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# The Tensor Product, Demystified

November 18, 2018 • Algebra

[Previously on the blog](#), we've discussed a recurring theme in mathematics: making new things from old things. Mathematics is a process of building new structures from existing ones. For example, when you have two integers, you can find their greatest common multiple. When you have some sets, you can form their Cartesian product. When you have two groups, you can construct their direct product. When you have a topological space, you can look at its subspace. When you have some vector spaces, you can ask about their intersection. The list goes on!



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Ps. 148

I'd like to focus on a particular way to build a new vector space: *the tensor product*. This construction is a bit mysterious, but I hope to help shine a little light on it. In particular, we won't talk about axioms, universal properties, or *functors*. Instead, we'll take an elementary, concrete approach.

Given two vectors  $\mathbf{v}$  and  $\mathbf{w}$ , we can build a new vector space. But what is that vector, *really*? Likewise, given two vector spaces, we can build a new vector space, also called their *tensor product*. But what is that vector space, *really*?

## Building new vectors from old

In this discussion, we'll assume  $V$  and  $W$  are finite dimensional. This means we can think of  $V$  as  $\mathbb{R}^n$  and  $W$  as  $\mathbb{R}^m$  for some  $n, m$ . So a vector  $\mathbf{v}$  in  $\mathbb{R}^n$  is really just a list of  $n$  numbers, and a vector  $\mathbf{w}$  in  $\mathbb{R}^m$  is a list of  $m$  numbers.

$$\vec{v} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

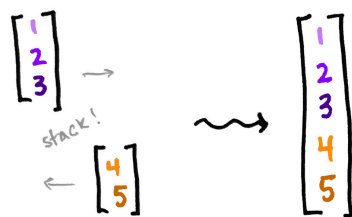
a vector in  $\mathbb{R}^3$

$\vec{w}$

a vector in  $\mathbb{R}^3$

Let's try to make new, third vector out of  $\mathbf{v}$  and  $\mathbf{w}$ . But we can't just add them together. We can stack them on top of each other, or we can first add them together and *then* stack them on top of each other.

option # 1

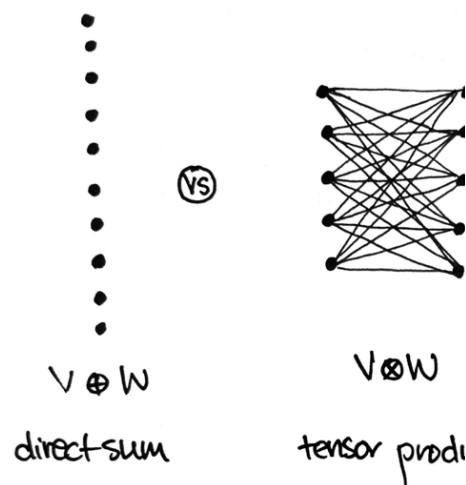


$$(\vec{v}, \vec{w}) \text{ in } \mathbb{R}^3 \oplus \mathbb{R}^2$$

direct sum

The first option gives a new list of  $n + m$  numbers, while the second gives a new list of  $nm$  numbers. The first gives a way to build dimensions *add*; the second gives a way to build dimensions *multiply*. The first is a vector  $(\mathbf{v}, \mathbf{w})$  in the same space as their direct product  $V \times W$ ; the second is a **tensor product**  $V \otimes W$ .

And that's it!



Forming the tensor product  $\mathbf{v} \otimes \mathbf{w}$  of two vectors is a Cartesian product of two sets  $X \times Y$ . In fact, that's exactly what you think of  $X$  as the set whose elements are the entries of  $\mathbf{v}$ .

If  $X$  and  $Y$  are finite sets, then this is their Cartesian product.

are these sets:

$$X = \{1, 2, 3\}$$

$$Y = \{4, 5\}$$

Cartesian product:

$$X \times Y = \{(1, 4), (1, 5), (2, 4), (2, 5), (3, 4), (3, 5)\}$$

So a tensor product is like a grown-up version of mult when you systematically multiply a bunch of numbers results into a list. It's multi-multiplication, if you will.

$$\begin{bmatrix} 2 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 2 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 5 \\ 1 \end{bmatrix} \otimes \begin{bmatrix} 1 \\ 0 \\ 4 \end{bmatrix} = \begin{bmatrix} 5 \\ 0 \\ 20 \\ 1 \\ 0 \\ 4 \end{bmatrix}$$

## There's a little more to the story

Does every vector in  $V \otimes W$  look like  $\mathbf{v} \otimes \mathbf{w}$  for some quite. Remember, a vector in a vector space can be wr *basis vectors*, which are like the space's building blocks: making new things from existing ones: we get a new  $\mathbf{v}$  sum of some special vectors!

If  $V$  has basis  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$

$$\mathbf{v} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = 1\mathbf{v}_1 + 2\mathbf{v}_2 + 3\mathbf{v}_3$$

So a typical vector in  $V \otimes W$  is a weighted sum of bas *basis vectors*? Well, there must be exactly  $nm$  of them,  $V \otimes W$  is  $nm$ . Moreover, we'd expect them to be built the basis of  $W$ . This brings us again to the "How can w

old things?" question. Asked explicitly: If we have  $n$  bases  $\mathbf{w}_1, \dots, \mathbf{w}_m$  for  $W$  then how can we set of  $nm$  vectors?

This is totally analogous to the construction we saw al and a list of  $m$  things, we can obtain a list of  $nm$  thing together. So we'll do the same thing here! We'll simply with the  $\mathbf{w}_j$  in all possible combinations, *except* "multi "take the tensor product of  $\mathbf{v}_i$  and  $\mathbf{w}_j$ ."

Concretely, a basis for  $V \otimes W$  is the set of all vectors  $\mathbf{v}_i \otimes \mathbf{w}_j$  where  $i$  ranges from 1 to  $n$  and  $j$  ranges from 1 to  $m$ . As an example, let  $n = 3$  and  $m = 2$  as before. Then we can find the six basis vectors of  $V \otimes W$  by constructing a 'multiplication chart.' (The sophisticated way to say that  $V \otimes W$  is a vector space on  $A \times B$ , where  $A$  is a set of generators for  $V$  and  $B$  is a set of generators for  $W$ .)

	$\vec{w}_1$	$\vec{w}_2$
$\vec{v}_1$	$\vec{v}_1 \otimes \vec{w}_1$	$\vec{v}_1 \otimes \vec{w}_2$
$\vec{v}_2$	$\vec{v}_2 \otimes \vec{w}_1$	$\vec{v}_2 \otimes \vec{w}_2$
$\vec{v}_3$	$\vec{v}_3 \otimes \vec{w}_1$	$\vec{v}_3 \otimes \vec{w}_2$

So  $V \otimes W$  is the six-dimensional space with basis

$$\{\mathbf{v}_1 \otimes \mathbf{w}_1, \mathbf{v}_1 \otimes \mathbf{w}_2, \mathbf{v}_2 \otimes \mathbf{w}_1, \mathbf{v}_2 \otimes \mathbf{w}_2, \mathbf{v}_3 \otimes \mathbf{w}_1, \mathbf{v}_3 \otimes \mathbf{w}_2\}$$

This might feel a little abstract with all the  $\otimes$  symbols don't forget—we know exactly what each  $\mathbf{v}_i \otimes \mathbf{w}_j$  looks like in terms of numbers! Which list of numbers? Well,

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \quad \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \quad \dots$$

$$\text{if } \mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \mathbf{v}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \mathbf{v}_3 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \mathbf{w}_1 =$$

$$\mathbf{v}_1 \otimes \mathbf{w}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{v}_1 \otimes \mathbf{w}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{v}_2 \otimes \mathbf{w}_1 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{v}_2 \otimes \mathbf{w}_2 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

So what is  $V \otimes W$ ? It's the vector space whose vectors are the  $\mathbf{v}_i \otimes \mathbf{w}_j$ . For example, here are a couple of vectors

$$7\mathbf{v}_1 \otimes \mathbf{w}_1 + 3\mathbf{v}_3 \otimes \mathbf{w}_2 = \begin{bmatrix} 7 \\ 0 \\ 0 \\ 0 \\ 0 \\ 3 \end{bmatrix} \quad -\mathbf{v}_1 \otimes \mathbf{w}_2 +$$

Well, technically...

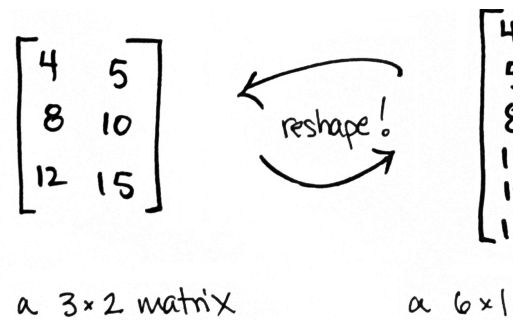
Technically,  $\mathbf{v} \otimes \mathbf{w}$  is called the **outer product** of  $\mathbf{v}$  and

$$\mathbf{v} \otimes \mathbf{w} := \mathbf{v} \mathbf{w}^\top$$

where  $\mathbf{w}^\top$  is the same as  $\mathbf{w}$  but written as a row vector. If the entries are complex numbers, then we also replace each entry with its complex conjugate. So technically the tensor product of vectors is matrix:

$$\vec{\mathbf{v}} \otimes \vec{\mathbf{w}} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 4 & 5 \end{bmatrix} = \begin{bmatrix} 1 \cdot 4 & 1 \cdot 5 \\ 2 \cdot 4 & 2 \cdot 5 \\ 3 \cdot 4 & 3 \cdot 5 \end{bmatrix}$$

This may seem to be in conflict with what we did above. Any  $m \times n$  matrix can be reshaped into a vector and vice versa. (So thus far, we've been exploiting the fact that  $\mathbb{R}^6$  is isomorphic to  $\mathbb{R}^3 \otimes \mathbb{R}^2$ .) You might refer to this as *matrix-vector duality*.



It's a little like a **process-state duality**. On the one hand,  $\mathbf{v} \otimes \mathbf{w}$  is, abstractly speaking, a vector. And a vector is a gadget that physicists use to describe the state of a quantum system; vectors encode states. The tensor product  $V \otimes W$  can be viewed in either way simply by list or as a rectangle.



By the way, this idea of viewing a matrix as a process or a state is also useful for *higher dimensional arrays*, too. These arrays are called **tensors**. If you do a bunch of these processes together, the resulting structure is called a **tensor network**. But manipulating high-dimensional arrays is very messy very quickly: there are lots of numbers that have to be kept together. This is like multi-multi-multi-multi...plication. (Tensor networks come with lovely pictures that make these concepts easier to understand. This goes back to [Roger Penrose's graphical calculus](#).) This is what we have here, but it'll have to wait for [another day](#)!

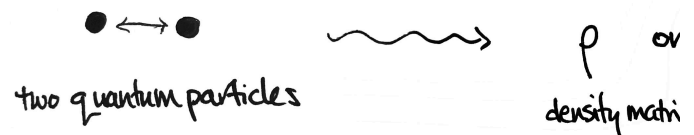
In quantum physics

## in quantum physics

One application of tensor products is related to the bra-ket notation. "A vector is the mathematical gadget that physicists use to describe the state of a quantum system." To elaborate: if you have a little quantum system, you'd like to know what it's doing. Or what it's capable of doing. Or what it'll be doing something. In essence, you're asking: What is its *state*? The answer to this question—provided by quantum mechanics—is given by a unit vector in a vector space (usually  $\mathbb{C}^n$ .) That unit vector encodes information about that system's state.



The dimension  $n$  is, loosely speaking, the number of possible outcomes you can observe after making a measurement on the particle. If you have two quantum particles, the state of that two-particle system is described by something called a *density matrix*  $\rho$  on the tensor product of the vector spaces  $\mathbb{C}^n \otimes \mathbb{C}^n$ . A density matrix is a generalization of a state vector for interactions between the two particles.



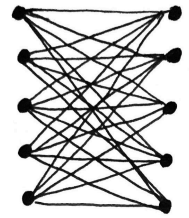
The same story holds for  $N$  particles—the state of an  $N$ -particle system is described by a density matrix on an  $N$ -fold tensor product of the vector spaces.



But why the tensor product? Why is it that this construction describes the interactions within a quantum system so well? We don't know the answer, but perhaps the appropriateness of the tensor product is not too surprising. The tensor product itself captures all ways in which two systems can interact.



"interact" with each other!



$V \otimes W$

tensor product

Of course, there's lots more to be said about tensor pr  
snippet of basic arithmetic. For a deeper look into the  
reading through Jeremy Kun's wonderfully lucid [How](#)  
[Tensorphobia](#) and [Tensorphobia and the Outer Produ](#)

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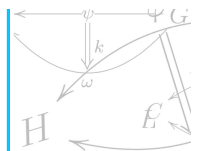
# A Quotient of the General Linear Group, Intuitively

December 15, 2016  
in [Algebra](#)

From a field  $F \rightarrow CV = \text{okay!}$  "vector" from a vector space  $V$   
but what if we want to multiply two vectors?  
? ? ?  
from another vector space  $U$   $u$  "times"  $v = \text{crazy!}$ ?  
what do we mean by "times"??  
**ANSWER: the TENSOR PRODUCT!**

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 4 years ago

Great article!

One thing that helped me conquer tensorphobia was the idea of bilinearity (or multilinearity more generally, I guess). I don't think it's the best way of first introducing the topic--this article's approach is probably best for that--but it helped make it clearer to me why tensors matter. I'll lay it out here in case it helps anyone.

First off, I think "bilinearity" is a misnomer. It makes it sound like it's sort of ultra-linear. Like, you have normal old linear maps, but then you have

maps, but then you have  
**\*bilinear\*** maps, which are  
**\*twice\*** as linear, somehow! But  
 no, bilinearity is a special case  
 of **\*non\***-linearity, not of  
 linearity. Namely, if you have a  
 product space  $V \times W$ , then a  
 bilinear map  $f : V \times W \rightarrow U$  out of  
 it satisfies:

$$\begin{aligned} f(v_1 + v_2, w) &= f(v_1, w) + f(v_2, w) \\ f(v, w_1 + w_2) &= f(v, w_1) + f(v, w_2) \\ f(a \cdot v, w) &= f(v, a \cdot w) = a \cdot f(v, w) \end{aligned}$$

As you can see, it's **\*like\*** the  
 conditions for linearity, but not  
 quite the same. If you hold the  
 right argument constant, you  
 have a linear map  $V \rightarrow U$ , and if  
 you hold the left argument  
 constant, you have a linear map  
 $W \rightarrow U$ . But if neither argument  
 is held constant, the map **\*isn't\***  
 linear. A linear map out of the  
 same space would satisfy:

$$\begin{aligned} f(v_1 + v_2, w_1 + w_2) &= f(v_1, w_1) + \\ &f(v_2, w_2) \\ f(a \cdot v, a \cdot w) &= a^2 \cdot f(v, w) \end{aligned}$$

As you can see, a linear map  
 acts more uniformly; the  
 conditions have to hold for both  
 conditions at once.  $f(ax, ay)$  is  
 $a \cdot f(x, y)$  for linear  $f$ , but it has to  
 be  $a^2 \cdot f(x, y)$  for bilinear maps;  
 you need to "pull out" the "a"  
 twice.

Okay, so anyway, bilinear maps  
 are a useful type of nonlinear  
 maps out of the product of two

vector spaces. For now, you kinda have to take my word for it that this class of maps is useful and interesting, but hopefully it shouldn't be too hard to believe. After all, linear maps are extremely useful, and bilinearity is linear in each of its arguments. It's like you're "encoding" two different classes of interrelated linear maps.

But, since bilinear maps fail to *actually* be linear as a whole, linear algebra isn't really equipped to deal with them directly! That's where tensor products come in. There is a one-to-one correspondence between BILINEAR maps  $f : V \times W \rightarrow U$  and LINEAR maps  $V \otimes W \rightarrow U$ ! (In fact, that's actually one way to *define* a tensor product in a more category-theoretic way; the "universal property" of the tensor product, at least in the category of vector spaces, is based on the preceding fact.) So, tensor products gives you a way to use the tools of linear algebra on bilinear maps, even though bilinear maps are nonlinear. Generalizing to multilinear maps, with higher-order products and tensor products, is pretty easy from here.

 5  0 Reply



...ti...

🕒 4 years ago

Thanks for sharing!



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Leon...



🕒 4 years ago

Thank you!



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Lisan...



🕒 4 years ago

Awesome !!! Thanks a bunch !!!



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Dinus...



🕒 3 years ago

great article. when i searched what is the tensor product for one of my undergraduate project there were lots of explanation which i cant understand. this is very good explanation and very easy to understand.



0



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Reply



Jan R...



🕒 3 years ago

Thanks for putting all this together :-)

 0  0 Reply



Danie...



 3 years ago

Love the article - it's really helping me familiarize myself with tensors ! Your content is beautifully put and very insightful. One question though, when you write the example of the tensor product of the vector spaces  $V$  and  $W$ , you write " $V \otimes W$  is the free vector space on  $A \times B$ , where  $A$  is a set of generators for  $V$  and  $B$  is a set of generators for  $W$ ". Should this be  $W$  ? Thanks !

 0  0 Reply



Tai...



 3 years ago

Whoops yes, fixed. Thanks!

 0  0 Reply



Abrah...



 2 years ago

Awesome article! Thanks in advance for your easy explanation.

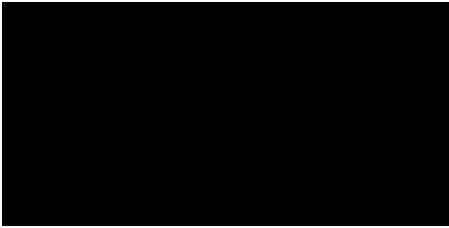
 2  0 Reply



hami...



 2 years ago



Amazing article. Thanks a lot