

## Projection Matrices

We discussed projection matrices briefly when we discussed orthogonal projection. In particular, we discussed the following theorem.

**Theorem.** Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be an orthonormal basis for a subspace  $W$  of  $\mathbb{R}^n$ . Form the  $n \times k$  matrix

$$\mathbf{U} = \left[ \begin{array}{c|c|c|c} \mathbf{u}_1 & \mathbf{u}_2 & \dots & \mathbf{u}_k \end{array} \right].$$

Then  $\text{proj}_W \mathbf{v} = \mathbf{U}\mathbf{U}^T \mathbf{v}$ .

The matrix  $\mathbf{U}\mathbf{U}^T$  is called the *projection matrix* for the subspace  $W$ . It does not depend on the choice of orthonormal basis.

What if we do not start with an orthonormal basis of  $W$ ?

**Theorem.** Let  $\{\mathbf{a}_1, \dots, \mathbf{a}_k\}$  be any basis for a subspace  $W$  of  $\mathbb{R}^n$ . Form the  $n \times k$  matrix

$$\mathbf{A} = \left[ \begin{array}{c|c|c|c} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_k \end{array} \right].$$

Then the projection matrix for  $W$  is  $\mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$ .

To see why this formula is true, we need a lemma.

**Lemma.** Suppose  $\mathbf{A}$  is an  $n \times k$  matrix whose columns are linearly independent. Then  $\mathbf{A}^T \mathbf{A}$  is invertible.

To see why this lemma is true, consider the transformation  $\mathbf{A} : \mathbb{R}^k \rightarrow \mathbb{R}^n$  determined by  $\mathbf{A}$ . Since the columns of  $\mathbf{A}$  are linearly independent, this transformation is one-to-one. Moreover, the null space of  $\mathbf{A}^T$  is orthogonal to the column space of  $\mathbf{A}$ . Consequently,  $\mathbf{A}^T$  is one-to-one on the column space of  $\mathbf{A}$ , and as a result,  $\mathbf{A}^T \mathbf{A} : \mathbb{R}^k \rightarrow \mathbb{R}^k$  is one-to-one. By the Invertible Matrix Theorem,  $\mathbf{A}^T \mathbf{A}$  is invertible.

Now we can compute the projection matrix for the column space of  $\mathbf{A}$ . (Note that  $W = \text{Col } \mathbf{A}$ .) Any element of the column space of the matrix  $\mathbf{A}$  is a linear combination of the columns of  $\mathbf{A}$ , that is,

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_k \mathbf{a}_k.$$

If we let

$$\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_k \end{bmatrix},$$

then

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_k \mathbf{a}_k = \mathbf{A}\mathbf{x}.$$

Given  $\mathbf{v}$  in  $\mathbb{R}^n$ , we denote by  $\mathbf{x}_p$  the  $\mathbf{x}$  that corresponds to the projection of  $\mathbf{v}$  onto  $W$ . In other words, let

$$\text{proj}_W \mathbf{v} = \mathbf{A}\mathbf{x}_p.$$

We find the projection matrix by calculating  $\mathbf{x}_p$ .

The projection of  $\mathbf{v}$  onto  $W$  is characterized by the fact that

$$\mathbf{v} - \text{proj}_W \mathbf{v}$$

is orthogonal to each vector  $\mathbf{w}$  in  $W$ , that is,

$$\mathbf{w} \cdot (\mathbf{v} - \text{proj}_W \mathbf{v}) = 0$$

for all  $\mathbf{w}$  in  $W$ . Since  $\mathbf{w} = \mathbf{A}\mathbf{x}$  for some  $\mathbf{x}$ , we have

$$\mathbf{A}\mathbf{x} \cdot (\mathbf{v} - \mathbf{A}\mathbf{x}_p) = 0$$

for all  $\mathbf{x}$  in  $\mathbb{R}^k$ . Writing this dot product in terms of matrices yields

$$(\mathbf{A}\mathbf{x})^T (\mathbf{v} - \mathbf{A}\mathbf{x}_p) = 0,$$

which is equivalent to

$$(\mathbf{x}^T \mathbf{A}^T) (\mathbf{v} - \mathbf{A}\mathbf{x}_p) = 0.$$

Converting back to dot products, we have

$$\mathbf{x} \cdot \mathbf{A}^T (\mathbf{v} - \mathbf{A}\mathbf{x}_p) = 0.$$

In other words, the vector  $\mathbf{A}^T (\mathbf{v} - \mathbf{A}\mathbf{x}_p)$  is orthogonal to every vector  $\mathbf{x}$  in  $\mathbb{R}^k$ . The only vector in  $\mathbb{R}^k$  with this property is the zero vector, so we may conclude that

$$\mathbf{A}^T (\mathbf{v} - \mathbf{A}\mathbf{x}_p) = \mathbf{0}.$$

We get

$$\mathbf{A}^T \mathbf{v} = \mathbf{A}^T \mathbf{A}\mathbf{x}_p.$$

From the lemma, we know that  $\mathbf{A}^T \mathbf{A}$  is invertible, and we have

$$(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{v} = \mathbf{x}_p.$$

Since  $\mathbf{A}\mathbf{x}_p$  is the desired projection, we have

$$\mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{v} = \text{proj}_W \mathbf{v}.$$

We conclude that the projection matrix for  $W$  is

$$\mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T.$$

Note that any projection matrix  $\mathbf{P}$  satisfies the two properties

1.  $\mathbf{P}^2 = \mathbf{P}$ , and
2.  $\mathbf{P}$  is symmetric.

It is also true that any matrix that satisfies these two properties is the projection matrix for some subspace of  $\mathbb{R}^n$  (see Exercise 36 in Section 7.1 of Lay).