

W07-01

Slide 01: In this unit, we will discuss how to compute the orthogonal projection of a vector onto a subspace.

Slide 02: Let V be a subspace of \mathbb{R}^n that we would like to project onto. Suppose \mathbf{w} is a vector in \mathbb{R}^n . If we have an orthogonal basis for V , comprising of vectors $\mathbf{u}_1, \mathbf{u}_2$ to \mathbf{u}_k , then the projection of \mathbf{w} onto V

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is given by the following expression. Notice that this expression is a linear combination of the orthogonal basis vectors \mathbf{u}_1 to \mathbf{u}_k . This is expected as the projection is a vector in V and thus should be expressible as a linear combination of the basis vectors.

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If we have an orthonormal basis for V instead, comprising of vectors $\mathbf{v}_1, \mathbf{v}_2$ to \mathbf{v}_k , then the projection of \mathbf{w} onto V , written as a linear combination of the orthonormal basis vectors is as shown.

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Take another look at the expression for the projection of \mathbf{w} onto V . Do you find this expression familiar? Have you seen the same expression from an earlier unit?

Slide 03: Indeed we have seen this expression before. Recall from an earlier unit, when we introduced the special class of bases known as orthogonal basis, we saw that if \mathbf{u}_1 to \mathbf{u}_k is an orthogonal basis for a vector space V , then to write any vector \mathbf{w} in V as a linear combination of the orthogonal basis vectors can be done without having to solve any linear systems to find the coefficients. This expression is precisely how \mathbf{w} can be written in terms of \mathbf{u}_1 to \mathbf{u}_k .

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Now compare to what we have just seen from the previous slide, which states that if \mathbf{u}_1 to \mathbf{u}_k is an orthogonal basis for the vector space V , then for any \mathbf{w} in \mathbb{R}^n , the projection of \mathbf{w} onto V is given by the same expression as we have seen above.

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Why are the two expressions identical and why does this make sense?

Slide 04: To see why this makes sense, consider the case where we are projecting a vector \mathbf{w} onto V .

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The figure here shows \mathbf{w} to be outside V ,

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so when we project \mathbf{w} onto V , we obtain a different vector \mathbf{p} whose expression is the linear combination of \mathbf{u}_1 to \mathbf{u}_k as shown.

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Recall that by orthogonal projection, we mean decomposing \mathbf{w} and writing it as the sum of two vectors \mathbf{p} and \mathbf{n} where \mathbf{p} belongs to V and \mathbf{n} is orthogonal to V .

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We have mentioned earlier what orthogonal projection means if the vector \mathbf{w} is already in the subspace V that we would like to project on, as shown in the figure on the left.

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The projection of \mathbf{w} onto V in this case is simply \mathbf{w} itself. This is where the expression from an earlier unit comes in, where we are writing \mathbf{w} as a linear combination of the orthogonal basis vectors \mathbf{u}_1 to \mathbf{u}_k .

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If thought of as a projection, we are decomposing \mathbf{w} as $\mathbf{w} + \mathbf{0}$. So indeed there is no conflict in seeing the same expression twice, since the notion of projection can take place whether or not \mathbf{w} belongs to V .

Slide 05: We are now ready to present a proof of this orthogonal projection theorem. Remember that this theorem allows us to compute orthogonal projection of any \mathbf{w} in \mathbb{R}^n onto the subspace V **provided** we have an orthogonal basis for V .

Slide 06: To prove the theorem, let \mathbf{p} be the expression given in the statement of the theorem.

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We will show that \mathbf{p} is indeed the projection of \mathbf{w} onto V .

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First, observe that \mathbf{p} is definitely a vector in V since it is a linear combination of the basis vectors \mathbf{u}_1 to \mathbf{u}_k .

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Next, we define the vector \mathbf{n} to be the difference between \mathbf{w} and \mathbf{p} . Why do we want to do this?

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By doing this, it would mean that \mathbf{w} is in fact decomposed into \mathbf{n} plus \mathbf{p} , where we have already noted that \mathbf{p} is a vector in V .

Slide 07: What remains for us to do, in order to conclude that \mathbf{p} is the orthogonal projection of \mathbf{w} onto V , is to show that the vector \mathbf{n} , in the way we have defined it, is orthogonal to the space V . If we can do this, then we would have established that \mathbf{p} will be the projection of \mathbf{w} onto V .

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Recall that to show that a vector is orthogonal to a space, we need to show that it is orthogonal to the vectors that spans the space. Thus, we will check whether \mathbf{n} is orthogonal to each of the \mathbf{u}_i 's.

Slide 08: So for each $i = 1$ to k , consider the dot product between \mathbf{n} with \mathbf{u}_i . Since \mathbf{n} is $\mathbf{w} - \mathbf{p}$, we have the expression on the right side of the equation,

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which can be simplified as follows by applying distributive law.

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We now write down the expression for \mathbf{p} as shown.

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Now consider the expression in the red box. Here we have the dot product between \mathbf{u}_i and a linear combination of \mathbf{u}_1 to \mathbf{u}_k .

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If we apply distributive law on this expression, we should have many terms as a result. Why is it that only one term remain? Namely, the only term that remains seems to be the $\mathbf{u}_i \cdot \mathbf{u}_i$ term.

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The reason is because that \mathbf{u}_1 to \mathbf{u}_k is an orthogonal set. This means that the vectors in the set are pairwise orthogonal, so the dot product between \mathbf{u}_i and \mathbf{u}_j will be zero whenever $i \neq j$. This is why only the $\mathbf{u}_i \cdot \mathbf{u}_i$ term remains.

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Now the dot product of \mathbf{u}_i with itself is the square of the length of \mathbf{u}_i . This will cancel with the denominator of the coefficient,

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which reduces the expression to $\mathbf{w} \cdot \mathbf{u}_i$ minus $\mathbf{w} \cdot \mathbf{u}_i$ which is 0.

Slide 09: We have thus established that \mathbf{n} is indeed a vector orthogonal to V since it is orthogonal to the vectors that spans V .

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The vector \mathbf{w} in \mathbb{R}^n , has now been written as the sum of two vectors \mathbf{n} and \mathbf{p} where

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\mathbf{n} is orthogonal to V while

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\mathbf{p} belongs to V . So the vector \mathbf{p} is indeed the orthogonal projection of \mathbf{w} onto V .

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The second part of the theorem actually follows immediately from the first part, since when \mathbf{v}_1 to \mathbf{v}_k is an orthonormal basis for V , they are orthogonal vectors of length 1.

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Thus the denominators of all the coefficients in the expression for \mathbf{p} is now 1 and we will obtain the expression for the projection of \mathbf{w} onto V accordingly.

Slide 10: Let us go through one example. Let V be a subspace of \mathbb{R}^3 spanned by the two vectors $(1, 0, 1)$ and $(1, 0, -1)$. Notice that these two vectors are not multiples of each other, meaning that they are linearly independent. Thus the two vectors form a basis for V .

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In addition, note that the dot product between the two vectors is 0, which means that these two vectors are orthogonal.

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So the two vectors actually form an orthogonal basis for V . Now that we have an orthogonal basis for V , we are ready to compute projection of vectors onto V .

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What is the projection of the vector $(1, 1, 0)$ onto V ?

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Using the orthogonal projection theorem, we have the following linear combination of $(1, 0, 1)$ and $(1, 0, -1)$ as the projection of \mathbf{w} onto V .

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This simplifies to the vector $(1, 0, 0)$ which is the projection of \mathbf{w} onto V .

Slide 11: What about this subspace V of \mathbb{R}^3 ? It is also spanned by two vectors $(1, 1, 1)$ and $(1, 3, -1)$ which are not multiples of each other. So these two vectors also forms a basis for V .

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However, in this case, the dot product between the two vectors is not zero, which means that this is not an orthogonal basis.

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Remember that to apply the orthogonal projection theorem, we need to have an orthogonal basis for the subspace V that we wish to project on.

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So how can we compute the projection of \mathbf{w} onto V in this case?

Slide 12: Clearly, we cannot apply the orthogonal projection theorem in this case.

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So when we are faced with the situation that we do not have an orthogonal basis for the subspace V , what can we do?

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In a subsequent unit, we will introduce a process where we can construct an orthogonal basis for a subspace V . Such a basis, once constructed, will then allow us to use the orthogonal projection theorem to compute projection.

Slide 13: In summary, we have seen in this unit

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how to compute orthogonal projection of a vector onto a vector space V . This can be done provided we have an orthogonal or orthonormal basis for V .