

Section 1.9

The Matrix of a Linear Transformation

Unit Coordinate Vectors

Definition

The **unit coordinate vectors** in \mathbf{R}^n are

$$\mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{e}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad \dots, \quad \mathbf{e}_{n-1} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{pmatrix}, \quad \mathbf{e}_n = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}.$$

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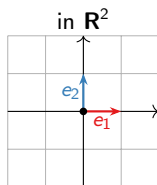
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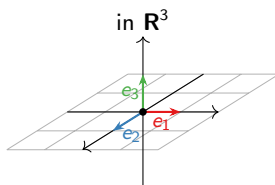
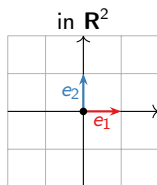
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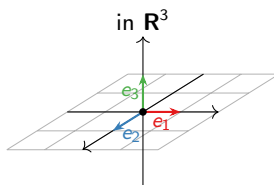
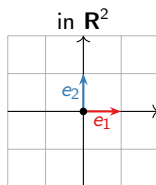
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Note: if A is an $m \times n$ matrix with columns v_1, v_2, \dots, v_n , then $Ae_i = v_i$ for $i = 1, 2, \dots, n$:

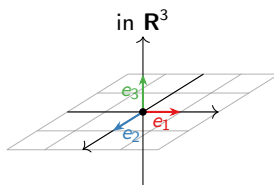
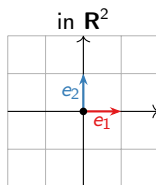
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Note: if A is an $m \times n$ matrix with columns v_1, v_2, \dots, v_n , then $Ae_i = v_i$ for $i = 1, 2, \dots, n$: multiplying a matrix by e_i gives you the i th column.

Linear Transformations are Matrix Transformations

Recall: A matrix A defines a linear transformation T by $T(x) = Ax$.

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Let $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ be a linear transformation. Let

$$A = \begin{pmatrix} \begin{array}{c} | \\ T(e_1) \\ | \end{array} & \begin{array}{c} | \\ T(e_2) \\ | \end{array} & \cdots & \begin{array}{c} | \\ T(e_n) \\ | \end{array} \end{pmatrix}.$$

This is an _____ matrix,

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Linear transformations are the same as matrix transformations.

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$T(x) = Ax$ \longleftarrow $m \times n$ matrix A

$T: \mathbf{R}^n \rightarrow \mathbf{R}^m$

Linear Transformations are Matrix Transformations

Continued

Why is a linear transformation a matrix transformation?

Linear Transformations are Matrix Transformations

Continued

Why is a linear transformation a matrix transformation?

Suppose for simplicity that $T: \mathbf{R}^3 \rightarrow \mathbf{R}^2$.

Linear Transformations are Matrix Transformations

Example

Before, we defined a **dilation** transformation $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ by $T(x) = 1.5x$.
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Linear Transformations are Matrix Transformations

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What is the matrix for the linear transformation $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ defined by

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(Check linearity. . .)

Linear Transformations are Matrix Transformations

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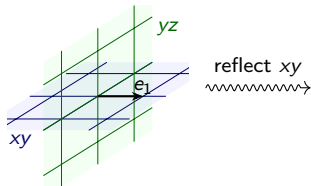
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Linear Transformations are Matrix Transformations

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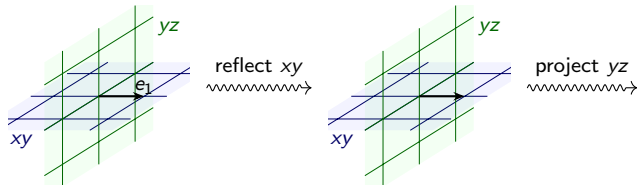


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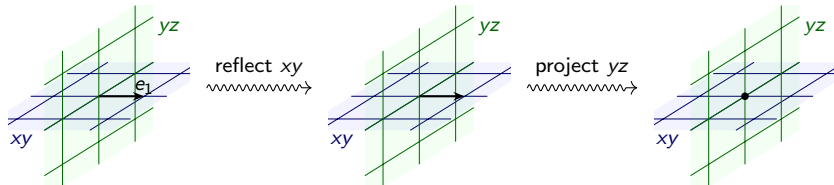


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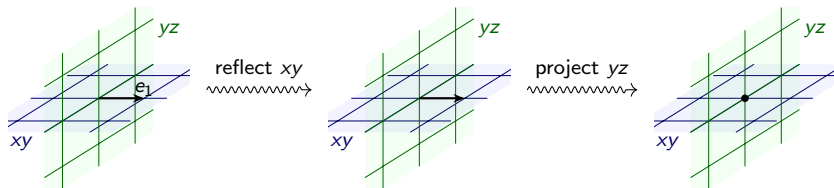


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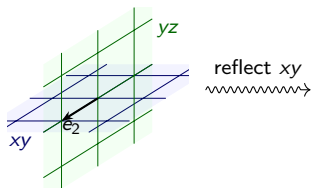
$$T(e_1) = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

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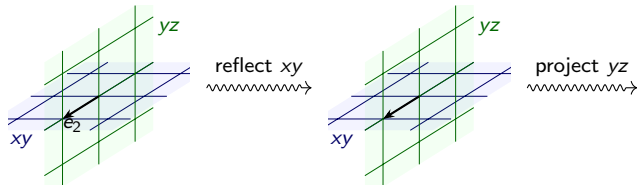


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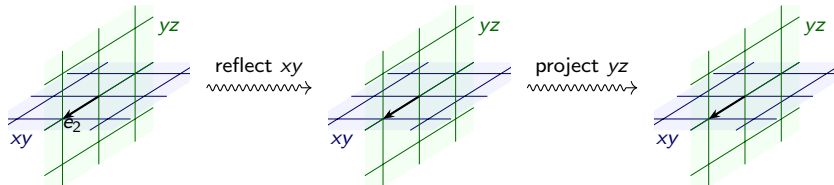


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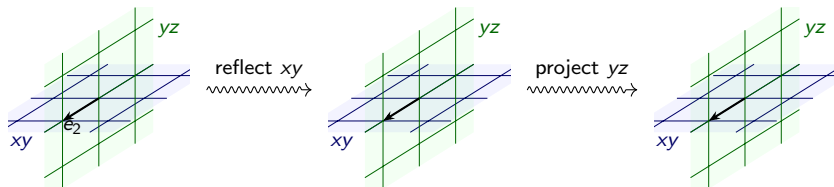


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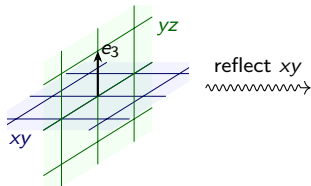
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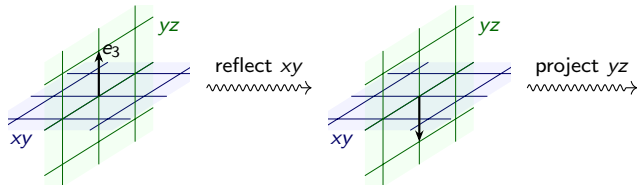


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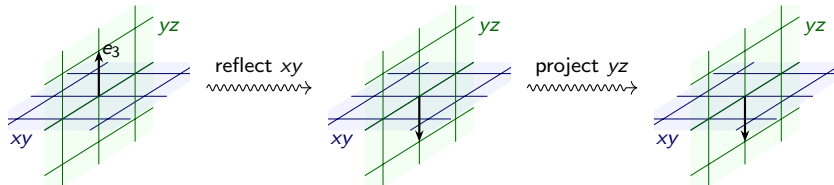


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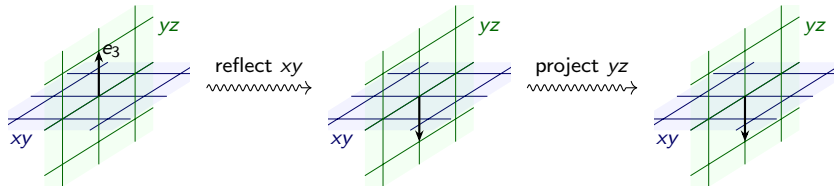


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There is a long list of geometric transformations of \mathbf{R}^2 in §1.9 of Lay. (Reflections over the diagonal, contractions and expansions along different axes, shears, projections, ...) Please look them over.

Onto Transformations

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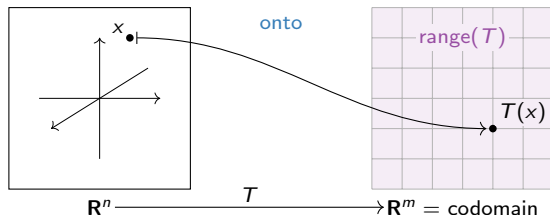
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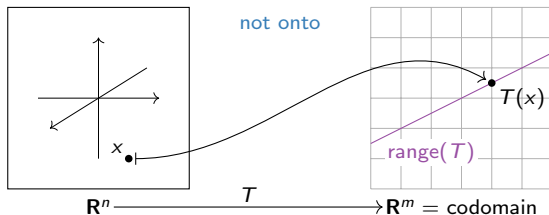
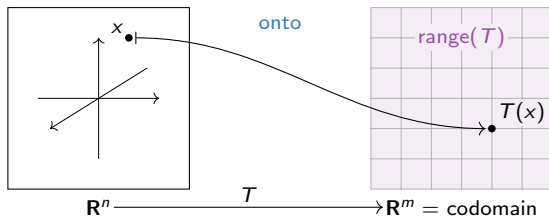
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If $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ is onto, what can we say about the relative sizes of n and m ?

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Let $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ be a linear transformation with matrix A . Then the following are equivalent:

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For instance, \mathbf{R}^2 is “too small” to map *onto* \mathbf{R}^3 .

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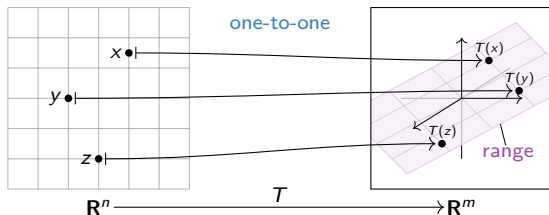
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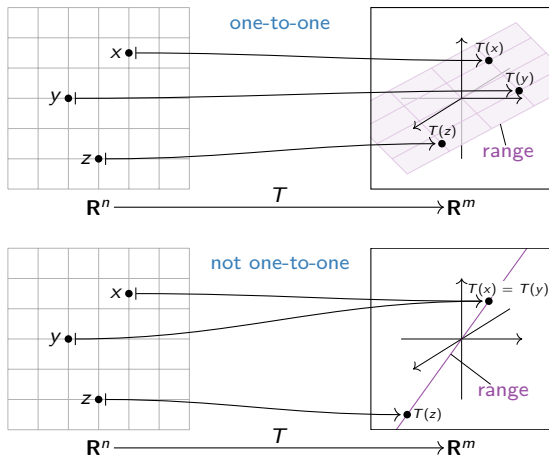
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