dr d\theta d\phi$$

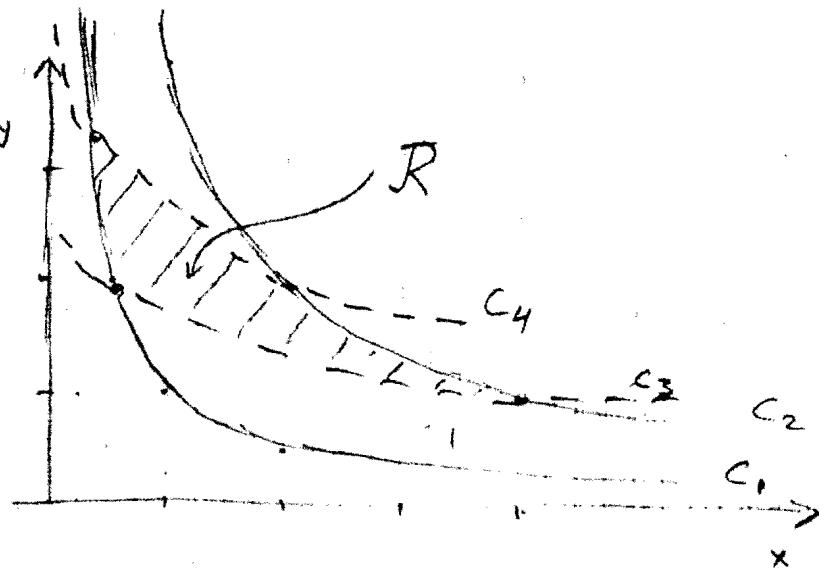
$$= -2\pi \cdot [\cos \theta]_0^{\pi/4} \int_0^R (5r^4 - 4r^3) dr$$

$$[\frac{5r^5}{5} - \frac{4r^4}{4}]_0^R$$

$$= 2\pi (1 - \sqrt{2}/2) (R^5 - R^4)$$

Same as before.

Q3 (a)



$$xy = 1 \rightarrow C_1$$

$$xy = 4 \rightarrow C_2$$

$$xy^3 = 4 \rightarrow C_3$$

$$xy^3 = 16 \rightarrow C_4$$

(b) Area: $\iint dxdy$ difficult to integrate.

as limits are complicated

Using the transformation $u = xy$

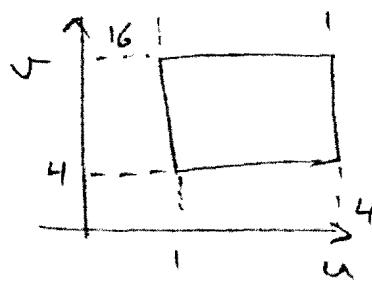
$$v = xy^3$$

area of integration becomes

$$\iint \frac{1}{J} du dv$$

$$J = \begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{vmatrix} = \begin{vmatrix} y & x \\ y^3 & 3xy^2 \end{vmatrix}$$

$$= 2v$$



$$\Rightarrow \text{area} = \int_4^{16} \int_1^4 \frac{1}{2r} du dr = \int_4^{16} \frac{1}{2r} dr \cdot [u]_1^4$$

$$= \frac{3}{2} \ln \frac{16}{4} = \frac{3}{2} \ln 4$$

(c) $\oint \underline{F} \cdot d\underline{L} = \iint_R \nabla \times \underline{F} dA$ by Stokes theorem

$$\nabla \times \underline{F} = \begin{vmatrix} i & j & k \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ \frac{1}{4}xy^4 & x^2y^3 & 0 \end{vmatrix} = (2xy^3 - xy^3) k \\ = xy^3 k$$

Integration over R of $\nabla \times \underline{F}$
is best done in $u-v$ space

$$\therefore \text{RHS} = \int_4^{16} \int_1^4 \frac{v}{2r} du dv = \frac{1}{2} \cdot 3 \cdot (16-4) \\ = 18$$

Q4

(a) Probability is $1 - \Phi\left(\frac{9-6}{\sqrt{2.25}}\right) = 1 - \Phi(2)$
 $= 0.0228$

(b) $P(\text{Disease}) = \frac{P(\text{positive test}) \cdot P(\text{positive diagnosis})}{P(\text{having disease} \& \text{test positive})}$
 $= \frac{0.6 \times 0.0228}{0.98} = 0.014$

(c) False positive rate = probability of having the test, but the patient does not have the disease. Call this $P(+, \bar{D})$

$$P(\text{positive test}) = P(+, \bar{D}) \cdot P(\text{negative diagnosis}) + P(\text{having disease} \& \text{test positive}) \cdot P(\text{Disease})$$
$$\Rightarrow P(+, \bar{D}) = \frac{0.0228 - 0.98 \times 0.6}{0.986} = 0.0091$$

(d) Poisson arrival of the test is $\lambda: 200 \times 0.0228 = 4.56$
 $\Rightarrow \text{Prob of } > 3 \text{ a day:}$

$$1 - \left(1 + 4.56 + \frac{4.56^2}{2} + \frac{4.56^3}{6} \right) e^{-4.56} = 0.668$$

5 (a)

$$\mathbf{A} = \begin{pmatrix} 4 & 2 & -2 & 0 \\ 2 & 3 & 1 & 2 \\ 2 & 2 & 0 & a \end{pmatrix} = 1 \begin{pmatrix} 4 & 2 & -2 & 0 \\ 2 & 1 & -1 & 0 \\ 2 & 1 & -1 & 0 \end{pmatrix} + 0.5 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 1 & 1 & a \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 1 & 1 & a \end{pmatrix} = 0 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 1 & 1 & 1 \end{pmatrix} + 0.5 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & a-1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 0 & a-1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & a-1 \end{pmatrix} = 0 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & a-1 \end{pmatrix} + 1 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 \\ 0.5 & 1 & 0 \\ 0.5 & 0.5 & 1 \end{pmatrix} \begin{pmatrix} 4 & 2 & -2 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 0 & 0 & a-1 \end{pmatrix} = \mathbf{L} \mathbf{U} \rightarrow \text{For rank 2, we need } a = 1. \quad [6]$$

(b) We have $\mathbf{Ax} = \mathbf{L}\mathbf{Ux} = \mathbf{Lc}$ where $\mathbf{c} = (c_1, c_2, c_3)^T = \mathbf{Ux}$. Thus,

$$\mathbf{Lc} = \begin{pmatrix} 1 & 0 & 0 \\ 0.5 & 1 & 0 \\ 0.5 & 0.5 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 1.25 \end{pmatrix} \rightarrow \begin{array}{l} c_1 = 1 \\ 0.5c_1 + c_2 = 2 \\ 0.5c_1 + 0.5c_2 + c_3 = 1.25 \end{array} \rightarrow \begin{array}{l} c_1 = 1 \\ c_2 = 1.5 \\ c_3 = 0 \end{array}$$

Hence, with $a = 1$ and writing $\mathbf{x} = (x, y, z, t)^T$, we have,

$$\begin{pmatrix} 4 & 2 & -2 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} 1 \\ 1.5 \\ 0 \end{pmatrix}$$

The free variables are z and t (because of the zeros in row 3, columns 3 and 4).

The particular solution \mathbf{x}_0 is found by setting $z = 0$ and $t = 0$. Thus,

$$\begin{aligned} 2y + 2z + 2t &= 1.5 \rightarrow y = 0.75 \\ 4x + 2y - 2z &= 1 \rightarrow 4x = 1 - 1.5 \rightarrow x = -0.125 \end{aligned}$$

Thus, $\mathbf{x}_0 = (-0.125, 0.75, 0, 0)^T$

[3]

The general solution is, $\mathbf{x} = \mathbf{x}_0 + \lambda \mathbf{u}_1 + \mu \mathbf{u}_2$ where \mathbf{u}_1 and \mathbf{u}_2 are in the null space and λ and μ are arbitrary scalar constants. To find \mathbf{u}_1 and \mathbf{u}_2 we solve $\mathbf{U}\mathbf{u} = \mathbf{0}$. Thus,

$$\begin{pmatrix} 4 & 2 & -2 & 0 \\ 0 & 2 & 2 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{array}{l} 2y + 2z + 2t = 0 \\ 4x + 2y - 2z = 0 \end{array}$$

Setting $z = 0$ and $t = 1$, we obtain $y = -1$ and $x = 0.5$. Thus, $\mathbf{u}_1 = (0.5, -1, 0, 1)^T$

Setting $z = 1$ and $t = 0$, we obtain $y = -1$ and $x = 1$. Thus, $\mathbf{u}_2 = (1, -1, 1, 0)^T$

[2]

[2]

Hence, the general solution is,

$$\mathbf{x} = \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} -0.125 \\ 0.75 \\ 0 \\ 0 \end{pmatrix} + \lambda \begin{pmatrix} 0.5 \\ -1 \\ 0 \\ 1 \end{pmatrix} + \mu \begin{pmatrix} 1 \\ -1 \\ 1 \\ 0 \end{pmatrix} \quad [1]$$

(c)

$$\begin{aligned} \mathbf{A} = \mathbf{LU} &= \begin{pmatrix} 1 \\ 0.5 \\ 0.5 \end{pmatrix} (4 \ 2 \ -2 \ 0) + \begin{pmatrix} 0 \\ 1 \\ 0.5 \end{pmatrix} (0 \ 2 \ 2 \ 2) + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} (0 \ 0 \ 0 \ 0) \\ &= \begin{pmatrix} 1 \\ 0.5 \\ 0.5 \end{pmatrix} (4 \ 2 \ -2 \ 0) + \begin{pmatrix} 0 \\ 1 \\ 0.5 \end{pmatrix} (0 \ 2 \ 2 \ 2) \\ &= \begin{pmatrix} 1 & 0 \\ 0.5 & 1 \\ 0.5 & 0.5 \end{pmatrix} \begin{pmatrix} 4 & 2 & -2 & 0 \\ 0 & 2 & 2 & 2 \end{pmatrix} = \mathbf{BC} \end{aligned} \quad [3]$$

(d) The column space of \mathbf{B} is in the column space of \mathbf{L} but does not span it. This is because it does not include the independent vector $(0, 0, 1)^T$.

The column space of \mathbf{C} is 2-dimensional whereas that of \mathbf{U} is 3-dimensional. Hence, strictly, the two column spaces are not comparable. However, because the final row of \mathbf{U} is composed of zeros, the column spaces are effectively identical.

[3]

6 (a) Let the eigenvalues and eigenvectors of the matrix \mathbf{B} be λ and \mathbf{u}_B respectively. Thus,

$$\mathbf{B}\mathbf{u}_B = \lambda\mathbf{u}_B$$

Premultiplying by the matrix \mathbf{X} gives,

$$\mathbf{X}\mathbf{B}\mathbf{u}_B = \lambda\mathbf{X}\mathbf{u}_B$$

From $\mathbf{A} = \mathbf{X}\mathbf{B}\mathbf{X}^{-1}$ (given) we have $\mathbf{AX} = \mathbf{XB}$. Thus,

$$\mathbf{A}(\mathbf{X}\mathbf{u}_B) = \lambda(\mathbf{X}\mathbf{u}_B)$$

Hence, λ are the eigenvalues and $\mathbf{X}\mathbf{u}_B$ are the eigenvectors of the matrix \mathbf{A} . [4]

(b) \mathbf{B} is an upper triangular matrix and so the eigenvalues are the elements on the diagonal,

$$\lambda_1 = 2, \quad \lambda_2 = 1, \quad \lambda_3 = 3. \quad [2]$$

The eigenvector corresponding to $\lambda_1 = 2$ is given by,

$$\begin{pmatrix} 2 - \lambda_1 & 0 & 1 \\ 0 & 1 - \lambda_1 & 1 \\ 0 & 0 & 3 - \lambda_1 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Thus, $e_3 = 0$ and $-e_2 + e_3 = 0 \rightarrow e_2 = 0$.

Hence, the normalised eigenvector is $\mathbf{u}_{B,1} = (1, 0, 0)^T$. [2]

The eigenvector corresponding to $\lambda_2 = 1$ is given by,

$$\begin{pmatrix} 2 - \lambda_2 & 0 & 1 \\ 0 & 1 - \lambda_2 & 1 \\ 0 & 0 & 3 - \lambda_2 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 2 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Thus, $e_3 = 0$ and $e_1 + e_3 = 0 \rightarrow e_1 = 0$.

Hence, the normalised eigenvector is $\mathbf{u}_{B,2} = (0, 1, 0)^T$. [2]

The eigenvector corresponding to $\lambda_3 = 3$ is given by,

$$\begin{pmatrix} 2 - \lambda_3 & 0 & 1 \\ 0 & 1 - \lambda_3 & 1 \\ 0 & 0 & 3 - \lambda_3 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} -1 & 0 & 1 \\ 0 & -2 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Thus, $e_1 = e_3$ and $-2e_2 + e_3 = 0 \rightarrow e_2 = e_3/2$.

Hence, the normalised eigenvector is $\mathbf{u}_{B,3} = (1, 1/2, 1)^T / (9/4)^{1/2} = (2/3, 1/3, 2/3)^T$. [2]

The eigenvalues of \mathbf{B} are distinct and so \mathbf{B} can be diagonalised. Writing $\mathbf{B} = \mathbf{U}_B \mathbf{D} \mathbf{U}_B^{-1}$ we construct \mathbf{U}_B from the eigenvectors and \mathbf{D} from the eigenvalues and obtain,

$$\mathbf{U}_B = \begin{pmatrix} 1 & 0 & 2/3 \\ 0 & 1 & 1/3 \\ 0 & 0 & 2/3 \end{pmatrix}, \quad \mathbf{D} = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

Note that \mathbf{U}_B is not an orthogonal matrix (\mathbf{B} is not symmetric) and hence $\mathbf{U}_B^{-1} \neq \mathbf{U}_B^T$. To find \mathbf{U}_B^{-1} we can use the Gauss-Jordan elimination method:

$$\begin{pmatrix} 1 & 0 & 2/3 & 1 & 0 & 0 \\ 0 & 1 & 1/3 & 0 & 1 & 0 \\ 0 & 0 & 2/3 & 0 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 2/3 & 1 & 0 & 0 \\ 0 & 1 & 1/3 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 3/2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & 1 & 0 & -1 \\ 0 & 1 & 0 & 0 & 1 & -1/2 \\ 0 & 0 & 1 & 0 & 0 & 3/2 \end{pmatrix}$$

$$\mathbf{U}_B^{-1} = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1/2 \\ 0 & 0 & 3/2 \end{pmatrix}$$

Thus,

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 2/3 \\ 0 & 1 & 1/3 \\ 0 & 0 & 2/3 \end{pmatrix} \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1/2 \\ 0 & 0 & 3/2 \end{pmatrix} \quad [4]$$

(c) As shown in Part (a) the eigenvectors \mathbf{u}_A of \mathbf{A} are the column vectors $\mathbf{X}\mathbf{u}_B$. Thus,

$$\text{For } \lambda_1 = 2, \quad \mathbf{u}_{A,1} = \mathbf{X}\mathbf{u}_{B,1} = \mathbf{X} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \mathbf{x}_1$$

$$\text{For } \lambda_2 = 1, \quad \mathbf{u}_{A,2} = \mathbf{X}\mathbf{u}_{B,2} = \mathbf{X} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \mathbf{x}_2$$

$$\text{For } \lambda_3 = 3, \quad \mathbf{u}_{A,3} = \mathbf{X}\mathbf{u}_{B,3} = \mathbf{X} \begin{pmatrix} 2/3 \\ 1/3 \\ 2/3 \end{pmatrix} = \frac{2}{3}\mathbf{x}_1 + \frac{1}{3}\mathbf{x}_2 + \frac{2}{3}\mathbf{x}_3$$

where \mathbf{x}_1 , \mathbf{x}_2 and \mathbf{x}_3 are the column vectors of \mathbf{X} .

[4]