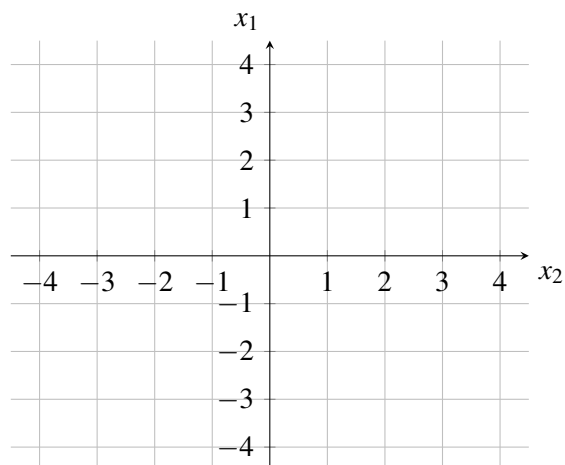


1. Projections

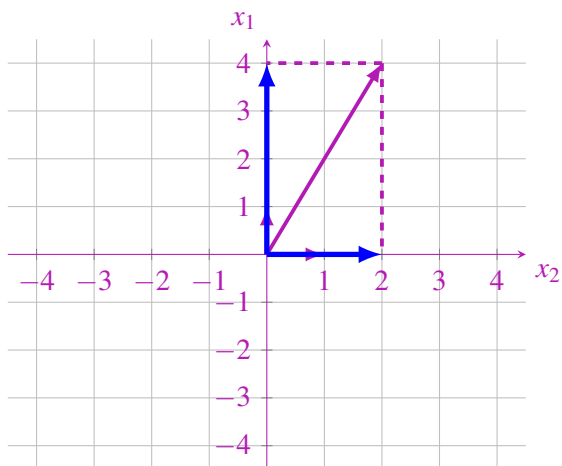
Learning Goal: The goal of this problem is to understand the properties of projection.

Relevant Notes: [Note 23](#) walks through mathematical derivations for projection.



- (a) Consider the vector $\vec{x} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$. Draw it on the graph provided. Also draw the vector $\vec{y}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $\vec{y}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$. Now, find the projections of \vec{x} on \vec{y}_1 and \vec{y}_2 geometrically. Compare with mathematical calculations.

Answer:



The projection of vector \vec{b} on vector \vec{a} is given by:

$$\text{proj}_{\vec{a}}\vec{b} = \frac{\langle \vec{b}, \vec{a} \rangle}{\|\vec{a}\|^2} \vec{a} \quad (1)$$

Now we have:

$$\langle \vec{x}, \vec{y}_1 \rangle = 2 \cdot 1 + 4 \cdot 0 = 2$$

$$\langle \vec{x}, \vec{y}_2 \rangle = 2 \cdot 0 + 4 \cdot 1 = 4$$

$$\|\vec{y}_1\| = \sqrt{1+0} = 1$$

$$\|\vec{y}_2\| = \sqrt{0+1} = 1$$

Hence projection of vector \vec{x} on vector \vec{y}_1 is

$$\text{proj}_{\vec{y}_1}\vec{x} = \frac{\langle \vec{x}, \vec{y}_1 \rangle}{\|\vec{y}_1\|^2} \vec{y}_1 = \frac{2}{1} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix} \quad (2)$$

$$\text{proj}_{\vec{y}_2}\vec{x} = \frac{\langle \vec{x}, \vec{y}_2 \rangle}{\|\vec{y}_2\|^2} \vec{y}_2 = \frac{4}{1} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \end{bmatrix} \quad (3)$$

- (b) Calculate the projection of $\vec{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ on $\vec{y} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$. Is it the same as the projection of \vec{y} on \vec{x} ?

Answer: Hence projection of vector \vec{x} on vector \vec{y} is

$$\text{proj}_{\vec{y}}\vec{x} = \frac{\langle \vec{x}, \vec{y} \rangle}{\|\vec{y}\|^2} \vec{y} = \frac{1 \cdot 3 + 1 \cdot 4}{3^2 + 4^2} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \frac{7}{25} \begin{bmatrix} 3 \\ 4 \end{bmatrix}. \quad (4)$$

Now the projection of vector \vec{y} on vector \vec{x} is given by

$$\text{proj}_{\vec{x}}\vec{y} = \frac{\langle \vec{y}, \vec{x} \rangle}{\|\vec{x}\|^2} \vec{x} = \frac{3 \cdot 1 + 4 \cdot 1}{1^2 + 1^2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{7}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3.5 \\ 3.5 \end{bmatrix}, \quad (5)$$

which is not the same as $\text{proj}_{\vec{x}}\vec{y}$.

- (c) Now consider the vectors $\vec{x} = \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}$, $\vec{y}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$ and $\vec{y}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$. Now, find the projections of \vec{x} on \vec{y}_1 and \vec{y}_2 . Also find the projection of \vec{x} on $\text{span}\{\vec{y}_1, \vec{y}_2\}$. Is $\text{proj}_{\vec{y}_1}\vec{x} + \text{proj}_{\vec{y}_2}\vec{x}$ equal to $\text{proj}_{\text{span}\{\vec{y}_1, \vec{y}_2\}}\vec{x}$? Explain your answer.

Answer: The projection of vector \vec{x} on vector \vec{y}_1 is

$$\text{proj}_{\vec{y}_1}\vec{x} = \frac{\langle \vec{x}, \vec{y}_1 \rangle}{\|\vec{y}_1\|^2} \vec{y}_1 = \frac{1 \cdot 1 + 2 \cdot 1 + 4 \cdot 0}{1^2 + 1^2 + 0^2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \frac{3}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1.5 \\ 1.5 \\ 0 \end{bmatrix}. \quad (6)$$

Similarly projection of vector \vec{x} on vector \vec{y}_2 is

$$\text{proj}_{\vec{y}_2}\vec{x} = \frac{\langle \vec{x}, \vec{y}_2 \rangle}{\|\vec{y}_2\|^2} \vec{y}_2 = \frac{1 \cdot 0 + 2 \cdot 0 + 4 \cdot 1}{0^2 + 0^2 + 1^2} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \frac{4}{1} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \\ 4 \end{bmatrix}. \quad (7)$$

We can use the least squares formula to find the projection of a vector on a subspace. The the projection of \vec{x} on $\text{span}\{\vec{y}_1, \vec{y}_2\}$ is the same as projection of \vec{x} on $\text{col}\{\mathbf{A}\}$, where matrix \mathbf{A} has the columns \vec{y}_1 and \vec{y}_2 .

$$\text{proj}_{\text{span}\{\vec{y}_1, \vec{y}_2\}} \vec{x} = \mathbf{A} \hat{\vec{\alpha}} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{x}, \quad (8)$$

where $\hat{\vec{\alpha}}$ is the least squares solution to $\mathbf{A} \vec{\alpha} = \vec{x}$.

Now we can calculate:

$$\mathbf{A}^T \mathbf{A} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

$$(\mathbf{A}^T \mathbf{A})^{-1} = \begin{bmatrix} 0.5 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A}^T \vec{x} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

So we can calculate

$$\text{proj}_{\text{span}\{\vec{y}_1, \vec{y}_2\}} \vec{x} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{x} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0.5 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1.5 \\ 4 \end{bmatrix} = \begin{bmatrix} 1.5 \\ 1.5 \\ 4 \end{bmatrix}$$

Now summing up the projections on \vec{y}_1 and \vec{y}_2 , we have

$$\text{proj}_{\vec{y}_1} \vec{x} + \text{proj}_{\vec{y}_2} \vec{x} = \begin{bmatrix} 1.5 \\ 1.5 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix} = \begin{bmatrix} 1.5 \\ 1.5 \\ 4 \end{bmatrix} = \text{proj}_{\text{span}\{\vec{y}_1, \vec{y}_2\}} \vec{x}$$

. Here $\text{proj}_{\vec{y}_1} \vec{x} + \text{proj}_{\vec{y}_2} \vec{x}$ is equal to $\text{proj}_{\text{span}\{\vec{y}_1, \vec{y}_2\}} \vec{x}$ since \vec{y}_1 and \vec{y}_2 are orthogonal, i.e. $\langle \vec{y}_1, \vec{y}_2 \rangle = 0$.

- (d) Find the expression for projection of $\vec{b} = \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}$ on the columnspace of matrix $\mathbf{A} = \begin{bmatrix} | & | \\ \vec{a}_1 & \vec{a}_2 \\ | & | \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 0 \end{bmatrix}$.

Is $\text{proj}_{\vec{a}_1} \vec{b} + \text{proj}_{\vec{a}_2} \vec{b}$ equal to $\text{proj}_{\text{Col}\{\mathbf{A}\}} \vec{b}$? (No need to do the calculations.)

If we set up a system of linear equations $\mathbf{A} \vec{x} = \vec{b}$, will there be a unique solution? (No need to solve the system.)

Answer: We use the least squares formula again to find the projection of a vector on a subspace. The the projection of \vec{b} on $\text{Col}\{\mathbf{A}\}$ is

$$\text{proj}_{\text{Col}\{\mathbf{A}\}} \vec{b} = \mathbf{A} \hat{\vec{x}} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{b}, \quad (9)$$

where $\hat{\vec{x}}$ is the least squares solution to $\mathbf{A} \vec{x} = \vec{b}$.

Now we can calculate:

$$\mathbf{A}^T \mathbf{A} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 3 & 1 \\ 1 & 1 \end{bmatrix}$$

$$(\mathbf{A}^T \mathbf{A})^{-1} = \begin{bmatrix} 0.5 & -0.5 \\ -0.5 & 1.5 \end{bmatrix}$$

$$\mathbf{A}^T \vec{x} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 7 \\ 1 \end{bmatrix}$$

So we can calculate

$$\text{proj}_{\text{Col}\{\mathbf{A}\}} \vec{b} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{b} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0.5 & -0.5 \\ -0.5 & 1.5 \end{bmatrix} \begin{bmatrix} 7 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0.5 & -0.5 \\ 0.5 & -0.5 \end{bmatrix} \begin{bmatrix} 7 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 3 \end{bmatrix}.$$

The columns of \mathbf{A} are not orthogonal, so $\text{proj}_{\vec{a}_1} \vec{b} + \text{proj}_{\vec{a}_2} \vec{b}$ is not equal to $\text{proj}_{\text{Col}\{\mathbf{A}\}} \vec{b}$.

There won't be a unique solution to $\mathbf{A}\vec{x} = \vec{b}$, since $\text{proj}_{\text{Col}\{\mathbf{A}\}} \vec{b} \neq \vec{b}$. This means that \vec{b} is not in the columnspace of \mathbf{A} , so the system is inconsistent.

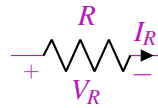
2. And You Thought You Could Ignore Circuits Until Dead Week

Learning Goal: The objective of this problem is to practice solving a noisy system using least squares method.

Relevant Notes: [Note 23](#) covers the details of least squares method.

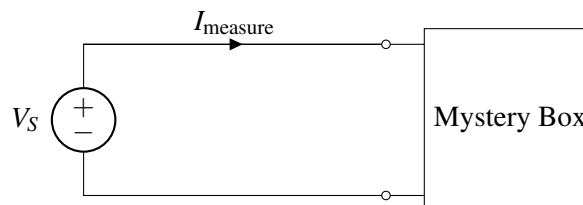
- (a) Write Ohm's Law for a resistor.

Answer: For the resistor



$$V_R = I_R R$$

- (b) You're given the following test setup and told to find R_{eq} between the two terminals of the mystery box. What is R_{eq} of the mystery box between the two terminals in terms of V_S and $I_{measure}$?



Answer:

$$R_{eq} = \frac{V_S}{I_{measure}}$$

$$R_{eq} = \frac{V_S}{I_{measure}}$$

- (c) You think you've figured out how to find R_{eq} ! You've taken the following measurements:

Measurement #	V_S	$I_{measure}$
1	2V	1A
2	4V	2A
3	6V	2A
3	8V	4A

Using the information above, formulate a least squares problem whose answer provides an estimate of R_{eq} .

Answer: According to Ohm's Law, $V_S = I_{measure}R_{eq}$. We have to calculate the least squares solution

\hat{R}_{eq} . We are estimating the resistance, so \hat{R}_{eq} corresponds to \hat{x} in the equation $\mathbf{A}\hat{x} = \vec{b}$, where $\vec{i} = \begin{bmatrix} 1 \\ 2 \\ 2 \\ 4 \end{bmatrix}$

and $\vec{v} = \begin{bmatrix} 2 \\ 4 \\ 6 \\ 8 \end{bmatrix}$ correspond to \mathbf{A} and \vec{b} , respectively:

$$\begin{aligned}\hat{x} &= (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{b} \\ \hat{R}_{eq} &= (\vec{i}^T \vec{i})^{-1} \vec{i}^T \vec{v} \\ &= ([1 \ 2 \ 2 \ 4] \begin{bmatrix} 1 \\ 2 \\ 2 \\ 4 \end{bmatrix})^{-1} [1 \ 2 \ 2 \ 4] \begin{bmatrix} 2 \\ 4 \\ 6 \\ 8 \end{bmatrix} \\ &= (1 \cdot 1 + 2 \cdot 2 + 2 \cdot 2 + 4 \cdot 4)^{-1} (1 \cdot 2 + 2 \cdot 4 + 2 \cdot 6 + 4 \cdot 8) \\ &= (25)^{-1} (54) \\ &= \frac{54}{25} \Omega = 2.16 \Omega\end{aligned}$$

(d) Find the least square error vector $\|\vec{e}\|$.

Answer: The minimum error is given by

$$\vec{e} = \mathbf{A}\hat{x} - \vec{b} = \vec{i}\hat{R}_{eq} - \vec{v} = \begin{bmatrix} 1 \\ 2 \\ 2 \\ 4 \end{bmatrix} \times 2.16 - \begin{bmatrix} 2 \\ 4 \\ 6 \\ 8 \end{bmatrix} = \begin{bmatrix} 2.16 \\ 4.32 \\ 4.32 \\ 8.64 \end{bmatrix} - \begin{bmatrix} 2 \\ 4 \\ 6 \\ 8 \end{bmatrix} = \begin{bmatrix} 0.16 \\ 0.32 \\ -1.68 \\ 0.64 \end{bmatrix}$$

3. Least Squares Fitting

Learning Goal: The objective of this problem is to set up a least squares problem for coefficients of non-linear equations.

Relevant Notes: [Note 23](#) covers the details of least squares method.

In an upward career move, you join the starship USS Enterprise as a data scientist. One morning the Chief Science Officer, Mr. Spock, hands you some data for the position (y) of a newly discovered particle at different times (t). The data has three points and **contains some noise**:

$$(t = 0, y = 0.5), \quad (t = 1, y = 3), \quad (t = 2, y = 18.5)$$

Your research shows that the path of the particle is represented by the function:

$$y = e^{w_1 + w_2 t} \tag{10}$$

You decide to fit the collected data to the function in Equation (??) using the Least Squares method.

series = qn You need to find the coefficients w_1 and w_2 that *minimize the squared error* between the fitted curve and the collected data points. So you set up a system of linear equations, $\mathbf{A}\hat{\alpha} \approx \vec{b}$ in order to find the approximate value of $\hat{\alpha} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$. What are the values of \mathbf{A} and \vec{b} ?

Answer: For $t = 0$, we have:

$$\begin{aligned} 0.5 &= e^{w_1+w_2(0)} \\ \implies \ln(0.5) &= w_1 + 0(w_2) \end{aligned}$$

Similarly for $t = 1$, and $t = 2$, we have:

$$\begin{aligned} \ln(3) &= w_1 + w_2 \\ \ln(18.5) &= w_1 + 2w_2 \end{aligned}$$

Hence we can write the following system of linear equations:

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \approx \begin{bmatrix} \ln(0.5) \\ \ln(3) \\ \ln(18.5) \end{bmatrix}$$

series = qn Mr. Spock thinks one of the data points is wrong and asks you to redo the fit with only two data points. What will happen to the norm of the error, $\|\vec{e}\| = \|\vec{b} - \mathbf{A}\hat{\alpha}\|$?

Answer: The linear system now has two unknowns (w_1, w_2) and two linearly-independent constraints (the two data points), so there will be an exact fit to the data: the norm of error $\|\vec{e}\| = \|\vec{b} - \mathbf{A}\hat{\alpha}\|$ will be 0. This is probably too good to be true!

series = qn Your colleague tries to repeat your fitting process with the same four data points in part (a), but they misread the equation relating t and y , i.e. they use the following function (which is **different than part (a)**):

$$y = e^{w_1 t + w_2 t^2} \tag{11}$$

Your colleague tries to find w_1 and w_2 by setting up a system of equations $\mathbf{A}\hat{\alpha} \approx \vec{b}$ and utilizing the equation:

$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \hat{\alpha} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{b}. \tag{12}$$

What will happen when your colleague tries to solve the above equation?

Answer: The new system of linear equations can be written as:

$$\begin{bmatrix} 0 & 0 \\ 1 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \approx \begin{bmatrix} \ln(0.5) \\ \ln(3) \\ \ln(18.5) \end{bmatrix}$$

Notice that the first and second columns of \mathbf{A} are the same. Since \mathbf{A} has linearly dependent columns, $\mathbf{A}^T \mathbf{A}$ will not be invertible, i.e. the equation for $\hat{\alpha}$ will not work.

4. Besto Pesto (Final Exam, Fall 2018) [PRACTICE]

Your TA Laura is struggling to keep her basil plant alive! She needs your help to determine how much water and sunlight her plant needs.

Let $x_h[k]$ be the plant's height on day k and $x_\ell[k]$ be the number of leaves on the plant on day k . The vector $\vec{x}[k] = \begin{bmatrix} x_h[k] \\ x_\ell[k] \end{bmatrix}$ defines the state of the plant. The evolution of the basil plant from one day to the next is defined by the **approximate** mathematical model:

$$\vec{x}[k+1] = \mathbf{A}\vec{x}[k] = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_h[k] \\ x_\ell[k] \end{bmatrix}. \quad (13)$$

- (a) Our first goal is to estimate the elements of state transition matrix, \mathbf{A} : $a_{11}, a_{12}, a_{21}, a_{22}$. To do this we count the leaves and measure the height for the first N time steps, i.e. we know $\{\vec{x}[0], \vec{x}[1], \dots, \vec{x}[N]\}$.

Setup a least squares problem to estimate $\vec{a} = \begin{bmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{bmatrix}$:

$$\hat{\vec{a}} = \underset{\vec{a}}{\operatorname{argmin}} \|\mathbf{M}\vec{a} - \vec{b}\|^2. \quad (14)$$

Write the matrix, \mathbf{M} , and vector, \vec{b} , that would be used in the above least squares problem for $N = 3$.

Answer: We can simplify the equations:

$$\begin{aligned} \begin{bmatrix} x_h[k+1] \\ x_\ell[k+1] \end{bmatrix} &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_h[k] \\ x_\ell[k] \end{bmatrix} \\ \implies x_h[k+1] &= a_{11}x_h[k] + a_{12}x_\ell[k] \\ x_\ell[k+1] &= a_{21}x_h[k] + a_{22}x_\ell[k] \end{aligned}$$

For $N = 3$, we have:

$$\begin{aligned} x_h[1] &= a_{11}x_h[0] + a_{12}x_\ell[0] \\ x_\ell[1] &= a_{21}x_h[0] + a_{22}x_\ell[0] \\ x_h[2] &= a_{11}x_h[1] + a_{12}x_\ell[1] \\ x_\ell[2] &= a_{21}x_h[1] + a_{22}x_\ell[1] \\ x_h[3] &= a_{11}x_h[2] + a_{12}x_\ell[2] \\ x_\ell[3] &= a_{21}x_h[2] + a_{22}x_\ell[2] \end{aligned}$$

$$\mathbf{M} = \begin{bmatrix} x_h[0] & x_\ell[0] & 0 & 0 \\ 0 & 0 & x_h[0] & x_\ell[0] \\ x_h[1] & x_\ell[1] & 0 & 0 \\ 0 & 0 & x_h[1] & x_\ell[1] \\ x_h[2] & x_\ell[2] & 0 & 0 \\ 0 & 0 & x_h[2] & x_\ell[2] \end{bmatrix} \quad (15)$$

$$\vec{b} = \begin{bmatrix} x_h[1] \\ x_\ell[1] \\ x_h[2] \\ x_\ell[2] \\ x_h[3] \\ x_\ell[3] \end{bmatrix} \quad (16)$$