

Mathematical Methods 1
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Linear Algebra
Lecture – 21
Gram-Schmidt orthogonalization

So, we have seen how in an n -dimensional vector space, if you consider an arbitrary you know set of n linearly independent vectors, these form a basis for you and a basis is a convenient you know set to work with because any vector in that space can be represented as a linear combination of vectors from this basis right.

Now, it turns out that although we can find a basis in many different ways, there are infinitely many ways; there are certain kinds of bases which are more convenient right; so these are called Ortho-normal basis right. And, 1 example is you know from two-dimensional vectors, we know that e_x and e_y right, we have perhaps without thinking about this, we have been using e_x and e_y to form the basis for all vectors in 2D right.

So, the vector unit vector along the x direction and the unit vector along the y direction; but in our recent discussion, we have seen that you know e_x and $e_x + e_y$ is an equally valid basis. We could have represented all our physics in terms of e_x and $e_x + e_y$ or $e_x + e_y$ and $e_x + 3e_y$ for example or any two linearly independent vectors from the plane is that valid basis.

But you know the reason why e_x and e_y is a particularly convenient basis is because you know there is no component of e_x along e_y and there is no component of e_y along e_x . So, the idea here is that e_x and e_y are normal, are orthogonal to each other and each of them are themselves normal right.

So, that, so what it means is each of the vectors of your basis is bringing a unique piece of information for the set right. Together of course, they add up to the entire information of the overall space and there is no redundancy.

The fact that they are a basis implies that there is no collective redundancy; but if you make it an orthonormal basis right, if all the basis vectors are orthogonal, it means that no vector carries any information contained in another vector. That is what the orthogonal basis does right.

So, in this lecture I am going to describe a systematic method by which you can start with a basis of any kind, it is a basis and that can be made into an orthonormal basis from this right. So, this is called the Gram-Schmidt orthogonal orthogonalization process ok.

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Gram-Schmidt orthogonalization

Orthonormal basis

We call a set of vectors *orthonormal* if they are all mutually *orthogonal*, and each vector is *normalized*. The Gram-Schmidt method is a systematic process for orthonormalizing a basis.

Let us provide a construction for starting with a basis of vectors, and obtaining an orthonormal basis.

Given a set B of m linearly independent vectors $\{|x_1\rangle, |x_2\rangle, \dots, |x_m\rangle\}$, a set S of mutually orthonormal vectors $\{|e_1\rangle, |e_2\rangle, \dots, |e_m\rangle\}$ can always be constructed such that each $|e_j\rangle$ is a linear combination of the vectors $\{|x_1\rangle, |x_2\rangle, \dots, |x_m\rangle\}$.

We start by normalizing the first vector, and calling it the first vector of the new set that we wish to construct:

$$|e_1\rangle = \frac{|x_1\rangle}{\sqrt{\langle x_1 | x_1 \rangle}}$$

So, yeah; so, like I said, we call a set of vectors orthonormal, if they are all mutually orthogonal to each other right and each vector is normalized. So, making each vector normalized simply means you take the if you take the inner product of a vector with itself, you have to get 1 right. So, this is you know some added convenience even if they are not normalized.

So, it is just one more step to normalize it as well. So, we will show here how there is a systematic way to take any basis and make it and get an orthonormal basis ok. Given a set B of m , m linearly independent vectors; a set, a set of s a set s of mutually in orthonormal vectors e_1 to e_m right; it can always be constructed that is the statement and so, how do we do this?

So, let us start by picking up any one vector right. Without loss of generality, we can just call it x_1 . So, e_1 is equal to x_1 divided by its norm right so that e_1 has been normalized. So, the first step is to take any one vector and normalize it and that will be your first element in the basis that you are creating.

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Next we construct a vector that is orthonormal to the first. This can of course be done in many different ways, so let us work with the second vector $|x_2\rangle$, without loss of generality. We need to remove the component of this vector along $|e_1\rangle$, which is done explicitly with:

$$|e_2'\rangle = |x_2\rangle - \langle e_1 | x_2 \rangle |e_1\rangle$$

Normalizing this vector yields the second vector of our set:

$$|e_2\rangle = \frac{|e_2'\rangle}{\sqrt{\langle e_2' | e_2' \rangle}}$$

We continue this procedure, each time considering a new vector from the original set, and removing from it all the components in directions already represented so far, before normalizing it. To create the $(i+1)^{\text{th}}$ vector of the new basis, we would first form the intermediate vector:

$$|e_{i+1}'\rangle = |x_{i+1}\rangle - \sum_{j=1}^i \langle e_j | x_{i+1} \rangle |e_j\rangle$$

and then normalize it to obtain:

Now, the next thing to do is pick any other vectors. Now, you could have picked any one of the other n minus 1 vectors. There is no problem, there is a lot of choice in this right and in fact, you can create lots of different basis, now orthonormal basis using this method. So, I am just showing you that it is possible to find such an orthonormal basis in a systematic way.

Now, pick x_2 without loss of generality and then, what you do is you peel off you know the component of x_2 which lies along e_1 right. So, x_2 is; so this is what we do. How do we find it? How do we peel off the component of x_2 along e_1 ? You just take the inner product of x_2 with e_1 you know and then, attach this vector e_1 and then, you remove that component from x_2 .

So, this is an intermediate vector which we call e_2 prime right and to get e_2 , we have to just normalize this vector right. We are interested in creating an orthonormal basis. So, we have to normalize this e_2 prime. So, then we have e_2 prime divided by square root of the inner product of e_2 prime with e_2 prime that is the second vector e_2 right. By construction,

you can check that this vector e_2 is going to be orthonormal, orthogonal to e_1 and it is also normalized by construction right.

So, the next step is similar right. Now, you have to peel off, you take the third vector x_3 without loss of generality and will peel off, you know the component of x_3 along x_2 and the component of x_3 along x_1 and normalize it and so on.

And then, you keep on doing this for the i th vector the $i + 1$ th vector is going to be obtained by peeling off from the $i + 1$ th vector you know components along every vector which has already been put into your box of orthonormal vectors that you creating right. So, e_{i+1}' is going to be x_{i+1} minus summation over $j \leq i$ of $\langle x_{i+1}, e_j \rangle e_j$ times the vector e_j right.

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The slide content is as follows:

$$|e'_{i+1}\rangle = |x_{i+1}\rangle - \sum_{j=1}^i \langle e_j | x_{i+1} \rangle |e_j\rangle$$

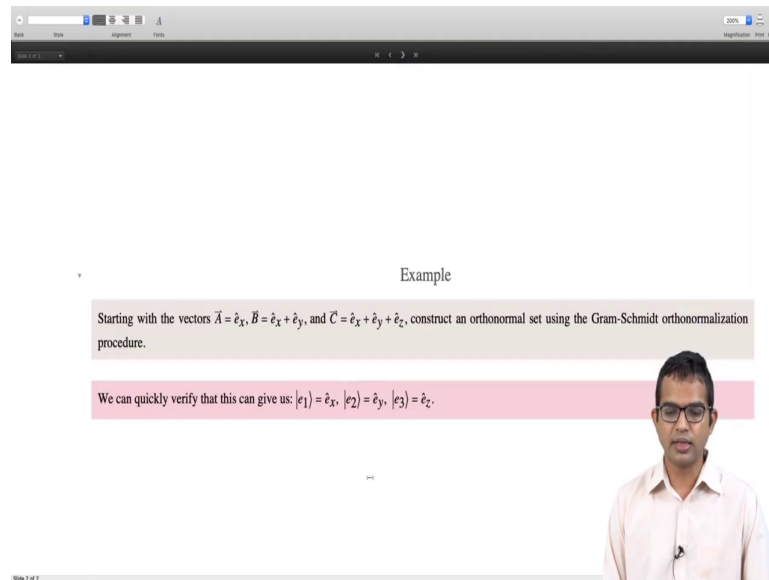
and then normalize it to obtain:

$$|e_{i+1}\rangle = \frac{|e'_{i+1}\rangle}{\sqrt{\langle e'_{i+1} | e'_{i+1} \rangle}}$$

Evidently, every vector in this new set is orthogonal to every other vector by construction. Moreover, each of them has been normalized, so we have created an orthonormal basis.

If you do this and then, you have to normalize it. You have to do each e_{i+1}' divided by square root of the e_{i+1}' inner product of e_{i+1}' with e_{i+1}' and then, so evidently, you can see that by construction every vector in this; in this set is orthogonal to every other vector. And also, every vector has been explicitly normalized. So, you have created an orthonormal basis.

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The image shows a presentation slide with a video overlay of a man in a white shirt and glasses. The slide content is as follows:

Example

Starting with the vectors $\vec{A} = \hat{e}_x$, $\vec{B} = \hat{e}_x + \hat{e}_y$, and $\vec{C} = \hat{e}_x + \hat{e}_y + \hat{e}_z$, construct an orthonormal set using the Gram-Schmidt orthonormalization procedure.

We can quickly verify that this can give us: $|\hat{e}_1\rangle = \hat{e}_x$, $|\hat{e}_2\rangle = \hat{e}_y$, $|\hat{e}_3\rangle = \hat{e}_z$.

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So, that is how this method works. To convince yourself that this works out, I have given you an example which is very very simple and you might not need an elaborate method for this. But let us say that you start with the vectors e_x , $e_x + e_y$ and $e_x + e_y + e_z$ right. I am thinking of a three-dimensional space, you know the usual kind of vectors. e_x is given; $e_x + e_y$ is given and $e_x + e_y + e_z$ is given right.

You can, if you take the first vector to be A itself right. So, it has already been normalized. So, you have e_1 is equal to e_x and then, when you go to the second step e_2 . To create e_2 , you are going to peel off you know the component of B along e_x which is in this case will turn out to be just e_y and then, you go to the third vector C . You peel off the vector along A and the vector along B . So, then you will get e_z .

So, e_x , e_y and e_z is not just a basis; but it is an orthonormal basis. So, I have given you a very trivial example. But even with this trivial example, let us say you start not with a vector A , you I said even within the Gram-Schmidt orthonormalization processes itself, you have a lot of choice.

Suppose, you start with a vector C , you know make C the vector e_1 right. You have to do $e_1 = C / |C|$ right. C divided by mod of C that will be e_1 and then, maybe you can take B as it using B you have to find e_2 and find A , use A to find e_3 right. That would be left as a

homework exercise. You could also try starting with B, you know try out various different combinations and see how you can generate different kinds of orthonormal basis right; that is all for this lecture.

Thank you.