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1.

$$(a) \begin{pmatrix} 5 & 3 & -2 \\ 4 & -2 & 3 \\ 1 & 1 & -5 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 4 \\ 12 \\ 24 \end{pmatrix} \#$$

(b)

>> A = [5, 3, -2;  
4, -2, 3;  
1, 1, -5;]

A =

$$\begin{matrix} 5 & 3 & -2 \\ 4 & -2 & 3 \\ 1 & 1 & -5 \end{matrix}$$

>> b = [4; 12; 24];  
>> w = A \ b

w =

$$\begin{matrix} 3.3043 \\ -8.0000 \\ -5.7391 \end{matrix}$$

#

$$(c) \begin{pmatrix} 5 & 3 & -2 & 4 \\ 4 & -2 & 3 & 12 \\ 1 & 1 & -5 & 24 \end{pmatrix} \xrightarrow[\text{R}_1 \text{ and } \text{R}_3]{\text{switch}} \begin{pmatrix} 1 & 1 & -5 & 24 \\ 4 & -2 & 3 & 12 \\ 5 & 3 & -2 & 4 \end{pmatrix}$$

$$\xrightarrow[-4R_1+R_2]{-5R_1+R_3} \begin{pmatrix} 1 & 1 & -5 & 24 \\ 0 & -6 & 23 & -84 \\ 0 & -2 & 23 & -116 \end{pmatrix} \xrightarrow{-\frac{1}{6}R_2} \begin{pmatrix} 1 & 1 & -5 & 24 \\ 0 & 1 & -\frac{23}{6} & 14 \\ 0 & -2 & 23 & -116 \end{pmatrix}$$

$$\xrightarrow[2R_2+R_3]{-\frac{3}{4}R_3} \begin{pmatrix} 1 & 1 & -5 & 24 \\ 0 & 1 & -\frac{23}{6} & 14 \\ 0 & 0 & \frac{46}{3} & -88 \end{pmatrix} \xrightarrow{\frac{3}{46}R_3} \begin{pmatrix} 1 & 1 & -5 & 24 \\ 0 & 1 & -\frac{23}{6} & 14 \\ 0 & 0 & 1 & -\frac{132}{23} \end{pmatrix}$$

$$\Rightarrow z = -\frac{132}{23} \quad y - \frac{23}{6}z = 14 \Rightarrow y = -8 \quad x + y - 5z = 24 \\ x = \frac{76}{23}$$

$$W = \begin{bmatrix} \frac{\gamma b}{23} & -8 & -\frac{132}{23} \end{bmatrix}^T \#$$

(d)

$$\left( \begin{array}{ccc|ccc} 5 & 3 & -2 & 1 & 0 & 0 \\ 4 & -2 & 3 & 0 & 1 & 0 \\ 1 & 1 & -5 & 0 & 0 & 1 \end{array} \right) \xrightarrow[\text{R}_2 \text{ and } R_1]{\text{Switch}} \left( \begin{array}{ccc|ccc} 1 & 1 & 5 & 0 & 0 & 1 \\ 4 & -2 & 3 & 0 & 1 & 0 \\ 5 & 3 & -2 & 1 & 0 & 0 \end{array} \right)$$

$$\begin{aligned} \xrightarrow[-4R_1 + R_2]{-5R_1 + R_3} & \left( \begin{array}{ccc|ccc} 1 & 1 & -5 & 0 & 0 & 1 \\ 0 & -6 & 23 & 0 & 1 & -4 \\ 0 & 1 & 23 & 1 & 0 & -5 \end{array} \right) \xrightarrow{-\frac{1}{6}R_2} \left( \begin{array}{ccc|ccc} 1 & 1 & -5 & 0 & 0 & 1 \\ 0 & 1 & -\frac{23}{6} & 0 & -\frac{1}{6} & \frac{1}{2} \\ 0 & 2 & 23 & 1 & 0 & -5 \end{array} \right) \\ \xrightarrow[-R_3 + R_1]{-R_2 + R_1} & \left( \begin{array}{ccc|ccc} 1 & 0 & -\frac{7}{6} & 0 & 0 & 1 \\ 0 & 1 & -\frac{23}{6} & 0 & -\frac{1}{6} & \frac{1}{2} \\ 0 & 2 & 23 & 1 & 0 & -5 \end{array} \right) \xrightarrow{-2R_2 - R_3} \left( \begin{array}{ccc|ccc} 1 & 0 & -\frac{7}{6} & 0 & 0 & 1 \\ 0 & 1 & -\frac{23}{6} & 0 & -\frac{1}{6} & \frac{1}{2} \\ 0 & 0 & \frac{145}{6} & 1 & 0 & -5 \end{array} \right) \\ \xrightarrow[\frac{3}{4b}R_3]{-R_1} & \left( \begin{array}{ccc|ccc} 1 & 0 & -\frac{7}{6} & 0 & 0 & 1 \\ 0 & 1 & -\frac{23}{6} & 0 & -\frac{1}{6} & \frac{1}{2} \\ 0 & 0 & \frac{145}{6} & 1 & 0 & -5 \end{array} \right) \xrightarrow{-\frac{b}{7}R_2 - R_1} \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 & \frac{7}{2} \\ 0 & 0 & \frac{145}{6} & 1 & 0 & -5 \end{array} \right) \xrightarrow{-\frac{23}{6}R_3 - R_2} \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 & \frac{7}{2} \\ 0 & 0 & 1 & 0 & 0 & -\frac{145}{6} \end{array} \right) \end{aligned}$$

$$A^{-1} = \begin{pmatrix} \frac{\gamma b}{23} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{4} \end{pmatrix}$$

$$W = A^{-1} \cdot b = \begin{pmatrix} \frac{\gamma b}{23} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{4} \end{pmatrix} \begin{pmatrix} 4 \\ -2 \\ 24 \end{pmatrix} = \begin{pmatrix} \frac{\gamma b}{23} \\ -8 \\ -\frac{132}{23} \end{pmatrix} \#$$

$$\geq \text{Proof: } u = \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} \quad v = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$$

$$A = u \cdot v^T = \begin{pmatrix} u_1 v_1 & u_1 v_2 & u_1 v_3 \\ u_2 v_1 & u_2 v_2 & u_2 v_3 \\ u_3 v_1 & u_3 v_2 & u_3 v_3 \end{pmatrix}$$

$$\det(A) = u_1 v_1 (u_2 v_2 \cdot u_3 v_3 - u_2 v_3 \cdot u_3 v_2) - u_1 v_2 (u_2 v_1 \cdot u_3 v_3 - u_2 v_3 \cdot u_3 v_1) + u_1 v_3 (u_2 v_1 \cdot u_3 v_2 - u_2 v_2 \cdot u_3 v_1)$$

$\therefore u$  and  $v$  is linear relation  $\therefore u = \lambda v$  or  $v = \lambda u$ .

So we can get  $\det(A) = 0$  #

3.

(a)

$$\|u+v\|^2 = (u+v)^T \cdot (u+v) = (u^T + v^T)(u+v) = \|u\|^2 + u^T v + u v^T + \|v\|^2$$

By vectors dot product, we can get  $u^T v = u v^T$

$$\therefore \|u+v\|^2 = \|u\|^2 + 2u^T v + \|v\|^2 \#$$

(b)

$$(u+v)^T (u-v) = (u^T + v^T)(u-v) = \|u\|^2 - u^T v + v^T u - \|v\|^2$$

By (a), we can get  $u^T v = u v^T$ .

$$\therefore (u+v)^T (u-v) = \|u\|^2 - \|v\|^2 \#$$

$$(c) u^T v = \|u\| \cdot \|v\| \cdot \cos \theta \Rightarrow |u^T v| = |\|u\| \cdot \|v\| \cdot \cos \theta|$$

$$\Rightarrow |\|u\| \cdot \|v\|| \geq |\|u\| \cdot \|v\| \cdot \cos \theta|$$

So we can get  $|U^T V| \leq \|U\| \cdot \|V\|$  #

$$(d) ① \|U+V\|^2 = \|U\|^2 + 2U^T V + \|V\|^2$$

$$\text{by (c) we can get } ① \leq \|U\|^2 + 2\|U\|\cdot\|V\| + \|V\|^2 = (\|U\| + \|V\|)^2$$

$$\text{So that } \|U+V\| \leq \|U\| + \|V\| \#$$

$$(e) ② \|U-V\|^2 = \|U\|^2 - 2U^T V + \|V\|^2$$

$$③ \|U-V\|^2 = \|U\|^2 - 2U^T V + \|V\|^2$$

$$④ + ③ = 2(\|U\|^2 + \|V\|^2) \#$$

(f)  $\because U^T V = 0 \therefore U$  and  $V$  is vertical

$$\|U+V\|^2 = \|U\|^2 + 2U^T V + \|V\|^2 \quad \because U^T V = 0 \quad \therefore \|U+V\|^2 = \|U\|^2 + \|V\|^2 \#$$

4.

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix}$$

$$AB = BA = \begin{pmatrix} 6 & 6 & 6 \\ 6 & 6 & 6 \\ 6 & 6 & 6 \end{pmatrix} \#$$

5.

$$\begin{aligned} (\text{std}(x))^2 &= \frac{\|x - C\left(\frac{x}{n}\right)\|_2^2}{n} \\ &= \frac{x^T x - 2x^T \left(C\left(\frac{x}{n}\right)\right) I + \left(C\left(\frac{x}{n}\right)\right)^T \left(C\left(\frac{x}{n}\right)\right)}{n} \end{aligned}$$

$$= \frac{x^T x}{n} - \frac{2\left(\frac{x}{n}\right)^T x}{n} + \frac{n \cdot \left(\frac{x}{n}\right)^2}{n}$$

$$\downarrow \text{rms}(x)^2 - \left(\frac{x}{n}\right)^2 \downarrow \text{avg}(x)^2$$

when we expand  $(\text{std}(x))^2$  and break it down into three terms with  $n$  as the denominator, we find that the first term is  $\text{rms}(x)^2$ , the second term can be written as  $\frac{n(x^T x)^2}{n}$  through matrix multiplication transformations and after expanding the product of the third term, we end up with  $n$  terms.

Therefore, when we combine the second and third terms we obtain exactly  $\text{avg}(x)^2$

$$\text{std}(x)^2 = \text{rms}(x)^2 - \text{avg}(x)^2$$

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6.

(a)

$$\tilde{a} = \begin{pmatrix} x_1 - \mu \\ x_2 - \mu \\ \vdots \\ x_T - \mu \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \tilde{b} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ x_1 - \mu \\ \vdots \\ x_T - \mu \end{pmatrix} \quad R(\tilde{v}) = \frac{\tilde{a}^\top \tilde{b}}{\|\tilde{a}\| \cdot \|\tilde{b}\|}$$

$$\text{When } \tilde{v} = 0, \quad \tilde{a} = \begin{pmatrix} x_1 - \mu \\ \vdots \\ x_T - \mu \end{pmatrix} \quad \tilde{b} = \begin{pmatrix} x_1 - \mu \\ \vdots \\ x_T - \mu \end{pmatrix}$$

$$R(0) = \frac{\tilde{a}^\top \tilde{b}}{\|\tilde{a}\| \cdot \|\tilde{b}\|} = 1$$

$$\text{When } \tilde{v} \geq T, \quad \tilde{a} = \begin{pmatrix} x_1 - \mu \\ x_2 - \mu \\ \vdots \\ x_T - \mu \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \tilde{b} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ x_1 - \mu \\ \vdots \\ x_T - \mu \end{pmatrix}$$

$$R(\tilde{v}) = \frac{0}{\|\tilde{a}\| \cdot \|\tilde{b}\|} = 0 \quad \#$$

$$(b) \quad \mu = (\tilde{v}^\top x)/T \Rightarrow \text{avg}(x)$$

$$z = \frac{1}{\sqrt{T} \|\tilde{v}\|} (x - \text{avg}(x))^\top$$

$$\begin{aligned} R(\tilde{v}) &= \frac{\tilde{a}}{\|\tilde{a}\|} \cdot \frac{\tilde{b}}{\|\tilde{b}\|} \\ &= \frac{(x - \mu, \tilde{v}_c)}{\sqrt{T} \|\tilde{a}\|} \cdot \frac{(\tilde{v}_c, x - \mu)}{\sqrt{T} \|\tilde{b}\|} \end{aligned}$$

$$= \frac{1}{T} (z, \tilde{v}_c)^\top \cdot (\tilde{v}_c, z)$$

$$\therefore \text{we can get } R(\tilde{v}) = \frac{1}{T} \sum_{t=1}^{T-\tilde{v}} z_t \cdot z_{t+\tilde{v}}. \quad \#$$

$$(c) \quad \because \text{mean is zero, norm is } \sqrt{T} \quad \therefore \frac{(x - \mu, \tilde{v}_c)}{\sqrt{T} \|\tilde{a}\|} = \frac{(x, \tilde{v}_c)}{T}$$

$$R(\tilde{v}) = \frac{1}{T} \sum_{t=1}^{T-\tilde{v}} x_t \cdot x_{t+\tilde{v}} = \frac{T-\tilde{v}}{T} (-1)^\tilde{v} \Rightarrow \frac{T-\tilde{v}}{T} \quad \tilde{v} \text{ is even}$$

$$\frac{\tilde{v}-T}{T} \quad \tilde{v} \text{ is odd} \quad \#$$

(d) We can assume  $R(7)$  is the most busy day all the week.  
So  $R(7)$  is likely  $R(7)$  but can not exceed it. #

7.

```
function result = matrix_multiply(A, B, flag)
    [rowsA, colsA] = size(A);
    [rowsB, colsB] = size(B);

    if colsA ~= rowsB
        error('Matrix dimensions are not same.');
    end

    result = zeros(rowsA, colsB);
    |

    if flag == 1
        disp('Dot product method')
        for i = 1:rowsA
            for j = 1:colsB
                result(i, j) = dot(A(i, :), B(:, j));
            end
        end
    elseif flag == 2
        disp('Linear combination of columns method')
        for i = 1:colsB
            result(:, i) = A * B(:, i);
        end
    elseif flag == 3
        disp('Linear combination of rows method')
        for i = 1:rowsA
            result(i, :) = A(i, :) * B;
        end
    elseif flag == 4
        disp('Sum of outer products method')
        for i = 1:colsB
            result = result + A(:, i) * B(i, :);
        end
    end
end
```

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8.

(a)

```
>> num_matrices = 10^6;
matrix_size = 3;
singular_count = 0;
>> for i = 1:num_matrices
    random_matrix = rand(matrix_size);
    if rank(random_matrix) < matrix_size
        singular_count = singular_count + 1;
    end
end
>> disp(['Out of ', num2str(num_matrices), ' random matrices with uniform entries, ', num2str(singular_count), ' were singular.']);
Out of 1000000 random matrices with uniform entries, 0 were singular.
```

```
>> while true
    random_matrix = randi([1, 3], matrix_size);
    if rank(random_matrix) < matrix_size
        break;
    end
end
```

```
disp('A singular matrix with entries from {1, 2, 3}:');
disp(random_matrix);
A singular matrix with entries from {1, 2, 3}:
```

3	3	1
2	2	1
3	3	3

#

(b)

```
>> num_matrices = 100000;
matrix_size = 4;
l_values = [1, 2, 3];

rank_counts = zeros(1, matrix_size + 1);

for i = 1:num_matrices
    random_matrix = zeros(matrix_size);
    for row = 1:matrix_size
        for col = 1:matrix_size
            random_index = randi(length(l_values));
            random_matrix(row, col) = l_values(random_index);
        end
    end

    rank_value = rank(random_matrix);
    rank_counts(rank_value + 1) = rank_counts(rank_value + 1) + 1;
end

rank_distribution = rank_counts / num_matrices;

disp('Rank Distribution: ');
disp(rank_distribution);
Rank Distribution:
```

0	0	0.0024	0.1696	0.8280	7
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