## Section 5.2. Geometric Transformations

**Note.** Gentle states that " $\ldots$  a vector represents a point in space...." This is misleading. We can associate a vector in  $\mathbb{R}^n$ ,  $v = (x_1, x_2, \ldots, x_n)$ , with a point in  $\mathbb{R}^n$ ,  $p = (x_1, x_2, \ldots, x_n)$  by representing the vector as an arrow in standard position with its tail at the origin and its head at point  $p$ . Often, vectors are notationally distinguished from points by using square brackets for vectors and parentheses for points: vector  $v = [x_1, x_2, \ldots, x_n]$  and point  $p = (x_1, x_2, \ldots, x_n)$ ; see my online notes for Linear Algebra (MATH 2010) on [1.1. Vectors in Euclidean Spaces.](http://faculty.etsu.edu/gardnerr/2010/c1s1.pdf) Vectors and points in  $\mathbb{R}^n$  are very different. For example, vectors can be added together and multiplied by scalars, but vectors don't have a location. Points cannot be added together nor multiplied by scalars, but points do have a location.

**Definition.** A transformation that preserves lengths and angles is an *isometric* transformation. A transformation that preserves angles is an *isotropic transforma*tion; a transformation that does not preserve angles is *anisotropic*. A transformation of the form mapping x to  $x+t$  (where  $x, t \in \mathbb{R}^n$ ) is a translation transformation.

Note. Gentle states that all transformations in this section "are linear transformations because they preserve straight lines" (page 175). This is unusual and a linear transformation  $T : \mathbb{R}^n \to \mathbb{R}^m$  is usually defined as satisfying  $T(ax + by) =$  $aT(x) + bT(y)$  for all  $a, b \in \mathbb{R}$  and  $x, y \in \mathbb{R}^n$ . In this case,  $T(0) = 0$  (since  $T(0) = T(0 + 0) = T(0) + T(0)$ , so a translation is not an example of a linear transformation in this sense. You see in Linear Algebra that all linear transformations (in this traditional sense) from  $\mathbb{R}^m$  to  $\mathbb{R}^n$  are represented by  $n \times m$  matrices; see my online Linear Algebra (MATH 2010) notes on [2.3. Linear Transformations](http://faculty.etsu.edu/gardnerr/2010/c2s3.pdf) [of Euclidean Spaces.](http://faculty.etsu.edu/gardnerr/2010/c2s3.pdf)

**Note.** Consider a vector x in  $\mathbb{R}^2$  with components  $x_1$  and  $x_2$ . In standard position in the  $\mathbb{R}^2$  ("geometric") Cartesian plane, we can represent x as an arrow from the origin to the point  $(x_1, x_2)$ . We now find a transformation that rotates x about the origin through an angle  $\theta$ . With  $\varphi$  as the angle between the positive x-axis and vector x we have  $x_1 = ||x|| \cos \varphi$  and  $x_2 = ||x|| \sin \varphi$ . With x rotated through angle  $\theta$  we produce  $\tilde{x}$  with endpoints at the origin and the point  $(\tilde{x}_1, \tilde{x}_2)$ .



Then the angle between the positive x-axis and  $\tilde{x}$  is  $\varphi + \theta$  and so  $\tilde{x}_1 = ||x|| \cos(\varphi + \theta)$ and  $\tilde{x}_2 = ||x|| \sin(\varphi + \theta)$ . The summation formulae for sine and cosine are

> $\cos(\varphi + \theta) = \cos \varphi \cos \theta - \sin \varphi \sin \theta$  $\sin(\varphi + \theta) = \sin \varphi \cos \theta + \cos \varphi \sin \theta.$

Now  $\cos \varphi = x_1 / ||x||$  and  $\sin \varphi = x_2 / ||x||$ , so

 $So$ 

$$
\tilde{x}_1 = ||x||((x_1/||x||)\cos\theta - (x_2/||x||)\sin\theta) = x_1\cos\theta - x_2\sin\theta
$$
  
\n
$$
\tilde{x}_2 = ||x||((x_2/||x||)\cos\theta + (x_1/||x||)\sin\theta) = x_2\cos\theta + x_1\sin\theta.
$$
  
\nSo with  $A = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$  we have  $\tilde{x} = Ax$ . Notice that A is an orthogonal matrix.

**Note.** In  $\mathbb{R}^3$ , to rotate a 3-vector about the *y*-axis we use

$$
B = \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}.
$$

The reason for the signs of the sines (!) is that in a right hand coordinate system if we rotate the  $xz$ -plane into the  $xy$ -plane as given in the image above then the y-axis points down into the image. Then a positive angle in the image represents a negative rotation about the y-axis (so to generate B we add a middle row and column to A with entries 0, 1, 0 [in this order] and then replace  $\theta$  with  $-\theta$  in the trig functions). Similarly, to rotate a 3-vector about the x-axis we use

$$
C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{bmatrix}.
$$

Of course a rotation about the z-axis in  $\mathbb{R}^3$  results from

$$
\left[\begin{array}{ccc} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{array}\right].
$$

Euler's Rotation Theorem implies that a rotation in  $\mathbb{R}^3$  about any axis (through the origin) is a combination of rotations about the x, y, and z axis. In fact, the collection of all rotation matrices form an infinite group called the "3D rotation group" or the " group of special orthogonal  $3 \times 3$  matrices" denoted  $SO(3)$ . The elements of  $SO(3)$  are all orthogonal  $3 \times 3$  matrices of determinant 1. The group of all  $3 \times 3$  orthogonal matrices form the "orthogonal group"  $O(3)$ .

Note. In sophomore Linear Algebra (MATH 2010) you encounter reflections in  $\mathbb{R}^2$  about the x-axis, y-axis, and the line  $y = x$  (see my online Linear Algebra notes on [2.4. Linear Transformations of the Plane\)](http://faculty.etsu.edu/gardnerr/2010/c2s4.pdf). This idea can be generalized to reflections about any line in any direction using unit vectors.

**Definition.** Let u and v for orthonormal vectors and let x be a vector in the space spanned by  $u$  and  $v$  (in which case we can represent the span of  $u$  and  $v$  as a two-dimensional plane). Then  $x = c_1u + c_2v$  for some scalars  $c_1$  and  $c_2$ . The vector  $\tilde{x} = -c_1u + c_2v$  is the *reflection* of x about v in the direction u.

Note. Geometrically, the reflection of x about  $v$  in the direction  $u$  can be represented as:



Note. As commented above, translations are not (in the usual sense) linear transformations and so cannot be represented by matrix multiplication. We introduce a new set of coordinates which allows us to "translate" vectors into the new coordinates and then represent translation by matrix multiplication in the new coordinates.

**Definition.** For point  $(x_1, x_2, \ldots, x_d) \in \mathbb{R}^d$  we introduce homogeneous coordinates  $(x_0^h)$  $\mathbf{b}_0^h, x_1^h, x_2^h, \ldots, x_d^h \in \mathbb{R}^{d+1}$  where  $x_1^h = x_0^h x_1, x_2^h = x_0^h x_2, \ldots, x_d^h = x_0^h x_d.$ 

**Note.** Homogeneous coordinates in  $\mathbb{R}^{d+1}$  represent points in  $\mathbb{R}^d$ . When  $x_0^h = 1$ , we have  $x_i^h = x_i$  for  $1 \leq i \leq d$ . When  $x_0^h = 0$  the homogeneous coordinate corresponds to no point in  $\mathbb{R}^d$ ; Gentle says that the collection of all homogeneous coordinate elements of  $\mathbb{R}^{d+1}$  where  $x_0^h = 0$  corresponds in projective geometry to the "hyperplane at infinity" (page 179).

Note. We now represent the translation transformation mapping x to  $x + t = \tilde{x}$ , where  $x, t \in \mathbb{R}^d$ ,  $x = [x_1, x_2, \dots, x_d]^T$ , and  $t = [t_1, t_2, \dots, t_d]^T$ , with matrix multiplication and homogeneous coordinates. First we represent the point  $(x_1, x_2, \ldots, x_d)$ in homogeneous coordinates as  $(1, x_1, x_2, \ldots, x_d)$ . Then for the  $(d + 1) \times (d + 1)$ matrix

$$
T = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ t_1 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ t_d & 0 & \cdots & 1 \end{bmatrix},
$$

the vector  $x^h = [1, x_1, x_2, \dots, x_d]^T$  satisfies  $Tx^h = [1, x_1 + t_1, x_2 + t_2, \dots, x_d + t_d]^T =$  $\tilde{x}^h$ . Now  $\tilde{x}^h$  is "associated" with the point  $(1, x_1 + t_1, x_2 + t_2, \ldots, x_d + t_d)$  which is the homogeneous coordinates associated with the vector  $x + t = \tilde{x} \in \mathbb{R}^d$ . Therefore, translation by any vector  $t \in \mathbb{R}^d$  is represented by matrix multiplication by T, but we must first convert/associate  $x$  with homogeneous coordinates, then perform the matrix multiplication, and finally convert/associate the homogeneous coordinates with the vector  $x + t$  in  $\mathbb{R}^d$ . As Gentle insightfully states: "We must be careful to distinguish the point  $x$  from the vector that represents the point" (page 179).

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